The EUROSIM Congress 2023, the 11th EUROSIM Congress, will be organized by DBSS, the Dutch Benelux Simulation Society, in Amsterdam, July 3-5, 2023.

This time, the EUROSIM Congress’s theme is Simulation for a sustainable future. And there is no better place than Amsterdam to celebrate the link between simulation and sustainability. You will be pleased to discover the fascinating and vibrant city of Amsterdam, a hub for many sectors such as logistics, ICT, art, financial and business services, and fashion, among others.

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Every year, the I3M Multiconference renovates the opportunity to bring together researchers, scientists and practitioners, from the Mediterranean Area, Latin & North Americas, Europe, Asia, Africa and Australia, who are concerned with Modeling and Simulation in Industry and Academia. As a result of the joint effort of many scientific excellences, I3M is an opportunity to meet and discuss and, as usually happens, constructive debate makes people open up, brings food for thoughts as well as new ways to explore encouraging multi-disciplinary cooperation and collaborations.

The Winter Simulation Conference 2023 highlights the vital role that simulation plays in designing, planning, and operating resilient systems under uncertainty. In an increasingly inter-connected world, it is more critical than ever to ensure that systems quickly recover from and adapt to major disruptions. With its uncertainty modeling and explainable analytics capabilities, simulation is one of the key technologies that lie at the heart of building resilient systems. We invite papers that emphasize the latest advances in simulation theory and applications showcasing the integrated use of simulation with technologies ranging from the Internet of Things and statistics to AI/ML and optimization. We particularly encourage applications of simulation to improve resiliency in a wide range of domains, including but not limited to aviation, disaster response, education, energy, finance, healthcare, infrastructure, manufacturing, national security, space systems, and supply chains.
Editorial

Dear Readers. This second issue of SNE Volume 33, 2023, underlines the astonishing broad area of modelling and simulation. Y. Shih’s contribution ‘Simulation of a Computational Model for the Performance Analysis of Mobile Police Bases in Apprehending Drug Dealers in the City of Itanhaém (São Paulo/Brazil)’ and the contribution ‘Raising Awareness of Climate Change Impact and Mitigation – The MyWorld Simulation Game’ by U. Beyer, M. Karnassnigg, and O. Ullrich deal with exciting applications: crime prevention and awareness of climate change impacts. Three other contrasting contributions present applications in biology, train scheduling, and software service architecture: ‘Modelling Blood Volume Oscillations in the Finger’ by H. Kastinger and M. Bachler investigates a new technology for blood pressure measurement, ‘Analysis of Train Delays using Bayesian Networks’ by V. Barzal et al. tries to find free slots for freight trains in dense train public transport systems, and ‘Service-Oriented Architecture for Automatic Evaluation of Urban Logistical Concepts’ by V. Ahlers et al. presents a framework for combining general traffic models with specialized logistical models into singular simulations. And we are glad, that D. Jammer, P. Junglas, and S. Pawletta have finalized a benchmark study for ARGESIM’s Parallel Benchmark CP2. The contribution ‘Solving ARGESIM Benchmark CP2 ‘Parallel and Distributed Simulation’ with Open MPI and Matlab PCT – Lattice Boltzmann Simulation’ continues and completes the benchmark contribution from SNE 32(4), which presented the Monte Carlo and PDE case studies with the same technique. The cover of this issue presents the second digital marbling graphics by Graham Horton, Professor for Simulation and Modelling at University Magdeburg, and ASIM-Member for many years, introduced in SNE 33(1). The title ‘Italian Sisters’ reflects a special technique for making marbled paper: two combs make the pattern.

I would like to thank all authors for their contributions for this issue, also and many thanks to the SNE Editorial Office for layout, typesetting, preparations for printing, electronic publishing, and much more. And we are glad that after the series of virtual conferences we can promote again EUROSIM face-to-face conferences: the EUROSIM Congress 2023, July 2023 Amsterdam, the conference ASIM SPL 2023, Ilmenau, in September 2023, 13M September 2023 in Athens, and others (see covers).

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SNE - Aims and Scope

Simulation Notes Europe (SNE) provides an international, high-quality forum for presentation of new ideas and approaches in simulation - from modelling to experiment analysis, from implementation to verification, from validation to identification, from numerics to visualization (www.sne-journal.org).

SNE seeks to serve scientists, researchers, developers and users of the simulation process across a variety of theoretical and applied fields in pursuit of novel ideas in simulation. SNE follows the recent developments and trends of modelling and simulation in new and/or joining areas, as complex systems and big data. SNE puts special emphasis on the overall view in simulation, and on comparative investigations, as benchmarks and comparisons in methodology and application. For this purpose, SNE documents the ARGESIM Benchmarks on Modelling Approaches and Simulation Implementations with publication of definitions, solutions and discussions. SNE welcomes also contributions in education in/for with simulation.

SNE is the scientific membership journal of EUROSIM, the Federation of European Simulation Societies and Simulation Groups (www.eurosim.info), also providing Postconference publication for events of the member societies. SNE, primarily an electronic journal e-SNE (ISSN 2306-0271), follows an open access strategy, with free download in basic version (B/W, low resolution graphics). Members of most EUROSIM societies are entitled to download e-SNE in an elaborate full version (colour, high resolution graphics), and to access additional sources of benchmark publications, models sources, etc. (via group login of the society), print-SNE (ISSN 2305-9974) is available for specific groups of EUROSIM societies.

SNE is published by ARGESIM (www.argesim.org) on mandate of EUROSIM and ASIM (www.asim-gi.org), the German simulation society. SNE is DOI indexed with prefix 10.11128. Author’s Info. Individual submissions of scientific papers are welcome, as well as post-conference publications of contributions from conferences of EUROSIM societies. SNE welcomes special issues, either dedicated to special areas and/or new developments, or on occasion of events as conferences and workshops with special emphasis.

Authors are invited to submit contributions which have not been published and have not being considered for publication elsewhere to the SNE Editorial Office.

SNE distinguishes different types of contributions (Notes), i.e.

- **TN** Technical Note, 6–10 p.
- **SN** Short Note, max. 5 p.
- **SW** Software Note, 4–6 p.
- **BN** Benchmark Note, 2–10 p.
- **ON** Overview Note – only upon invitation, up to 14 p.

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Simulation of a Computational Model for the Performance Analysis of Mobile Police Bases in Apprehending Drug Dealers in the City of Itanhaém (São Paulo/Brazil)

Yung-Chin Shih

Abstract. In this study, we developed a computational model to simulate the number of arrests made by police when using mobile police bases. For a simulation period of twelve months, the results showed that the use of more mobile bases did not always lead to more arrests. Moreover, it was observed that if an equal distribution of police officers might jeopardize the number of arrests. We proposed an unbalancing model in order to promote the police workload distribution. The simulation showed that the model can produce beneficial results with only a negligible imbalance.

Introduction

According to [3], Brazil is plagued by extreme violence, originating from multiple factors such as its poor economic condition, cultural differences, and wars between factions. For every 100,000 inhabitants, Brazil sees 27.1 homicides. Much of this violence can be attributed to organized crime linked to drug trafficking [11]. Therefore, curbing any form of trafficking, regardless of its nature, can help reduce the number of homicides.

As mentioned by [12], the implementation of the pacifying police unit (UPP) is strategic and is based on geoprocessing; it identifies regions with a high incidence of theft. However, as the world of crime is highly dynamic, many of the UPPs in Brazil have been replaced by mobile modules.

A mobile module (or mobile police base) is a vehicle that reinforces ostensive patrolling, bringing the police closer to the community, in addition to favoring the arrival of police officers when there is a crime-related incident.

According to [7], by patrolling through mobile police bases, the police can typically respond to drug-related incidences within a 2 km radius, while some authors argue that they can cover a greater distance. The module can remain at a certain location and even change its position in response to a specific incident.

In each mobile police base, a team patrols its surroundings and makes arrests if it identifies a crime. Police officers can perform patrolling in several ways, including in motorcycles, bicycles and even on foot. According to [2], the use of mobile police bases aims to reduce crime with low offensive potential.

According to [6], the city of Itanhaém (located on the coast in the state of São Paulo, Brazil) has the highest rate of exposure to violent crime. Hence, attempts to reduce such crime are prioritized by municipal authorities.

With this background, in this study, we developed a computational model to predict the performance of mobile police bases in apprehending illicit drug traffickers. The model runs using information for the city in question.

The remainder of this paper is divided into four sections. The first part describes the methods used in the research, methodology, and types of data collected. The model used in the simulation is discussed in the second part. The third part presents the results (obtained under balancing/unbalancing conditions), discussion of the results, and statistical analysis. Finally, we draw our conclusions.
1 Material and Methods

The method chosen for this study was a computational model, and the computational results obtained were quantitative, that is, the number of arrests, the size of the vulnerable area not covered by police bases, and the total cost incurred by the city hall. Because of the difference between the distribution of traffickers along the crowns as shown in fourth section, the performance of the arrests carried out by the police may be jeopardized, mainly due to the lack of police. Thus, an unbalanced model based on the workload was proposed. The results were compared with those obtained using a balanced model.

2 Model Police / Drug Dealer

2.1 Determining the coverage and vulnerable areas

Let \( N_{\text{current period}} \) mobile police bases be randomly located in the total area of the city, denoted by \( A_T \). The coordinates can be represented as \((x_1, y_1) \neq (x_2, y_2) \) \( \ldots \neq (x_N \in \text{N current period}, y_N \in \text{N current period}). \)

If the coverage radius of each mobile police base \( i \) is \( R_i \), the areas supervised by all the police bases are (1):

\[
OA_{\text{N current period}} = \sum_{i=1}^{N_{\text{current period}}} \pi \cdot R_i^2
\]  

As \( R_i \) can assume any positive value, an overlapping area may exist. If there are \( K \) overlapping areas with \( L \) police stations, the area occupied by police bases can be expressed as (2):

\[
OA_{\text{N current period}} = OA_{\text{N current period}} - \sum_{L=2}^{N} \sum_{ii=1}^{K} OLA_{L,ii},
\]

with \( L \forall i \) 

(2)

Here, \( OLA \) represents the overlapping area, and \( ii \) is the counter of \( K \) overlapping areas. For example, if \( L = 3 \) and \( K = 4 \), there are four occurrences in the three overlapping areas.

Thus, the remaining (vulnerable) areas represented by the \( RA \) variable can be expressed as follows (3):

\[
\left\{ \begin{array}{l}
RA_1 = A_T - OA_{\text{N current period}} \quad \text{without overlapping area} \\
RA_2 = A_T - OA_{\text{N current period}} \quad \text{with overlapping area}
\end{array} \right.
\]  

(3)

If \( RA \) (both) is the region farthest from police stations, i.e., the preferred area for criminals, then the objective is to minimize \( RA \).

The focus here is on dealing with nonoverlapping areas covered by the police bases; then, we can represent \( RA_1 \) through expressions (4), (5), and (6):

\[
\min RA_1 = A_T - \max \{OA_{N_{\text{current period}}}\},
\]

where \( 0 < \max \{OA_{N_{\text{current period}}}\} < A_T \) to avoid \( RA_1 \) becoming negative

\[
\min RA_1 = A_T - \pi \cdot \max \{\sum_{i=1}^{N_{\text{current period}}} R_i^2\},
\]

with \( \pi \cdot \max \{\sum_{i=1}^{N_{\text{current period}}} R_i^2\} \leq A_T \)

\[
\min RA_1 = A_T - \pi \cdot \{\max R_1^2 + \max R_2^2 + \cdots + \max R_{N_{\text{current period}}}^2\}
\]  

(6)

From these expressions, it can be noted that while the coverage radius are different, it is not possible to maximize each \( R_i \). To achieve the highest value of \( R_i \), it is necessary for all of them to be equal by positioning them as equidistantly as possible between the police bases. Thus, the expression can be summarized as (7):

\[
\{ \min RA_1 \} = A_T - \max N_{\text{current period}} \cdot \pi \cdot R^2
\]

(7)

2.2 Equidistant positioning of police bases

To calculate \( RA_i \), two unknown variables must be defined that are mutually dependent. In the case of equidistant positioning of police bases, the variables are inversely proportional \( (N_{\text{current period}} \text{ and } R) \).

Once \( N_{\text{current period}} \) is defined, it is possible to determine \( R \) and the desired value of \( RA_i \).

For a given value of \( N_{\text{current period}} \), with the expression proposed by [1], called the distribution degree \( \phi \) shown in (8), the shortest distance between equal objects can be as large as possible. The unit of measurement was distance.

The authors used this expression to define the positions of machines in a physical arrangement study on factories. The lower the degree of distribution, the better the result; that is, the more equidistant the objects. Similarly, police bases (equal objects) can be positioned equidistantly.

To obtain one of the variables described in (8), we need to solve (9):

\[
\phi = \sum_{j=1}^{N_{\text{current period}}} \sum_{n_{j=1}}^{N_{\text{current period}}} \frac{\delta_{n_j}}{N_{\text{current period}} N_{\text{current period}}}
\]

\[
\delta_{n_j} = \sum_{k \neq j}^{N_{\text{current period}}} d_{n_j k}
\]

(9)
Here $d_{j,k}^*$ is the distance between the $n$th object of type $j$ and the closest object of type $k$; $n_j = n$th object type $j$; $NO_j$ is the number of objects type $j$; $NO$ represents the types of objects.

In this study, $NO$ is equivalent to $N_{current\ period}$. To computationally obtain the distribution degree, the authors argue that metaheuristics should be used to obtain the degree of distribution. This is because the computer tests different positions that each police base can occupy, thus reducing the computational search time. The distribution degree is calculated for each police base allocation.

[8] developed a genetic algorithm (GA) to determine the distribution degree. When performing numerous trials (generations in the GA), the degree of distribution that presents the lowest degree of distribution is the most equidistant possible placement between objects. Therefore, in this study, we developed a GA. Because there are only two types of objects (police bases and positions not occupied by police bases), it is not necessary to perform many generations in this case.

Finally, it is possible to calculate the distance between police bases and consequently obtain the desired circular radius (10):

$$ R = \frac{1}{2} \cdot \min \{ SQRT[(x_{i=1} - x_{i=2})^2 + (y_{i=1} - y_{i=2})^2]; $$

$$ SQRT[(x_{i=1} - x_{i=3})^2 + (y_{i=1} - y_{i=3})^2]; \ldots; $$

$$ SQRT[(x_{i=n_{current\ period} - 1} - x_{i=n_{current\ period}})^2 + (y_{i=n_{current\ period} - 1} - y_{i=n_{current\ period}})^2] \} \quad (10) $$

### 2.3 Determining the number of police officers and the number of drug dealers

The total number of police officers working in a region is considered. Using $Y\%$ (the percentage of police officers working in regions covered by police bases), we can calculate the number of police officers per police base ($n_{police}$). Therefore, the difference $(1 - Y)\%$ corresponds to the effective personnel working in vulnerable regions, under the condition that $Y\% > (1 - Y)\%$.

For the total number of traffickers operating in the region, if $X\%$ operate in vulnerable areas, the remaining $(1 - X)\%$ will operate in coverage areas. These established inequalities are due to the fact that police officers are at a greater risk in more vulnerable areas. However, it is not convenient for traffickers to operate close to police bases because these areas are a greater risk to drug dealers.

For a given number of mobile bases $N$, it is possible to estimate the number of traffickers for each mobile base $i$, represented by $n_{drug\ dealers}$.

The coverage area of each police base $i$ was divided according to the number of circular crowns adopted. For example, if there are two crowns, the height of each crown is $R/3$. In addition to the two circular crowns, there will be an internal circle with a radius of $R/3$. It can simply be called a circular crown. In these cases, the area of the circular crown will always be greater than that of the adjacent interior.

The number of police officers at each police base was equally distributed by each circular crown. For traffickers, the preferred active areas are external crowns.

The initial positions of each police officer and each drug dealer are different from each other and randomly established. For each police base, we established an arrest performance target to be used in decisions regarding hiring police officers. The target is defined as the expected number of arrests $EA$.

### 2.4 Determining the number of arrests and calculating the total cost

As the simulation progressed, new values of $N$ were obtained, which in turn changed the configuration of police base positions in the $A_f$ area. This is because, if the number of arrests reaches a certain number, physical structures with a mobile police base will be required.

The number of arrests carried out on a given day depends on the movement of traffickers and police. However, their movements differed. In the case of the drug dealer, the subject performed move–stop exercises.

Considering the time for which the dealer remained in the same location, this can be represented by the following equation (11):

$$ \frac{s_1}{v_1} + \frac{s_2}{v_2} + \ldots + \frac{s_n}{v_n} = WJDD - n, \ t_{standing\ up} $$

where $WJDD$ represents the work journey of the drug dealer; $n$ represents the number of dealers; $s$ represents the distance traveled by each trafficker; $v$ represents the displacement speed of each trafficker; and $t$ represents the time spent at the site.

If the speed is constant throughout the path taken and represents all the distances by $S$, then the expected number of locations reached on the workday is (12):

$$ \frac{N_{locations\ drug\ dealer}}{WJDD} = \frac{\bar{S}_{drug\ dealer}}{v_{drug\ dealer}} + t_{standing\ up} \quad (12) $$
In the case of a police officer, movement is interrupted only to apprehend a drug dealer. Hence, we omitted the stop time. The expression is shown in (13):

$$\frac{s_1}{v_1} + \frac{s_2}{v_2} + \cdots + \frac{s_n}{v_n} = WJPO$$

(13)

where \(WJPO\) is the work journey of the police officer; \(n\) represents the number of police officers; \(s\) represents the distance traveled by each officer; \(v\) represents the displacement speed of each officer.

If we assume that the displacement speed of the police officer is constant and if we represent all distances covered by \(\bar{s}\), then the expected number of locations that can be reached by the police officer can be expressed by (14):

$$N_{locations \ police \ officer} = WJPO / (\bar{s}_{police \ officer} / v_{police \ officer})$$

(14)

When both the locations are identical, an arrest occurred. Due to the arrests, the number of drug dealers is subtracted. The police officer sends a delinquent to a mobile base for registration and other administrative activities. Therefore, the on-site actions of the officer become inactive. Activity is resumed only the following day. The subtracted number of traffickers is immediately replenished the following day.

Therefore, the ratio of the number of police officers to the number of drug dealers can be expressed as (15):

$$f = \frac{N_{locations \ police \ officer}}{N_{locations \ drug \ dealer}}$$

(15)

For example, if the result of \(f\) is eight, this indicates that there are eight police movements reaching eight different locations against one trafficker movement (reaching one new location). The variable \(f\) corresponds to the number of loops to be executed by the computational model to determine new police positions based on the number of positions reached by the traffickers. Thus, for each \(i\), if the positions of the dealer and police are the same, then an arrest is made (\(A_i\) in the next period).

For the entire month, the total cost can be calculated as follows (16):

$$TC = PC \cdot \sum_{i=1}^{N} n_{i \ police} + N \cdot SPC$$

(16)

where \(TC\) represents the total cost; \(N\) represents the number of mobile bases; \(n\) represents the number of police officers in base \(i\); \(PC\) represents the cost incurred in the deployment of each police officer; and \(SPC\) represents the monthly cost to maintain the physical structure of each mobile base.

The sum of officers from all mobile police bases must be calculated because this information is used to determine the number of officers from each mobile base in the next period. The total number of police officers operating in the coverage areas is (17):

$$Total \ number \ of \ police \ officers_{current \ period} = \sum_{i=1}^{N_{current \ period}} n_{i \ current \ period}$$

(17)

### 2.5 Updating the number of police officers and drug dealers for the next period

The updated number of police officers for the following month is based on the expected number of arrests.

The number of police officers required for each mobile base \(i\) in the next period can be calculated using Equations (18), (19), and (20). If the number of arrests in the current period is higher than the expected number, the difference must be added to the current number of police officers (18). If the number of arrests in the current period is lower, the difference must be subtracted (19). However, the number of police officers remains the same (20):

- If \(A_i \ current \ period > EA_i\) then \(n_{i \ next \ period} = n_{i \ current \ period} + [A_i \ current \ period - EA_i]\)
- If \(A_i \ current \ period < EA_i\) then \(n_{i \ next \ period} = n_{i \ current \ period} - [EA_i - A_i \ current \ period]\)
- If \(A_i \ current \ period = EA_i\) then \(n_{i \ next \ period} = n_{i \ current \ period}\)

We established the total number of police officers required for the next period (21) as follows:

$$Total \ number \ of \ police \ officers_{next \ period} = \sum_{i=1}^{N_{current \ period}} n_{i \ next \ period}$$

(21)

The difference \(d\), expressed in Equation (22), between the number of police officers in the current period and that in the next period determines whether there is a need to change the number of mobile bases. When the cost of the required police officers is greater than the cost of the mobile base, more mobile bases are added:

$$d = Total \ number \ of \ police \ officers_{next \ period} - Total \ number \ of \ police \ officers_{current \ period}$$

(22)

Thus, we can establish some rules, as expressed in equations (23) and (24).
If the difference is positive,

\[ d \cdot PC \geq SPC + n_i \text{current period} \cdot PC \]

\[ N_{\text{next period}} = N_{\text{current period}} + 1 \quad (23) \]

If the difference is lower than or equal to zero, then

\[ |d| \cdot PC \geq SPC + n_i \text{current period} \cdot PC \]

\[ N_{\text{next period}} = N_{\text{current period}} - 1 \quad (24) \]

With the new value of \( N_{\text{next period}} \) determined, we can obtain \( R \) and consequently the values of \( RA_1 \) and so on.

Regarding the number of traffickers, the integration of new traffickers occurs daily. This implies that the total number of traffickers is always fixed.

### 3 Simulation of the Model

#### 3.1 Definition of the input parameters and discussion of results

The city of Itanhaém (in São Paulo) has an estimated area of 600 km² according to [9]. [4] recorded that the average number of traffickers operating throughout the city was 1000; in coverage areas, \( Y\% = 75 \), and in vulnerable areas, there were 75 police officers.

The available software for building such a computational model is Dev-Pascal (Pascal language); the simulation was done on a PC with i3-10100 CPU, 8 cores, 3.6 GHz, 4.1 GB of RAM.

Based on the information obtained from the aforementioned city, in this study, the dimensions were set to 25 km \( \times \) 25 km; the total number of available police officers in the city was 250 and updated monthly. The total number of traffickers operating throughout the city was fixed at 1000. The initial number of mobile police bases was 13 (\( N_{\text{initial}} = 13 \)), and \( Y\% \) was taken as 70%. Therefore, the number of police officers working in mobile bases was 250; \( Y\% = 175 \), and in vulnerable areas, there were 75 police officers.

The number of traffickers with \( X\% = 80 \%) \) that are active in vulnerable areas was 1000; in coverage areas, \( X\% = 800 \), and the number of traffickers was 200. Thus, per police base, the estimate was

\[ n_i \text{police officer} = \text{trunc} \left( \frac{175}{N_i \text{initial}} \right) \]

#### Simulation of the Model

#### Discussion of results

With the new value of \( N_{\text{next period}} \) determined, we can obtain \( R \) and consequently the values of \( RA_1 \) and so on.

Regarding the number of traffickers, the integration of new traffickers occurs daily. This implies that the total number of traffickers is always fixed.

The number of traffickers acting per police base could be estimated through

\[ n_i \text{drug dealers} = \text{trunc} \left( \frac{200}{N_i \text{initial}} \right) \]

The dealer distributions for the three circular crowns (interior to exterior) were 0%, 20%, and 80%. The trafficker’s working day was established using the relationship

\[ WJDD = 3h; \quad v_{\text{drug dealer}} = 0.7 \text{ m/s}; \]

\[ \bar{s}_{\text{drug dealer}} = 10 \text{ m}; \quad t_{\text{standing up}} = 0.25 \text{ h}; \]

\[ v_{\text{police officer}} = 35 \text{ km/h}; \quad \bar{s}_{\text{police officer}} = 40 \text{ m}. \]

\( PC \) corresponds to the salary and fuel cost per month = R$5000.00; \( SPC \) corresponds to fuel and van rental per month = R$20,000.00; \( EA_1 = 40\% \) de \( n_{\text{police}} \).

To run the GA, we adopted five generations with 20 individuals in the population, a probability of 80% for crossover, and 2% for mutation.

Table 1 presents the simulation results. The first row shows the performance when 13 mobile police bases were used, resulting in 359 arrests. The outermost circular crowns saw a greater number of arrests. The vulnerable area not covered by police bases was 461.64 km², with a total cost of R$1105K.

The results showed that the installation of more police bases cannot always increase the number of arrests, as was the case between months 4 and 8.

By separately analyzing the results of the two groups, we found that the number of arrests in the case of more mobile bases (from 10 to 13) averaged at 723.25, and the number of arrests in the case of fewer bases (7 to 9) averaged at 768.5. From this, we can conclude that, in general, a greater number of mobile bases does not necessarily result in a greater number of arrests.

The results showed that working with more police bases does not always result in a lower vulnerable area.

By observing the three-performance metrics, namely the “total arrests,” “vulnerable area,” and “total cost,” it is not possible to reach the best value considering all the three metrics. Notably, the reduction in the number of police bases seems to have a strong influence on the total cost reduction, as was the case in months six to seven as well as in months three to four.

Statistically, if the number of traffickers at the beginning of each month is defined as 1000, and if we consider that 80% of months should have more than 1000 arrests, it indicates that the number of circular crowns used produces a satisfactory performance.
Statistically, if the number of traffickers at the beginning of each month is defined as 1000, and if we consider that 80% of months should have more than 1000 arrests, it indicates that the number of circular crowns used produces a satisfactory performance.

We observed that three of the 12 months saw arrests above 1000. Therefore, the value of \( p \) was \( 3/12 = 0.37 \).

So we can formulate the following hypotheses for the statistic \( p \):

- \( H_0: \ p \geq 0.8 \), if 80% of the months have a performance equal to or greater than 1000 arrests, \( \text{(25)} \)
- \( H_1: \ p < 0.8 \) otherwise.

The value of \( Z_{\text{test}} \) used in the comparison is

\[
Z_{\text{test}} = \left( \hat{p} - p \right) / \sqrt{\frac{p(1-p)}{n}}
\]

The \( Z_{\text{test}} \) value was \(-3.72\). For a level of significance of 5%, because \(-3.72\) is lower than \( Z_{\alpha=5\%} = -1.64 \), the decision was to reject \( H_0 \).

### 3.2 Adaptation of the simulation model with implementation of imbalance

#### Imbalance per area

Note that when dividing the circumference and equally distributing the number of police in \( n \) crowns, the number of arrests in the internal crown was zero, and drug dealers were not involved.

In this study, traffickers preferred external crowns farthest from the base. This led to the idea that police officers should be reallocated differently.

This paper presents two unbalanced procedures. First, police officers are proportionally allocated according to the crown area.

Let \( p' \) be the probability that a police officer is located in a given circular crown. If the area increases, then to maintain the same \( p' \), a larger number of police officers should be assigned \( \text{(27)} \).
The steps are as follows:

- Calculate the total coverage area of base \( i \).
- Calculate the percentage of each circular crown \( p' \);
- Calculate the number of police officers per ring crown using the following equation:

  \[
  \text{Police officers in the circular crown} = n_i \text{police officer} \cdot p'
  \]  

  (27)

Table 2 shows the results of the imbalance model. Clearly, a different number of police officers should be allocated to each crown to ensure a greater number of arrests. Surprisingly, the performance in the case of four circular crowns approached that in the case of three circular crowns. Thus, the variation in the number of circular crowns with unbalancing does not seem to jeopardize the arrest performance.

<table>
<thead>
<tr>
<th>Total arrests</th>
<th>Month</th>
<th>Balanced</th>
<th>Area-based imbalance</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1</td>
<td>359</td>
<td>552</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>1157</td>
<td>508</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>1023</td>
<td>1007</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>354</td>
<td>1591</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>783</td>
<td>3187</td>
</tr>
<tr>
<td></td>
<td>6</td>
<td>706</td>
<td>936</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>863</td>
<td>981</td>
</tr>
<tr>
<td></td>
<td>8</td>
<td>1015</td>
<td>741</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>893</td>
<td>1834</td>
</tr>
<tr>
<td></td>
<td>10</td>
<td>868</td>
<td>348</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>600</td>
<td>841</td>
</tr>
<tr>
<td></td>
<td>12</td>
<td>420</td>
<td>378</td>
</tr>
<tr>
<td></td>
<td>Total</td>
<td>9041</td>
<td>12904</td>
</tr>
</tbody>
</table>

Table 2: Comparison between the results obtained under balancing and unbalancing conditions by areas.

We note that the unbalancing model outperforms the balancing model.

Applying the statistical test to the unbalanced results, we find that the \( p \) value was \( 4/12 = 0.33 \). The value of \( Z_{act} \) was \( -4.07 \), resulting in the rejection of the hypothesis \( H_0 \).

We conclude that, although the number of arrests improved overall, it still did not satisfy the monthly performance conditions.

During the simulation, with the change in the number of mobile bases, the number of traffickers per police base also changed. This means that, for the current approach, the number of police officers remain unchanged, consequently reducing the number of arrests.

This suggests that a greater workload should be assigned to police officers in the most distant crowns. This is the method proposed in this study.

**Imbalance per workload**

The results suggest that a greater workload should be assigned to police officers in the most distant crowns. This is the method proposed in this study, Equation (28):

\[
\beta = \frac{n_i \text{police officers} \cdot W \cdot \rho}{\sigma + \sum_{i=0}^{\Delta} i \cdot \alpha}
\]  

(28)

where \( \beta \) corresponds to the workload of the outermost crown; \( \alpha \) corresponds to the imbalance factor; \( \Delta \) corresponds to the number of circular crowns.

Once this value was determined, the other values were assigned according to the circular crown. The closer to the mobile police base, the lower the workload, indicating fewer police officers.

The workload \( CHS \) (29) of the \( \epsilon \)th crown counted from the base can be expressed as follows:

\[
CHS = \beta + (\Delta - \epsilon) \cdot \alpha \cdot \beta
\]  

(29)

Note that in Equation (28), \( \alpha \) depends on the number of police officers operating in each police base (which varies temporally), \( \beta \) value, the number of circular crowns, and police’s work journey. This means it is not possible to establish a fixed value of \( \alpha \).

To define the required range of \( \alpha \) for the simulation, we use the input parameters under the condition that \( \beta \geq 0 \). If the police officer workday is 8 hours, then we can deduce that for \( \beta \geq 8 \), we must use \( \alpha > -1 \), in addition to \( \alpha \neq 0 \). Only negative values were used because of the higher workload in the external crown. Therefore, the \( \alpha \) values for the three crowns were \(-0.9, -0.45, -0.15\).

Table 3 shows the results. The benefits can be verified when a higher unbalancing factor (close to balanced \( \alpha = 0 \)) is used.

As the value decreases (the greater the unbalancing), the benefits become evident, confirming that the unbalancing should be based on the workload.
So results are:
- For \( \alpha = -0.15 \), \( p = 4/12 \),
- for \( \alpha = -0.45 \), \( p = 10/12 = 0.83 \),
- and for \( \alpha = -0.9 \), \( p = 12/12 = 1 \).

The corresponding \( Z_{\text{act}} \) values were calculated to be 4.07, 0.26, and 1.73, respectively.

Table 4 presents a comparison with the reference \( z \) at a significance level of 5%.

<table>
<thead>
<tr>
<th>Months of the year</th>
<th>( \alpha = -0.15 )</th>
<th>( \alpha = -0.45 )</th>
<th>( \alpha = -0.9 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Crown 1</td>
<td>0 0 0 0 0 0 0 0 0 0 0 0</td>
<td>0 0 0 0 0 0 0 0 0 0 0 0</td>
<td>0 0 0 0 0 0 0 0 0 0 0 0</td>
</tr>
<tr>
<td>Crown 2</td>
<td>121 365 381 158 544 190 103 138 227 250 312 68</td>
<td>139 343 268 149 317 501 219 199 247 285 229 457</td>
<td>166 211 703 296 529 353 853 885 437 441 468 499</td>
</tr>
<tr>
<td>Crown 3</td>
<td>413 936 981 543 1298 516 376 402 582 696 857 230</td>
<td>506 1274 1096 816 1238 1811 1108 995 1042 1087 1073 1656</td>
<td>2259 3004 4620 3923 4211 3499 4311 4557 4188 4399 4584 4118</td>
</tr>
<tr>
<td>Unbalanced</td>
<td>534 1301 1362 701 1842 706 479 540 809 946 1169 298</td>
<td>645 1617 1364 965 1555 2312 1327 1194 1289 1372 1302 2113</td>
<td>2425 3215 5323 4219 4740 3852 5164 5442 4625 4840 5052 4617</td>
</tr>
</tbody>
</table>

Table 3: Results with unbalancing.

Table 4: Statistical comparison with unbalancing.

4 Conclusions

In this study, a computational model was constructed for effective policing against drug trafficking, with the objective of simulating performance projections in cases where mobile police bases were implemented, aiming to increase the number of arrests, reduce the number of vulnerable areas, and reduce associated costs.

Using the computational model constructed for this purpose, it was possible to obtain the desired results for a performance comparison. The computational model proved to be versatile, as it could simulate possible changes in the positions of mobile bases, police officers, and particularly drug dealers.
The results showed that when there were areas with police supervision with a low circulation of traffickers, the expected result should be zero arrests, which was verified. This finding suggested that an unbalanced model can be used to reallocate officers in a different manner. When considering the imbalance by the area size (i.e., the larger the area, the higher the number of police officers, or vice versa), more arrests could be made.

Despite the benefits of unbalancing by area size, redistribution might fail. This is because, as the simulation progresses, the numbers of drug dealers and police officers might change. Hence, an unbalanced model was proposed according to the workload of the police officers. With the proposed model and using an intermediate lower value of the unbalancing factor, the obtained results were more promising.

As a proposal for future work, some restrictions could be considered, such as in areas that are difficult for police to access and are already dominated by armed traffickers. This prevents the installation of mobile bases in such locations. As a suggestion for future work, due to the limitations of the software used in this study, it was not possible to analyze the arrests made in areas not covered by police bases. As stated by [10], it is in the peripheries that crime is more organized, suggesting a way that aims to maintain a protection network and eliminate possible competitors. We also suggest the development of new unbalanced models in future studies.

References


Abstract. The aim of this work is to describe a model of the relationship between blood volume and transmural pressure of a vessel of the finger, where the finger is pressed against a surface with increasing pressure. The focus lies on the amplitudes of the blood volume oscillations, as they can be measured easily. The model leads to a formula for the blood volume oscillation amplitudes. This equation is then fitted to measured data to estimate systolic and diastolic blood pressure. The focus lies on the fitting process and the quality of the fitting results.

Introduction

This work introduces and describes a model of the relation between blood volume and transmural pressure of a vessel of the finger which is pressed against a surface. More of interest than the absolute volume are in fact the blood volume oscillations, as they can be measured easily using an optical sensor, with respect to changes in the transmural pressure. This model for the blood volume oscillations shall be used for estimating the blood pressure without a cuff at the finger by fitting the model function to measured data and obtaining the systolic and diastolic blood pressure as fitting parameters.

1 The Model

The idea of the model is to describe a relation between blood pressure inside a vessel, pressure acting on the vessel from outside and the blood volume inside the vessel. The vessel of interest, the transverse palmar arch artery, is located in the finger tip and the situation of interest is the fingertip being pressed on some surface with increasing pressure. This approach can be compared to the blood pressure measurement with a cuff around the upper arm as there is also an external pressure acting on the vessel. Similar to the finger pressing on a surface, the cuff is exerting decreasing pressure on the vessel from outside, while pressure oscillations are measured. First, let us take a look at the absolute volume of the vessel and afterwards at the blood volume oscillations in dependency of the external pressure acting on the vessel from outside.

1.1 The Vessel Volume

The transverse palmar arch artery lies parallel to the surface of the finger and the bone underneath. Let us assume the artery as a cylindric tube of length $L$. When the finger presses on a surface, the pressure also acts onto the wall of the vessel. This external pressure ($P_{\text{ext}}$) is directed inwards. [1] The second relevant component is the blood pressure ($P_b$). This is the pressure applied on the vessel wall by blood running through the arteries. It is directed opposite to $P_{\text{ext}}$, in direction of the outer normal vector. Hence, the pressure acting on the vessel wall is a combination of $P_b$ and $P_{\text{ext}}$. This combination is defined as transmural pressure $P_t$ as follows:

$$||P_t|| := ||P_b|| - ||P_{\text{ext}}||.$$ 

A positive transmural pressure $P_t$ means that the blood pressure $P_b$ exceeds the external pressure $P_{\text{ext}}$. This leads to the vessel being pressed apart. A negative transmural pressure $P_t$, on the other hand, means that $P_{\text{ext}}$ exceeds $P_b$ and therefore the vessel is compressed. [1]

To find a relation between the transmural pressure $P_t$ and the blood volume $V$ of the artery, $A$ is defined as the cross-sectional area of the artery. First we are looking at the artery in relaxed state: $P_t = 0$. In this case, we assume $A$ as a circular disc with radius $R \in \mathbb{R}$. 

In the second case we are looking at is $P_l < 0$. In this case, the artery is compressed. Figure 1 shows the cross-section with side length $a \in \mathbb{R}^+$ and radius of the semicircles $0 \leq r \leq R$. \[1\]

The arterial wall is not stretched. We want to use this fact to find a relation between $R$, $r$ and $a$. Let $\gamma_0$ be the boundary of $A$, the cross-section of the artery in relaxed state, then $L\gamma_0 = 2R\pi$. Let $\gamma$ be the boundary of the deformed cross-section, then $L\gamma = 2r\pi + 2a$.

\[
\begin{align*}
L\gamma &= L\gamma_0 \quad (1) \\
2r\pi + 2a &= 2R\pi \quad (2) \\
a &= \pi(R - r) \quad (3) \\
\Rightarrow a &\in [0, R\pi]. \quad (4)
\end{align*}
\]

If the artery is relaxed then $a = 0$ and if $a = R\pi$ the vessel is fully occluded. \[1\]

The next step is to find a relation between $r$ and $P_l$. Let us define $r(t) := R(\sqrt{1 - e^{\alpha t}})$, $P_l < 0$, inspired by Charles. F. Babbs [3]. The parameter $\alpha > 0$ is related to the stiffness of the vessel wall. Now we can take a closer look at $V$.

\[
\begin{align*}
V(P_l) &= L\gamma(P_l) = L(r(P_l)^2\pi + 2r(P_l)a) = (5) \\
&= LR^2\pi e^{\alpha P_l} = V_0 e^{\alpha P_l}, \quad (6)
\end{align*}
\]

where $V_0 := LR^2\pi$ is the volume of the relaxed vessel. We can see that a negative transmural pressure leads to an exponential reduction in blood volume. \[1\]

The next step is to look at the case $P_l > 0$. The wall is stretched uniformly in angular direction by the pressure. Hence, the cross-section $A$ forms a circle with radius $r > R$. Let us define $r$ in dependency of $P_l$ based on Charles F. Babbs [3] as follows:

\[
r(P_l) := R\sqrt{1 + \frac{\alpha}{\beta} (1 - e^{-\beta P_l})}. \quad (7)
\]

The parameter $\beta > 0$ is also related to the stiffness of the arterial wall and is inversely proportional to the elasticity of the vessel wall. Again we want to take a look at the volume $V$:

\[
\begin{align*}
V(P_l) &= LA(P_l) = L(r(P_l)^2\pi = \\
&= LR^2\pi(1 + \frac{\alpha}{\beta} (1 - e^{-\beta P_l})) = \\
&= V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta P_l})). \quad (8)
\end{align*}
\]

\[1\]

All together we can define the vessel volume $V$ as follows:

**Definition 1.1 (Vessel volume; P. Baumann [1], p. 42)**

Let $L > 0$ be the length of the cylindric shaped transverse palmar arch artery and $R > 0$ the radius of the cross-section in a relaxed state ($P_l = 0$). Furthermore, let $\alpha, \beta > 0$ be fixed parameters describing the vessel wall elasticity and assume the deformation behaviour described previously. Then, $V$ is given by

\[
V : \mathbb{R} \rightarrow \mathbb{R}, P_l \mapsto \begin{cases} \\
V_0 e^{\alpha P_l}, & P_l < 0 \\
V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta P_l})), & P_l \geq 0.
\end{cases} \quad (9)
\]

With this definition we can see that $V \in C^1(\mathbb{R})$ and that it is monotonically increasing. Figure 2 shows the function of the vessel volume $V$ for estimated values for $\alpha$ and $\beta$. \[1, 3\]

![Figure 1](image1.png)

**Figure 1:** Schematic representation of the cross-section of the artery for a transmural pressure $P_t < 0$ [2].

![Figure 2](image2.png)

**Figure 2:** $V$ for $\alpha = 0.11$ and $\beta = 0.03$. 
1.2 The Oscillation Curve

The blood pressure $P_b$ is not constant. In contrary, it varies between a minimum (diastolic, $P_{dia}$) and a maximum pressure (systolic, $P_{sys}$). If the external pressure increase at the finger is sufficiently slow, we can assume that $P_{ext}$ is constant during each heartbeat. That means that the change of $P_t$ is determined just by the variance of the blood pressure $P_b$. Furthermore, the differences of $P_t$ lead to a varying amount of blood volume. Hence, $V$ oscillates between its maximum and minimum during each cardiac cycle. Since $V$ is monotonically increasing with $P_t$, it attains its minimum and maximum at the minimum and maximum of $P_b$. [1]

To make the next step let us first define pulse pressure $PP := P_{sys} - P_{dia}$. Now, we can define the maximum difference of blood volume $\Delta V$ during one heartbeat as follows:

$$\Delta V(P_m) := V(P_m + PP) - V(P_m)$$

$P_m$ denotes the minimal transmural pressure which can also be calculated as $P_m = P_{dia} - P_{ext}$. [1]

The next step is to establish a connection between systolic and diastolic blood pressure and $\Delta V$. First we will look at the case $P_m \leq -PP$:

$$\Delta V(P_m) = V_0 e^{\alpha (P_m + PP)} - V_0 e^{\beta P_m}$$

The next case is $P_m \in (-PP, 0)$:

$$\Delta V(P_m) = V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta (P_m + PP)})) - V_0 e^{\alpha P_m}$$

Last, we will take a look at the case $0 \leq P_m$:

$$\Delta V(P_m) =$$

$$= V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta (P_m + PP)})) - V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta P_m}))$$

[1]

As already mentioned above, $P_m = P_{dia} - P_{ext}$.

Therefore

$$P_m + PP =$$

$$= P_{dia} - P_{ext} + P_{sys} - P_{dia}$$

$$= P_{sys} - P_{ext}$$

Finally, we can combine these results and formulate an equation for $\Delta V$:

$$\Delta V(P_{ext}) =$$

$$\begin{cases} V_0 \cdot e^{\alpha (P_{sys} - P_{ext})} - \alpha \cdot V_0 \cdot e^{\alpha (P_{dia} - P_{ext})}, & P_{ext} \geq P_{sys} \\ V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta (P_{sys} - P_{ext})}) - \alpha \cdot V_0 \cdot e^{\alpha (P_{dia} - P_{ext})}, & P_{dia} < P_{ext} < P_{sys} \\ V_0(1 + \frac{\alpha}{\beta} (1 - e^{-\beta (P_{sys} - P_{ext})}) - V_0(1 + \frac{\alpha}{\beta} e^{-\beta (P_{dia} - P_{ext})}), & P_{ext} \leq P_{dia}. \end{cases}$$

Figure 3 shows the oscillation amplitude curve $\Delta V$ for an assumed systolic and diastolic pressure of 120mmHg and 80 mmHg and $\alpha = 0.11, \beta = 0.03$. [3, 4]

![Figure 3: $\Delta V$ for $P_{sys} = 120\text{mmHg}, P_{dia} = 80\text{mmHg}$ and $\alpha = 0.11, \beta = 0.03$.](image)

2 Estimating Blood Pressure via Model Fitting

This model, the formula for the blood volume oscillations $\Delta V$ in particular, in this form can be used for estimating blood pressure. To do so, equation 14 can be used as model function and be fitted to blood volume oscillations measured from the transverse palmar arch artery.
2.1 Measuring of the Blood Volume Oscillations

The blood volume oscillations ($\Delta V$) are measured alongside the contact pressure ($P_{ext}$) by the smartPWA device. The smartPWA (smart Pulse Wave Analysis) device is a bio-signal acquisition sensor device specially developed and built by the AIT Austrian Institute of Technology (Vienna, Austria) for research projects. The device is intended to be held by the user with both hands, like a gamepad, as shown in Figure 4. The user’s right index finger touches an optical sensor for photoplethysmography (PPG) to measure blood volume changes in the finger’s microvascular bed. Below the PPG sensor is a pressure sensor that registers the contact pressure of the right index finger.

The PPG, and pressure signals are converted from analogue to digital signals at 256 Hz and 24-bit resolution. Communication with a smartphone or tablet computer is established via Bluetooth low energy (BLE) and the measured signals are streamed continuously to a mobile app for further data processing.

Let us take a closer look at the measurement of the blood volume oscillations. The finger pressure on the sensor is increased steadily by the user up to 200 mmHg over 30 seconds and at the same time the blood volume oscillations are measured with the PPG sensor. To obtain the best possible measurement, the contact pressure $P_{t}$ between finger and sensor should increase linearly with respect to time $t$ and the slope should not be steeper than $7 \frac{mmHg}{s}$. Unfortunately, the pressure sensor currently in use is not able to detect pressure below 50 mmHg and therefore the pressure detection starts only at this value. Figure 5 shows an example for the contact pressure in orange.

To remove high frequency noise and slow signal drift, and enhance the amplitude of the volume oscillations, the measured signal from the photoplethysmography sensor needs to be filtered. We assume a heart rate of 60-120 beats per minute. Therefore, the filtering is done by applying a high-pass and a low-pass filter which combine to a band-pass filter with a frequency range of 1.9 Hz - 8.5 Hz. An example of this filtered signal is shown in blue in figure 5.

2.2 Realization of the Model Fitting

The idea is to obtain the systolic and diastolic blood pressure as parameters of fitting the function $\Delta V$ to the measured data. This is done by classical model fitting. The fitting is realized in MATLAB (The MathWorks, Natick, Massachusetts, MA, USA) using the curve fitting toolbox. The fitting is done using the non-linear least squares method with a trust region algorithm as it is suitable if coefficient constraints are specified.

The measured data is prepared for the fitting process by selecting the peaks of the oscillations (marked in red in Figure 6). These peaks are used as the data points the model function shall be fitted to.

From input pressure $P_{ext}$, only those values are used for which there is also a peak present at the same time.
Figure 6: Oscillation amplitudes and corresponding contact pressure for an example measurement by the smartPWA. The peaks of the oscillations are marked in red.

The model function is, as previously mentioned, the function $\Delta V$. Hence, there are five model parameters which need to be fitted: the two parameters $\alpha$ and $\beta$ which are related to the stiffness of the arterial wall, the systolic and diastolic blood pressure which are the two values we want to obtain with this whole process, and the initial volume of the vessel $V_0$.

Initial Values and Limits. The MATLAB curve fitting tool box together with the Trust-Region algorithm allows us to set initial values, lower and upper limits for these five parameters. To find useful values here, let us take a closer look at $\alpha$ and $\beta$ first. We have already seen that $\alpha > 0$ and $\beta > 0$. Furthermore, it should always be true that

$$\frac{\beta}{\alpha + \beta} \approx \frac{1}{3}. \quad [1]$$

There is also an estimation of $\alpha$ and $\beta$ for the brachial artery where the estimated values are

$$\alpha = 0.11 \quad \text{and} \quad \beta = 0.03. \quad [16]$$

For these values the approximation (15) is far missed since $\frac{0.03}{0.11 + 0.03} \approx 0.21$. However, this can be explained by the fact that there are lots of different shapes of blood pressure curves. For some experiments, in this work the parameters $\alpha$ and $\beta$ are also set to the values given in (16) and therefore the model function in these cases is $\Delta V(P_{sys}, P_{dia}, V_0, P(t))$. [3]

These facts need to be taken into consideration when choosing initial values and limits for $\alpha$ and $\beta$. As initial values, either the values given in equation (16) or $\frac{\text{upper limit} - \text{lower limit}}{2}$ for both are chosen. The lower limit is set 0.001 for both. Regarding the upper limits, there are several options, but especially equation (15) should be kept in mind. The used values are summarized in table 1. The values of each column are used in combination.

<table>
<thead>
<tr>
<th></th>
<th>$\alpha$</th>
<th>$\frac{\text{upper limit} - \text{lower limit}}{2}$</th>
<th>$\beta$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0.5</td>
<td>$\frac{2}{3}$</td>
<td>0.2</td>
</tr>
<tr>
<td></td>
<td>0.5</td>
<td>$\frac{1}{3}$</td>
<td>0.1</td>
</tr>
</tbody>
</table>

Table 1: Different options for the upper limits for $\alpha$ and $\beta$.

The values of each column are used in combination.

Regarding the systolic and diastolic blood pressure, the initial values are chosen to be either $\frac{\text{upper limit} - \text{lower limit}}{2}$ for both or 120 for $P_{sys}$ and 80 for $P_{dia}$, since these values are commonly known as reference values for blood pressure [4]. The upper and lower limit are set to the maximum and the minimum of the contact pressure $P_{ext}$, respectively.

Finally, for the initial volume $V_0$, we choose half of the difference between the maximum and minimum of oscillations of the blood volume as initial value and set the upper and lower limit to the said maximum and minimum.

The Data. From 26 subjects, 443 measurements have been taken at up to five points in time. If possible, for each subject and at each point in time, more than one measurement has been performed, to be able to compare repeated blood pressure estimations. Of the 26 subjects, 15 are male and 11 are female. From 6 of the 26 subjects, the age is unknown. The 1st quartile, median, and 3rd quartile of the birth-years of the other 20 subjects are 1982, 1985, and 1990. The earliest birth-year is 1972 and the latest is 1996.

2.3 Categorising of Measured Data

To optimize the results, one approach is to be more strict regarding the quality of the measurements. Two ways of deciding which measurement is good enough to be taken into account, are investigated. The first approach is to investigate the linearity of the contact pressure $P_{ext}$ over time.
Figure 7: Plot of the contact pressure exerted on the pressure sensor by the finger and the linear polynomial fitted to this curve.

To do so, a linear polynomial is fitted to the contact pressure put to the sensor by the finger. This is done by the ‘fit’ function in MATLAB and the outcome can be seen in Figure 7. The quality of the measurement is then assessed with respect to the goodness of this linear fit. To find a threshold to decide which measurements to keep and which to reject, the sum of squares error (SSE$^\text{lin}$), $R^2$$_\text{lin}$ or root mean squared error (RMSE$^\text{lin}$) for a manually chosen subset of measurements are calculated, and the second quartile or median are used as guideline for the choice of a threshold.

Another approach to categorize the measurements is to accept or reject them with respect to the goodness of the model fitting (SSE$^\text{mod}$, $R^2$$_\text{mod}$, RMSE$^\text{mod}$). In this case, any criterion can be used and the threshold is chosen in the same way as above.

3 Results

The model fitting has been performed for all 443 measurements with different initial values and limits. Figure 8 shows the plot of the fitting curve and the data points for an example measurement using the above specified settings with initial values $\frac{\text{upper limit} - \text{lower limit}}{2}$ and upper limits 0.2 and 0.1 for $\alpha$ and $\beta$, respectively. The initial values for the systolic and diastolic blood pressure are also set to be $\frac{\text{upper limit} - \text{lower limit}}{2}$.

The fitted curve resembles the main properties of the model function $\Delta V$ (see figures 3 for reference). Furthermore, the fitted curve also fits to the data points quite well.

However, there are not only positive results. Figure 9 shows an example where the model fitting did not work as expected.

To evaluate the quality of the fit more objectively, the goodness of the fit parameters provided by the MATLAB curve fitting toolbox are taken into consideration. For the above combinations of initial values and limits, these parameters are summarized in table 2. As can be seen clearly, there are almost no differences between the goodness of the fit parameters for all three combinations.
3.1 Categorising of Measured Data

The categorization of the measurements to decide which are good enough for the estimation of the systolic blood pressure and which not is the next topic we want to look at. As discussed, two methods to do so are investigated in this work: fitting a linear polynomial to the contact pressure curve (‘linear’) and looking at the goodness of the fit of the model fitting (‘model’). The threshold, i.e., the first quartile, median, and third quartile of the RMSE_{lin}, SSE_{lin}, and $R^2_{lin}$ of the linear polynomial fitting calculated for a manually selected subset are given in table 3.

$$
\begin{array}{cccc}
\alpha & \beta & \text{SSE} & R^2 & \text{RMSE} \\
0.5 & 0.5 & 1.0e+07^* [0.1963, 0.4722, 1.1609] & 0.4726, 0.7502, 0.8876 & 281.4590, 432.8865, 674.1670 \\
2.3 & 2.3 & 1.0e+07^* [0.1957, 0.4710, 1.1669] & 0.4819, 0.7522, 0.8905 & 281.4213, 432.3620, 676.7246 \\
0.2 & 0.1 & 1.0e+07^* [0.1963, 0.4793, 1.1743] & 0.4549, 0.7483, 0.8866 & 281.6088, 432.8869, 677.1325 \\
\end{array}
$$

Table 2: The 1st quartile, median and 3rd quartile of the goodness of the fit parameters for 443 measurements. SSE...sum of squares error, $R^2$...coefficient of determination, RMSE...root mean squared error.

To find a threshold for the model fitting based approach, the first quartile, median, and third quartile of the goodness of the fit criteria $\text{SSE}_{mod}$, $R^2_{mod}$, and $\text{RMSE}_{mod}$ are calculated for different methods of robust fitting for the work-set. The results are given in the below table 4.

$$
\begin{array}{cccc}
\text{SSE}_{mod} & R^2_{mod} & \text{RMSE}_{mod} \\
1.0e+07^* [0.1743, 0.4793, 1.2426] & 0.7609, 0.8779, 0.9365 & 276.6219, 449.2120, 697.3921 \\
\end{array}
$$

Table 4: First quartile, median and third quartile of the dispersion of systolic blood pressure and the goodness of the fit criteria $\text{SSE}_{mod}$, $R^2_{mod}$ and $\text{RMSE}_{mod}$ for the subset.

To find a threshold for the model fitting based approach, the first quartile, median, and third quartile of the goodness of the fit criteria $\text{SSE}_{mod}$, $R^2_{mod}$, and $\text{RMSE}_{mod}$ are calculated for different methods of robust fitting for the work-set. The results are given in the below table 4.

Table 3: First quartile, median and third quartile of the dispersion of the goodness of the fit criteria $\text{SSE}_{lin}$, $R^2_{lin}$, and $\text{RMSE}_{lin}$ for the linear fitting to the contact pressure calculated for a subset.

$$
\begin{array}{cccc}
\text{SSE}_{lin} & R^2_{lin} & \text{RMSE}_{lin} \\
1.0e+04^* [3.2299, 4.0901, 5.9620] & 0.9886, 0.9922, 0.9937 & 2.2565, 2.5403, 3.0695 \\
\end{array}
$$

Table 5 shows the results of the categorization using different thresholds. The goodness of the fit parameters $\text{SSE}_{mod}$, $R^2_{mod}$, $\text{RMSE}_{mod}$ and the number of rejected measurements are included.

4 Discussion and Conclusion

The first example, shown in Figure 8, indicates that the model fitting approach is working well and could be the right way to go. The data points already show a distinct shape and the fitting works well. In contrast, the data points in the second example are more scattered. The fitting does not work well in this case. This indicates that the quality of the measurements has a great influence on the quality of the fit.

The goodness of the fitted parameters, the $R^2$ in particular, also show a wide range. This also indicates that there are great differences in the quality of the fit for different measurements. Anyhow, the fitting works well for a part of the measurements, which is indicated by, e.g., the high third quartile of the $R^2$. Furthermore, this already indicates that the goodness of fit parameters, especially $R^2$, can be used to distinguish between good and bad measurements.

The results of the categorization of the measured data further support the above observations. Categorizing the measured data by the goodness of the fit of the linear polynomial fitting only leads to small improvements of the goodness of the model fitting although depending on the threshold already lots of measurements get excluded.
<table>
<thead>
<tr>
<th>Method</th>
<th>Exclusion (linear)</th>
<th>Exclusion (model)</th>
<th>$\text{SSE}_{\text{mod}}$</th>
<th>$R^2_{\text{mod}}$</th>
<th>$\text{RMSE}_{\text{mod}}$</th>
<th>Rejected measurements</th>
</tr>
</thead>
<tbody>
<tr>
<td>-</td>
<td>-</td>
<td>-</td>
<td>$1.0e+06^*$ [19.63 47.93 117.43]</td>
<td>0.4549 0.7483 0.8866</td>
<td>281.6088 432.8869 677.1325</td>
<td>0</td>
</tr>
<tr>
<td>linear</td>
<td>$\text{RMSE}_{\text{lin}} \geq 3.1$</td>
<td>-</td>
<td>$1.0e+06^*$ [1.4586 4.0650 8.6325]</td>
<td>0.5328 0.7825 0.8937</td>
<td>245.4182 412.7944 615.4822</td>
<td>183</td>
</tr>
<tr>
<td>linear</td>
<td>$\text{RMSE}_{\text{lin}} \geq 2.54$</td>
<td>-</td>
<td>$1.0e+06^*$ [1.6480 4.3642 8.7856]</td>
<td>0.6331 0.8152 0.9064</td>
<td>249.9744 420.3820 626.3038</td>
<td>303</td>
</tr>
<tr>
<td>linear</td>
<td>$\text{SSE}_{\text{lin}} \geq 1.0e+4^*$ [5.9620]</td>
<td>-</td>
<td>$1.0e+06^*$ [1.4424 4.0421 8.4975]</td>
<td>0.5327 0.7853 0.8934</td>
<td>244.8690 411.6259 609.2378</td>
<td>188</td>
</tr>
<tr>
<td>linear</td>
<td>$R^2_{\text{lin}} \leq 0.9886$</td>
<td>-</td>
<td>$1.0e+06^*$ [1.4256 4.0168 8.4789]</td>
<td>0.5370 0.7897 0.8964</td>
<td>244.2957 410.2751 610.4114</td>
<td>198</td>
</tr>
<tr>
<td>model</td>
<td>-</td>
<td>$\text{SSE}_{\text{mod}} \geq 1.0e+07^*$ [1.2426]</td>
<td>$1.0e+06^*$ [1.4066 3.2432 6.0491]</td>
<td>0.5116 0.7740 0.8933</td>
<td>238.4647 356.4845 500.8422</td>
<td>107</td>
</tr>
<tr>
<td>model</td>
<td>-</td>
<td>$R^2_{\text{mod}} \leq 0.7609$</td>
<td>$1.0e+06^*$ [1.5425 3.9385 8.1521]</td>
<td>0.8314 0.8923 0.9337</td>
<td>259.8585 403.3576 603.6834</td>
<td>235</td>
</tr>
<tr>
<td>model</td>
<td>-</td>
<td>$R^2_{\text{mod}} \leq 0.8779$</td>
<td>$1.0e+06^*$ [1.3604 3.3716 7.4848]</td>
<td>0.9034 0.9279 0.9546</td>
<td>239.8867 371.3517 563.2087</td>
<td>324</td>
</tr>
<tr>
<td>model</td>
<td>-</td>
<td>$\text{RMSE}_{\text{mod}} \geq 697$</td>
<td>$1.0e+06^*$ [1.4127 3.3117 6.1229]</td>
<td>0.5069 0.7679 0.8935</td>
<td>238.9084 360.1372 509.4549</td>
<td>104</td>
</tr>
</tbody>
</table>
combined \( \text{RMSE}_{\text{lin}} \geq 3.1 \) \( R^2_{\text{mod}} \leq 0.8779 \) \( 1.0e+06^* \) \[\begin{array}{c}
1.4211 \\
3.8300 \\
7.4848
\end{array}\] 0.9034 0.9282 0.9552 238.6218 396.8745 563.2087 364

combined \( \text{RMSE}_{\text{lin}} \geq 3.1 \) \( R^2_{\text{mod}} \leq 0.7609 \) \( 1.0e+06^* \) \[\begin{array}{c}
1.8223 \\
3.9946 \\
7.7939
\end{array}\] 0.8343 0.8921 0.9347 271.4662 412.8987 582.1261 305

combined \( \text{RMSE}_{\text{lin}} \geq 2.54 \) \( R^2_{\text{mod}} \leq 0.8779 \) \( 1.0e+06^* \) \[\begin{array}{c}
2.3574 \\
4.2898 \\
8.3234
\end{array}\] 0.9035 0.9259 0.9514 301.0549 411.8377 602.5087 391

combined \( \text{RMSE}_{\text{lin}} \geq 2.54 \) \( R^2_{\text{mod}} \leq 0.7609 \) \( 1.0e+06^* \) \[\begin{array}{c}
2.1357 \\
4.7765 \\
8.0394
\end{array}\] 0.8435 0.9011 0.9396 286.2074 431.9334 589.1964 363

Table 5: First quartile, median and third quartile of the goodness of the fit parameters \( \text{SSE}_{\text{mod}}, R^2_{\text{mod}} \) and \( \text{RMSE}_{\text{mod}} \) for different methods and threshold for categorizing the measurements.

Categorizing with respect to the various goodness-of-the-fit-parameters of the model fitting leads to better results. Nevertheless, in some cases less than 100 of the 443 measurements would be accepted. The combination of the two categorization methods does not lead to a big difference compared to only using the goodness of the model fitting parameters as criterion. Overall, good results can be achieved when choosing the right criterion. The \( R^2 \) of the model fitting seems to be the best choice. However, it is necessary to find a balance between not excluding too many measurements and still obtaining good results.

To sum up, the model fitting works well for a part of the measurements, but not so good for a great part of the measurements. There are some indicators that this can be traced back to the quality of the measurements. Still, the approach itself shows great promise. Hence, before looking into the resulting absolute values for the systolic and diastolic blood pressure, more investigations on how to improve the fitting by filtering out bad measurements, increasing the measurement quality itself, or preprocessing the measured data is needed. Furthermore, at the same time the possibilities how to categorize the measurements into good and bad ones should be investigated further to find the right compromise between excluding and including measurements.

References


[2] Stadium Shape
https://www.calculatorsoup.com/calculators/geometry-plane/stadium.php; accessed on 15.06.2022.


Service-Oriented Architecture for Automatic Evaluation of Urban Logistical Concepts

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Abstract. With the rapidly increasing availability of logistical concepts new tools are required to provide accurate estimations of concept effectiveness to urban planners. In this paper we describe a service-oriented architecture providing automated concept evaluation through agent-based traffic simulation, combining general traffic models with specialized logistical models into singular simulations. Key focus of the paper is the integration architecture utilized to combine the different tools from a previously developed manual workflow. The architecture is designed to incorporate multiple different tools into a combined workflow, providing non-domain users with a highly specialized tool.

Introduction

Urban logistics is currently changing at a rapid pace. An ever increasing consciousness for environmental factors, new regulatory measures, new services and a pandemic are moving populations from traditional solutions towards utilization of novel logistical concepts [1, 2]. This development confronts urban planners with new challenges as the number of design options for a city logistics system continues to increase, challenging the selection of the most suitable measures to improve the quality of urban mobility and thus environmental, economic, and social effects. New tools are needed to support decision making in municipal planning agencies to increase utilization and effectiveness of logistics solutions [3]. A team consisting of traffic engineers, business IT specialists, computer scientists, and municipal government employees was tasked in the research project USEfUL to create a simulation-based decision support tool for urban logistics [4, 5].

An agent-based simulation approach using the multi-agent transport simulation (MATSim) [6] was chosen to evaluate the impacts of logistical concepts on urban traffic patterns within the city of Hannover. For the evaluation, different key performance indicators (KPIs) were defined. Based on these simulation results, the observed impacts on the KPIs were evaluated in order to derive recommendations for urban planners. Consequently a hand-crafted traffic model [7], adapted to the resulting needs and behavioral patterns stemming from logistical concepts was developed, which in turn fed a semi-automated evaluation tool chain. Due to the comparatively long run times of the agent-based simulations and the high computational effort, all scenarios were simulated manually and only the corresponding impacts were stored in the tool. These results are on display on the project’s web page [8], where users can gather information about the concepts, the evaluation areas and predicted effects on different KPIs.

However, the workflow with the semi-automatic traffic model generation and the evaluation framework is currently only applicable to the city of Hannover. In particular, the setup of the traffic model is limited by available data sources and therefore not directly transferable to other cities and municipalities. To provide similar assistance to other communities, a completely automated architecture needs to be designed and implemented. The resulting tool has to utilize public and private data sources to create an accurate baseline model of the area of interest, before standardized changes to the model implement novel logistical concepts.

To tackle these issues – and increase the usefulness of the web-based tool – the follow-up project USEfUL-XT aims to increase the transferability of the USEfUL workflow. In this paper we present the architecture to
combine previously manual workflows using a service-oriented approach. A high-level view of the different services is given, as the implementation of the services is still in progress. The result is the concept of an integrated, service-oriented and cloud-based framework capable of automatically collecting and processing the required data, creating a simulation model and performing the resulting simulation runs. As the specified simulation scenario only need to be simulated once, the computation needs are also dynamic and efficient resource usage demands a dynamically scalable solution with adequate storage space to store results.

Section 1 will give an overview of related work. In the following section 2 the current workflow of the project is presented, which is then expanded to an automatic workflow in section 3. Remaining challenges are discussed in section 4 and a conclusion is given in section 5.

1 Related Work

A similar tool to the USEfUL web application was provided by [9]. The Urban Transport Roadmap provides statistical evaluations for many regulatory measures and estimates the impacts of technological development on urban traffic. However, the tool focuses on very rough estimates for many different cities and allows in-depth analysis of combinations of measures.

Another theoretical framework for automated evaluation of urban traffic was proposed by [10]. In their proposed system, a user can utilize a partially automated co-creation process to construct simulation models of a city. After model creation, partial changes to the model are proposed to the user, which, when chosen, are in turn simulated within multiple simulation tools. At the same time a rule-based system watches the simulation for inconsistencies and provides input to the user to correct these. The publication focuses on the process of creating the tool, rather than the tool itself. It is unclear if the tool can provide automated model creation for multiple areas and logistical concepts. Furthermore, the proposed system of [10] does not include an expert system to fast-track concept evaluation. The proposed system remains theoretical.

In another work, [11] present a generalized pipeline for the generation of agent-based transport models in France. This pipeline presents a standardized approach for processing data sources to synthesize a synthetic population and assign mobility behaviour to the agents. The paper also includes a detailed validation of the method. However, the described method is focused entirely on the generation of a model for private traffic omitting commercial traffic and is adapted to the specifically in France available data.

An example for a microscopic modeling approach for commercial traffic is presented in [12]. Therein, the authors then employ land use and structural data to generate businesses and the respective employees and vehicle fleets. By applying the German commercial travel survey, each vehicle is equipped with a daily schedule.

Meanwhile, many other simulation models and studies have been created to identify issues and solutions for urban logistics [13]. However, model creation was not completely automated and often only small areas of urban logistics were modelled. Our proposed system approaches urban logistics from a more abstract point of view, allowing a rough estimation of impacts for many concepts and areas, through automated model creation, simulation and evaluation.

2 Current Solution

The current workflow combines manual data collection, preparation, and model calibration as well as simulation and evaluation. A lightweight, web-based decision support tool was created, not slowed down by long simulation run times, as users want to use the tool at a quick glance. The goal of the process was the easy-to-understand representation of logistic concept impacts on specific KPIs, which are presented to users in a web-based decision support tool. Traffic and logistic simulations are performed using MATSim to benchmark the impact of new logistical concepts. The process can be split into seven parts, as shown in Figure 1.

The first step of the workflow is data collection, as many different sources of information have to be utilized to manually collect the necessary traffic supply and demand information. Several open data sources like OpenStreetMap (OSM) and public transport providers are considered to create a road network with an integrated public transport system for the traffic supply needed in the MATSim model.

To generate a realistic traffic demand, multiple studies and non-public data from companies and municipalities are consolidated into a synthetic population. Furthermore, different input surveys are used to derive the travel or logistic demand from the synthetic population. From the collected data a basecase model is created,
representing the status quo of traffic within the research area. A manual calibration of the model using traffic counts, as well as a validation against different mobility parameters is performed to achieve a high model accuracy. Building upon the previously created basecase, demand changes due to the implementation of novel logistic concepts are replicated in the model and need to be simulated subsequently. Each concept is parameterized in different dimensions (e.g., usage of electric vehicles, total acceptance percentage, etc.), studying concepts more in-depth. A large amount of simulation runs were needed for every area of interest and each parameter combination, since multiple logistical concepts were compared to provide a useful decision support.

Simulation results are in turn evaluated with regards to eight KPIs: acceptance, emissions, area usage, costs, implementation potential, traffic impacts, economic feasibility, ecological break-even-point. For the evaluation, the basecase is considered as the starting point and trends on the KPIs through utilization of logistic concepts are derived. After all possible scenarios are simulated, evaluation data is combined in a database queryable through a web-application. Users can now select a research area, a concept (possibly modify it through parameters) and are presented with the evaluated simulation results.

While this workflow produces good results for predefined areas, adaption to new areas is labor intensive and slow, diminishing the usefulness for other parties. Furthermore, since all possible combinations of areas, concepts and parameters need to be simulated a priori, timely simulation requires a large amount of computational resources. A optimized process would allow the users to select the area of interest, a specific concept and then receive the simulation results on-demand (either through a new simulation study or previously performed simulations). Consequently, we designed this concept of an automated simulation framework. This goal-process is shown in Figure 2.

The improved workflow also utilizes a rule-based expert-system to give impact estimations almost instantly, as opposed to the simulation, where turnaround times of multiple days, if not weeks are to be expected. To achieve this workflow, major challenges must be solved, most importantly the automatic creation of models based on automatically collected data. While the previous process already uses automated tools at each step in the workflow, data transfer between steps was handled manually, increasing the total time needed. Implementing the new workflow requires a clear definition of software interfaces, processes and deployment into an easily scalable environment.

### 3 Service-Oriented Cloud-Based Workflow

Service orientation and cloud usage are computing patterns that address multiple issues also encountered in the simulation workflow of USEfUL. Service orientation allows the easy integration of quickly evolving tools into combined workflows while providing abstract interfaces, decoupling different steps of software processes [14]. This in turn allows the easy combination
of many different software tools into a combined application. Cloud-based computing provides quickly scalable infrastructure by combining different applications on the same hardware, therefore reducing the cost for hosting of applications.

With a cloud-based simulation service, users can request simulations ad-hoc, paying only for the computational power that is actually needed to run the simulation, instead of investing in expensive simulation hardware. Since primary user groups are located in communal administration and other non-software firms, this approach increases the attractiveness of the solution to the end user. While the architecture is aimed at a cloud deployment, it is not necessary to rely upon an external cloud provider. Through the usage of a private cloud, the entire software can run on-premise, if e.g. data security or privacy requirements dictate it. In general service-orientation is implemented by encapsulating specific tasks of the workflow into independent services that can be recombined for other workflows. Figure 3 shows the proposed service-oriented architecture for the automated solution.

The workflow of the proposed system is controlled by the already existing WEBAPP, providing user interaction and service orchestration. Since only a single workflow – the evaluation of logistic scenarios – is to be expected, a flexible workflow engine is not necessarily needed and implementation complexity can be reduced. In case the workflow changes for the different simulations, an increased workload to implement the changes is to be expected, but does not outweigh the reduction of implementation complexity. As the WEBAPP has already been developed in USEFUL, it can easily be adapted to the necessary changes.

The Service Bus provides stable service endpoints for usage through the WEBAPP (or other applications) and connects them to the service implementations. This effectively decouples declaration and definition of different tasks, allowing easy redevelopment of different parts of the overall workflow without leading to widespread changes. Many different existing solutions can be adapted for this particular purpose. Communication between the Service Bus and other components is technology-dependent and can be implemented in multiple different ways. As the chosen simulation tool MATSim utilizes xml-files for configuration and modeling, data transfers will also utilize the xml-standard to achieve design homogeneity.

The Services implement the steps of the workflow and utilize technologies independent of the rest of the application. This allows technology selection according to the domain-specific needs, as well as flexibility when services are updated to new methods. A description of the services is given in the following sections.

Some cross-cutting concerns are life cycle management and security, as these are needed at all levels of the application. While life-cycle management needs to be decentralized due to the technical nature of the domain, security can be handled at the Service Bus and deployment environment. E.g. access to services from external sources can be blocked by a firewall and internal compartmentalization is achieved by extending the existing role-based-access-control to service bus level.
3.1 DataServices

The data services, are responsible of adapting different sources of data into the system. To create simulation models for urban logistics, different information sources are summarized to create the traffic supply as well as the traffic demand.

As these data sets are built from many different sources, the services are responsible for aggregating this data and providing the necessary information in a standardized format. The data services consolidate the processed data in a data warehouse and allocate it to the associated services for further processing. While some data might be pulled from standardized web services, other need to be manually interpreted, prepared and provided in a suitable format. Consequently most of the technical and logical complexity of handling different data is localized in these services.

Figure 4 shows a high-level view of one of the main data services, the TRAFFIC DEMAND SERVICE. This Service is responsible for providing information about traffic demand and population behavior, which in turn is used to generate traffic demand within the MODELCREATORSERVICE. Information for this service is gathered from public APIs like OSM/Overpass, public statistical data, manually extracted from documents, and non-public studies. Data pulled from public APIs can easily be converted into the necessary XML structure, utilized by consuming services. Other Data, like the aforementioned public statistical data, needs manual conversion and own storage infrastructure, as document structures vary vastly between different sources or iterations of the same studies. For example, the study Mobility in Germany 2017 added a new excel-based tool for data exploration, which was not available for the studies predecessor, created in 2008 [15]. While automatic extraction of the study is possible, the created tool would only work for this iteration of the study. The resulting effort/effect-balance is not favourable for automation. To make manually extracted data available for the automatic workflow, it is entered into a database, which provides a machine readable data format. Since the TRAFFIC DEMAND SERVICE collects data from different sources and provides a unified interface for consumers, it can be viewed as a data warehouse.

As part of the modelling process, the traffic demand service also provides data for the automated model calibration service. Thus, mode share information of the simulation area as well as other mobility parameters must be specified and provided by the user. An essential element for the quality of the model is the availability of real traffic data from the simulation area particularly consisting of vehicle count data of different sections of the network. While this detector data can be accessed online for specific network connections, it is often not directly available. Therefore, this missing component must be provided by the user.

3.2 ModelCreatorService

After all necessary input data is collected and processed into the data warehouse, the MODELCREATORSERVICE is executed. This service contains three consecutive modeling steps that gradually generate an agent-based transport model for individual traffic along with a service-based model for commercial traffic (Figure 5).

In the first step, a synthetic population consisting of households with corresponding agents is generated based on the grid cells that are deposited in the data warehouse. Grid cells are selected according to the area of interest included in the execution request. Each of the agents in the population represents one person living in the specific cell and is assigned with certain sociodemographic attributes, such as gender or age, as well as attributes that impact the mobility, such as the possession of a driving licence or public transport ticket. In addition to these agents, companies are synthesized as
location for work activities and as instances that generate commercial or logistic traffic. The companies are differentiated by industry sectors and also obtain attributes such as the number of vehicles and the number of agents working at that location.

In the second step of the model creation, each agent is assigned a mobility behavior corresponding to the predefined attributes. The behavior is defined by a sequence of activities the agent performs during a day and the necessary trips that the agent must perform to reach those activities. These patterns are provided by the data services and also contain additional information such as the transportation modes used and the travel distance for the trips. The location choice for the activities is then carried out on the basis of these information along with land-use data.

The last element of the ModelCreatorService is the generation of commercial trips. In this process, the defined vehicles for each company are assigned with a sequence of services they have to operate on the simulated day. The services act in a similar way to the activities of the agents and are linked to the population and other companies. For example, if a craftsman performs a service for an agent, the agent has to be at home. A special focus is laid on logistical aspects of the model. Specifically, the parcel demand is used to generate service sequences for parcel delivery vehicles of multiple delivery services. The resulting vehicle routing problem is solved using a ruin-and-recreate-algorithm and resulting trips are assigned to the vehicles [16]. The output of the ModelCreatorService consists of several independent XML-files, representing the model in its entirety. This model then has to be calibrated and validated using additional data sources from the warehouse that were not used for the model creation.

3.3 ConceptScenarioService

After the basecase has been created, further changes to the model are necessary to inspect different logistical concepts. In the ConceptScenarioService sub-components make changes to the model according to the concepts. For example, some of the traffic resulting from grocery runs is replaced by a centralized grocery distribution system. To add a new concept, new components can easily be added to the service, without changing other parts of the system. Since the logistical concepts are not in service at the research area, validation of the changed model proves difficult.

3.4 SimulationRunnerService

The simulation service is responsible for scheduling and executing the simulations. A detailed analysis through simulations can be provided by utilizing a worker cloud where simulation tasks can be asynchronously executed.

As shown in Figure 6, Simulation Requests are accepted by the SimulationRunnerService, where they can be stored in a queue until resources are available for execution. The Simulation Request contains XML-Files which were generated in the ModelCreatorService along with configuration parameters for the simulation, which have to be provided by the user. Data is then fed into the Simulation Workers, which execute MATSim simulations on dynamically scaling infrastructure e.g. through the use of Docker containers. After the simulation runs are finished, data is pushed into a Result Store for sub-
sequent requests to prevent repeated simulation of the same model/parameter-combination.

3.5 AnalysisServices

The final step before presenting concept impacts to the user is the analysis of simulation results. The analysis is carried out by applying models for different KPIs to parts of the simulation output (e.g. additional traffic caused by the given concept). As models are created by domain experts, an integration for widely used tools like Microsoft Excel is required. The different models compare information from the simulation of the modified models to the basecase and derive trends for KPIs. This allows the easy-to-understand representation in the final visualization for the user.

One example of ANALYSIS SERVICES is the AREA MODEL, which gives an estimation on the expected change of area needed for the implementation of a given logistical concepts. The AREA MODEL uses information like the amount and type of delivery vehicle used to implement a given concepts and then calculates additional area needs for e.g. construction of micro hubs. This is in turn compared to the reduction of e.g. parking spaces to calculate the net difference. Since the effects of concepts are also interconnected, the AREA MODEL feeds its result directly into the COST MODEL, which predicts cost changes.

3.6 RuleService

As an alternative to the time-consuming simulation path (shown in Figure 2), a rule-based expert system can be utilized to estimate the impacts KPIs. The RULE SERVICE uses the JSONRULES ENGINE and multiple sets of Rules to estimate Results, as each concept reacts differently to structural and behavioural peculiarities of different concepts. While not providing the accuracy achieved by the simulations, the RULE SERVICE allows the preliminary study of concepts effects on non-simulated areas to identify promising combinations for further study.

The rules are built upon previous simulation results for each logistical concept and model the connection between concept effects and structural attributes of areas. As no high accuracy is expected until a large amount of simulations have been carried out, the first iteration of the rules engine only predicts three categories of change, instead of a numerical scale. As more and more simulations are carried out, new results are incorporated into the rules, thus improving model prediction. However, the rule engine cannot predict changes in radically new areas or for unknown logistical concepts.

3.7 WebApp

After the impacts of the concept have been evaluated, a web-based decision support tool presents the results to the user. This WEBAPP is built upon Laravel [17], using a modern design to attract users. It uses a rule-based expert system [18]. Functionalities for selecting areas and concepts as well as dynamical presentation of results are already implemented, therefore only additions to existing components are necessary.

4 Remaining Challenges

While the technical integration of all existing tools can easily be achieved through the use of available software solutions, some procedural issues remain. First and foremost the validation of the simulation models remains problematic, as multiple data sources for similar information are required. Data has proven to be relatively scarce, as privacy of residents or workers needs to be protected and some data are simply not collected. Furthermore, validation is currently a time-consuming, manual process, that has not been fully automated yet.

Another challenge is the calibration of the simulation models, as this is again a manual process, requiring a human-in-the-loop approach to properly interpret information and adapt parameters accordingly. An automated process could involve a meta-optimization approach such as evolutionary algorithms, but would in turn increase resource demand and runtime. Since open data projects are utilized, data quality is inconsistent, as data tends to be biased. One of the main reasons for this are the inhomogeneous contribution patterns of its contributors [19]. This issue is not fixable through the proposed system, as automatic data quality improvement is out of scope for its requirements.

5 Conclusion

In this paper an architecture for the automation of simulation and decision support has been presented. Built upon a previously developed workflow, the system aims to allow users to evaluate novel logistical concepts in specific research areas. The system shall provide rough
estimates on effects on key performance indicators of traffic like emissions and population behaviour. This allows users to evaluate logistical concepts in theory and aids in preliminary selection, reducing the need for cost-intensive prototype-projects in the real world.

While the technical aspects of the proposed system do not pose show-stopping challenges, domain specific workflows, like simulation validation need to be developed further to implement the system. In future work a prototypical implementation as well as the definition and evaluation of automated processes for simulation validation and calibration are planned.

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Raising Awareness of Climate Change Impact and Mitigation - The MyWorld Simulation Game

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Abstract. As climate change increasingly impacts natural systems and living conditions all over the world, it becomes even more important that the public understand the probable course of climate change, its impact on world-wide and local ecology and living conditions, and the extent to which mitigation measures can alleviate that impact. Researchers can contribute to this by developing simulation models of the course and impacts of climate change, and make them available to the general public, packaged with a user-friendly interface and progressively detailed information.

This paper introduces MyWorld, a simulation-based game that enables the player to experience the course and impact of climate change and to experiment with potential mitigation measures. The freely available game aims at facilitating a better understanding of the probable progression of climate change and its impact on living conditions in the world in general and specific countries in detail.

Introduction
Climate change increasingly impacts all areas of human existence, including such diverse fields as urban infrastructure systems and housing [1], century-old cultural heritage assets [2][3], medicine and human health [4], and agriculture [5]. With the range and velocity of climate change increasing, it becomes even more important that the general public understands its impacts and potential remedies.

Here, simulation can help. Wrapped in a user-friendly game-like environment, a (simplified) simulation model of global climate change and its impacts on living conditions, broken down to a user’s local environment and extended by estimates of the effectiveness of mitigation measures, can increase the awareness of changes ahead and improve decision quality regarding potential changes in economy, societal organization, and personal lifestyle. To reach those goals, a model does not have to be as precise as an analytical simulation for the validation of real-world decisions and strategies. In return, players have no patience to wait for hours for the conclusion of a simulation run – an underlying simulation model has to be fast, even at the cost of precision.

This paper presents the MyWorld simulation-based game [6] aimed at facilitating a better understanding of the probable progression of climate change and its impact on living conditions in the world in general and specific countries in detail, as well as the effect of potential mitigation measures on those living conditions. The game presents the players with a visualisation of the course of climate change in any specific country, gives them the opportunity to experiment with mitigation measures, and simulates the impact of those measures on the CO2 balance and the further course of climate change itself. Playing the game thereby raises awareness of climate change impacts, potential mitigation measures and their effects and interdependencies.

Literature describes a number of serious games in the area of climate change and sustainability (for an overview, see Stanitsas et al. [7] or Gerber et al. [8]). Only a minority of approx. 10 % of those primarily aim at raising awareness [8], many of the others are more oriented towards strategy design [9], collecting data [10][11], or direct behavioral change [12]. Some are table-top games without computer support [9][13][14][15], while others are at least partially computer-supported [10][12][16][17][18][19][20]. Many of those examine specific (sub-) systems, like water management [17][19], household energy efficiency [12], and cultural heritage protection [20].
A few of these serious games warrant a closer examination: Van Pelt, et al. [17] describe a simulation-based game aimed at conveying the inherent uncertainties of climate change in the area of water management to expert stakeholders; their experimental setup includes ex ante and ex post surveys and control groups. They conclude that simulation-based games can be valuable in supporting the communication of climate change uncertainties.

The game proposed by Gangollels, et al. [12] aims at facilitating behavioral change in the area of household energy conservation, specifically in affordable housing communities. Based on their observations, they identify a significant potential for greenhouse gas savings that could be realised by individual behavior change. They propose to facilitate that behavior change by exposing members of the general public to a simulation-based game.

Bontchev, et al. [20] examine, based on the example of their own computer-based game, what factors robustly contribute to the awareness of climate resilience, especially in the area of built heritage sustainability. They find that learnability factors, e.g., ease of understanding the mechanics and gameplay, consistency in its mechanics, and direct feedback to a player’s actions, contribute strongly.

De Kraker, et al. [19] propose the usage of a multiplayer game to foster social learning, based on an example of river water management. They find that the game facilitates a convergence of perspective in stakeholders from various backgrounds especially if the simulation model’s mechanics and intermediate computational results – and therefore, by extension, the underlying interdependencies of the real-world system – are clearly communicated and not hid as a black box.

Based on that – admittedly, non-comprehensive – literature review, there seems to be a research gap regarding an awareness-raising game that covers the whole climate system and allows players to experiment with the impact of mitigation measures as they are discussed in politics and media. The review also showed that a simulation-based game should be clear about the underlying model’s mechanics, should contain information on any sub-system and parameter that can be accessed by a player if they desire so, and should expose the player to direct feedback regarding every decision they make. The MyWorld game aims to fill that gap.

The paper continues with a short introduction to the MyWorld game as seen by the user (Section 1), followed by a description of the underlying simulation model (Section 2). It concludes with some thoughts on the context the game can be used in and an outlook on further research (Section 3).

1 The MyWorld Game

MyWorld offers a number of main views on the developments and dependencies of the underlying climate change model; it tries to convey to the user a sense of what those developments might actually mean to them and their local context. The main views are a world map and country information, a local and global ‘climate press’, the CO2 balance of the world or specific countries, and the mitigation measures selected by the user. While aimed at being easily understood by novice users, MyWorld also puts a focus on offering detailed information on all aspects of climate change. The game can be downloaded freely from its web site [6].

1.1 World Map and Country Information

The first major view depicts a world map on which the player can select a country. Detailed data on that country will be shown in the country data area on the right of the map (see Figure 1). A date slider enables the player to select a year – between 1960 and 2100 – for that they would like to examine data. Beneath the data slider an info area depicts in what years, given the current set of parameters and mitigation measures, the model estimates that global temperature will break the 2 °C, 3 °C, 4 °C, and 6 °C thresholds.

The country data area provides a wide range of both world-wide and country-specific data for a selected year, including population data, CO2 emitters, energy production, societal and humanitarian factors, climate- and weather-related data, ecologic and economic data. Each of these items provides an information button revealing progressively more detailed information of the parameter and the underlying interdependencies, and a thumbnail-sized graph that depicts its development over time. The graphs can be expanded for each variable for closer examination.

1.2 Climate Press

To increase the degree of immersion and to give a better understanding of course and impact of climate change as well as the effect of mitigation measure, MyWorld includes a ‘simulation’ of both local and regional press articles. These ‘articles’, for any combination of simulation time, selected country, and selected mitigation measures, give a summary of current developments world-wide or. Technically, these ‘articles’ are composed out of simple boilerplate text modules based on the current development of the simulated climate and living conditions at any point in simulation time.
This is a significantly shortened example of the ‘news article’ localized for Germany for the year 2030 with some activated mitigation measures:

“Germany wants to be climate neutral by 2045 – By 2039, a 75 % reduction is targeted. Food and tobacco industry, chemical industries, agriculture and livestock, and cement manufacturing will be equipped with filters for carbon capture and storage of 75% of all emissions. Power generation will be CO2eq-free by 75%, with at least 901 TWh CO2eq-free energy generated. For this purpose, at least 128,742 wind turbines or 64 nuclear power plants are required. The purchasing power of the population will be reduced by 5.8%.”

1.3 CO2 Balance

For each point in simulation time, as indicated with the date slider, the game depicts the CO2 balance resulting from the currently simulated climate system, including the mitigation measures selected by the user (see Figure 3). MyWorld distinguishes between all major carbon sources and sinks, including oceanic CO2 uptake, human-caused emissions and carbon capture and storage (CCS), photosynthesis, freshwater, weathering, volcanic eruptions, as well as other natural sources. The game also summarizes whether the CO2 level in the atmosphere under the given conditions is growing or shrinking, i.e., whether the mitigation measures selected by the user are sufficient to stop climate change in the long run.
1.4 Mitigation Measures

*MyWorld* offers its users the opportunity to define and implement mitigation measures at any point in simulation time and enables them to investigate how these measures might influence the further course of the earth’s climate. As a starting point for tinkering, the model presents the measures that various nations agreed upon as part of their Nationally Determined Contributions (NDCs). To be easily understood, measures form English sentences, e.g., “For Australia, phase out CO2e down to 0.1 times the value of 1990 by the year 2040, starting in 2025.”

2 The MyWorld Simulation Model

To be of use in a near real-time game environment, the simulation model has to be executed fast, therefore putting limits on its complexity and preciseness. Simplifications are justifiable, as the model is not aiming at evaluating real-world decisions and strategies for their effectiveness and efficiency, but to give the user an impression on what impact climate change might have on their living conditions, and what effect mitigation measures might have on the further course of climate change.
The following sections first describe the sources of the utilized data, then the basic simulation model based on a set of difference equations [21][22], and finally details the way potential mitigation measures selected by the user are integrated with the calculations.

2.1 Data and Constant Sources

For the MyWorld climate model, only data and constants from reputable data sources such as the UN, IPCC, WorldBank, or Our World in Data, are used. A separate view enumerates the sources of all data, including where to find them on the internet (see Figure 5). As with most complex models, experts disagree on some of the constants, such as certain demographic indicators. These constants can be adjusted by the player to suit their needs. The following paragraphs exemplarily describe some of the most important data series, their sources, and the way of integration with the MyWorld model.

The CO2e-level in the atmosphere is continuously measured by NOAA [23]. CO2e stands for “CO2 equivalents” and contains the greenhouse gases listed by the International Panel on Climate Change (IPCC) [24], primarily water vapour (H2O), carbon dioxide (CO2), methane (CH4), nitrous oxide (N2O) and ozone (O3) as well as entirely human-made greenhouse gases, such as halocarbons and other chlorine- and bromine-containing substances, like sulphur hexafluoride (SF6), hydrofluorocarbons (HFCs), and perfluorocarbons (PFCs).

Data on the energy mix per world region is collected by Our World in Data [25], albeit incompletely. Some of the data series show erratic trends, which is due to the fact that the data sets are incomplete. MyWorld estimates the future energy demand of a nation based on the current energy consumption per inhabitant. This method is inherently inaccurate as it does not take into account whether energy consumption per capita will increase or decrease. The current trends are diverging for different nations, so that an exact prediction seems hardly possible.

Current and future population numbers and birth rates per region are estimated by the UN Department of Economic and Social Affairs [26]. While the estimates by principle cannot be exact, the UN data collection is generally seen as the most reliable available.

Types of landscapes, especially percentage of tree cover, are monitored by the OECD [27]. MyWorld models changes in land cover as a simple function of deviation of temperature to the historical mean. This approach is very simplistic because it does not consider geography, global flow systems, or peculiarities of different nations. Predictions on local levels can therefore be inaccurate.

2.2 Basic Model

Climate change will have a significant – potentially devastating – impact on future living conditions. That said, the authors of the game want to avoid MyWorld presenting an overly pessimistic view of the future.

Figure 5: MyWorld informs the user on data used in the underlying climate model, including data sources.
The game therefore estimates the impact of climate change on natural systems and living conditions on the conservative side, turning optimistic when modelling the capacity of nature and society to mitigate the impact and to adapt to it.

The underlying simulation model employs an iterative difference equation system [28]. The system calculates the values of approx. 60 variables with an increment of one year from 1960 up to 2100. The variables come from the fields of population, climate, energy, CO2 emitters, society, ecology, and economy, and are explained to the player in detail. An individual difference equation is defined for each variable, which allows the most probable extrapolation of the value curve. Some equations use heuristics and constants taken from publications of the highest available quality, as described above. The historical variables collected from the data sources are in some cases used to adjust the parameters of the heuristics as well as possible, given the goals stated above regarding erring on the positive side.

All equations and heuristics can be controlled and partially changed in the ‘Model’ view (again, see Figure 5) by the user. That view also indicates for each equation whether it is based on historical data, is generated by a reliable approximation method, or consists of a plausible but basically uncertain estimation.

The predictive significance of iterated differential equations decreases with increasing $t$ [29]. Therefore, as already described, the MyWorld simulation model used does not produce exact predictions of the future. Rather, it is aimed at creating plausible scenarios and developments that give the player an understanding of dependencies in the world climate system and the general impact of climate change on their future living conditions.

2.3 Mitigation Measures

As its main gaming element, MyWorld allows the player to define their mitigation and adaptation strategy by defining countermeasures in the ‘Measures’ view.

For this purpose, the player can switch on and off the default measures committed by individual countries or change their parameters, such as percentages or time targets. In additional information boxes, the player can find out how binding the commitments of the nations are, e.g., whether they are anchored in law or merely non-binding declarations of intent.

In addition to those default measures, the player can define their own measures by influencing the course of certain variables. For example, they can stipulate that a particular nation will reduce certain emissions by a fixed factor by a certain date. Depending on the adjustment of the mitigation measures, the simulation engine calculates a new course for all variables and adjusts all views accordingly. As a result, the player can develop an understanding of which measures work how well, and how ambitious the world’s mitigation measures have to be to avoid negative impacts on living conditions and natural systems. That didactic effect includes also an understanding of the timeframe in which certain measures have to be implemented in order for the entire ecological and societal system to thrive in the upcoming decades.

The following model variables can be modified by user-defined measures:

- **Birth rate:** MyWorld assumes that human-induced climate change is directly related to the number of people living on earth. If the user does not modify this variable, the model assumes the UN-predicted birth rates [26].

- **Death rate:** The death rate, along with the birth rate, is also an important factor for the size of the world population. Unmodified by the user, MyWorld assumes that the death rate stays at the level of 2020.

- **State Fragility:** The Fragile State Index [31] is a measure of a nation’s internal security. By default, this factor assumes changes in fragility dependent on the development of the nation's gross domestic product.

- **Gross Domestic Product:** The gross domestic product is a measure of a nation's productivity and thus of the standard of living of its citizens. By default, MyWorld adapts the gross domestic product dependent on the mean annual temperature of a nation.

- **Forestation:** Forested areas are an important factor influencing the removal of CO2 from the atmosphere. If a user changes this factor by planning (de-)forestation campaigns, MyWorld automatically adjusts the land use of the nation. This means that the other landscape types (i.e., grassland, wetlands, shrubland, desert, sparse vegetation, and cropland) are adjusted so that the total area remains unchanged. Without specific measures present, this factor is derived from the temperature-dependent utilization model.

- **CO2e per capita:** In MyWorld, this factor is the central control variable for influencing the CO2e balance of a nation. More detailed factors, such as the energy mix or the mix of emitters, cannot be influenced directly by the user, as only insufficient models exist describing how changes in more detailed factors affect a nation.
Negative Emissions: This factor indicates how much CO2e is removed from the atmosphere by human activities. For such a carbon capture and storage strategy, various methods are potentially feasible. However, currently there are no significant capacities of negative emissions [32][33], so they are assumed to be zero by default. Since it is also currently unclear whether and when such processes could be used on a global scale, MyWorld does not make any pre-assumptions but leaves them to the player to schedule.

MyWorld does not prescribe in detail how targets set by mitigation measures are achieved, or why the targets might be feasible or unfeasible to reach. For many of the mentioned variables (e.g., forestation, negative emissions, and state fragility) no detailed understanding of their feasibility of implementation or impact on the course of climate change exist yet.

However, to get a better understanding of the course and impact of climate change, it is sufficient to show to a user what might happen if certain targets are reached or not reached – independent of the means of their implementation.

3 Conclusion

This paper presented MyWorld, a simulation-based computer game that aims at facilitating a better understanding of the progression of climate change and its impact on a user’s regional living conditions, as well as the effect of potential mitigation measures on climate change and those living conditions. The paper described the game both from the user’s point of view as well as the underlying simulation model.

The simulation model has been developed, implemented, and verified. A user-friendly graphical interface has been developed, giving users access to the model parameters and results. The game now is ready for extensive evaluation with stakeholders. In further research steps, the game will therefore be co-evaluated with university students from the earth sciences, municipal climate protection managers, other stakeholders, and the general public.

References


Abstract. Bayesian networks can be used for analysis and representation of dependencies in large data sets. Due to their property of operating with graphs, they are suitable for analyzing delays in rail networks. After getting an overview of the theory of Bayesian networks, this article deals with recent literature about Bayesian networks and train delays. Furthermore, the presented methods will be applied to data from the Austrian railway network.

Introduction

Since public transport plays a major role in society, it is on societies behalf to keep delays to a minimum. In order to reach the goal of minimizing delays, it is important to understand not only the reasons of delays but also the propagation of delays within the public transport network. The availability of large data sets, powerful computers and suitable mathematical tools allows analyzing such delays in a sensible way.

1 Bayesian Network

A Bayesian network is a model represented by a directed acyclic graph (DAG) $G = (V, E)$, whereat $V$ is the vertex set and $E$ is the edge set. Every vertex represents a random variable. There exists an edge $e \in E$ between two vertices $v,w \in V$, if and only if there is a probabilistic dependency between the two random variables represented by $v$ and $w$ ([1]).

Depending on the model, the random variable (i.e. vertex) can take on different types of values ([2], [3]), e.g. unordered values (e.g. blue, green, yellow), ordered values (e.g. low, medium, high) or continuous values ($\mathbb{R}$). Every random variable also has its probability distribution, which needs to be added to the graphical representation. If the random variables take on continuous values (e.g. when dealing with continuous data), we often consider multivariate normal variables, i.e. the random variables are normally distributed and linked by linear constraints. An example of this case is illustrated in Figure 1.

Figure 1: Considering that a train at station Himberg is 10 minutes late, then it is expected to be 6 minutes late ($3 + 0.3 \cdot 10 = 6$) at station Gramatneusiedl, with a standard deviation of 2 minutes.

When using Bayesian networks for modelling real-world problems, we often just have data describing the values of our random variables, but no information about the graph structure, i.e. the edges/dependencies between the vertices/random variables. In this case, we want to derive the graph structure from the values of the random variables. The Max-Min Hill-Climbing method (MMHC) enables this derivation and will be used in Section 2 and Section 3. A detailed introduction to this algorithm can be found in [4].
2 Analyzing the Long Island Rail Road System

In [1], Ulak et al. investigate network-wide pattern of rail transit delays in the Long Island Rail Road System, i.e. a railway system in New York that has a length of 620 miles and consists of a total of 11 rail lines. In Section 3, the methods used in [1] will be modified and used for data of the Austrian railway system. Therefore, Section 2 recapitulates the methods and ideas of [1] without going too deep into detail or presenting the results.

In order to build a Bayesian Network as described in Section 1, every station is considered as a random variable. It takes on the average delay at this station within a certain time interval. There should exist an edge between two vertices (i.e. between two stations) if and only if there is a dependence between those two vertices (i.e. if a delay at one station affects the occurrence of a delay at the other station).

It is possible to classify each train as either eastbound or westbound. When analyzing the delay propagation, i.e. when creating the Bayesian Network, either all eastbound or all westbound trains are considered, but not both east- and westbound trains.

By analyzing the data of an app used by the passengers, the authors receive the delay of every train at every station (with a few exceptions). The observation period is divided into 1-hour intervals. For every station the average delay per train within the examined hour is calculated. A data sample consists of the average delay time at every station within one hour. By using the Max-Min Hill-Climbing method (MMHC) and the maximum likelihood method, the graph structure and the corresponding parameters are compiled from the independent data samples. The R-package bnlearn was used for this task. Detailed documentation about this package can be found in [3].

In order to receive an expressive score for every station's role in delay propagation, two metrics are introduced. The Inducer score of station $s$ indicates how much a delay at station $s$ influences the delays at all the other stations. The Susceptible score of station $s$ indicates how much the delay at station $s$ is influenced by the delays at all the other stations. Similar scores will be introduced in Section 3 in a more precise way.

3 Analyzing the Austrian Railway System

In this section we want to modify the methods of [1] and apply them to data of the Austrian railway network. For this purpose, delay data provided by the Austrian national railway company Österreichische Bundesbahnen (ÖBB) is used. It not only contains information about the delay at every train station, but at every operation control point, a so-called "Betriebsstelle". These are specific points within the network. Every train station is also a operation control point. In the following, the word "station" will be used for operation control point.

3.1 Differences to the Long Island Rail Road System

The different topology of the railway network and the different timetables cause different problems that need to be faced:

- In [1], the length of every time interval is chosen to be 1 hour. Considering longer time intervals “flattens” the delays, since temporal peaks are divided by a higher number of trains. So shorter time intervals provide better “resolution” of the occurrence of delays.

On the other hand, it takes some time to propagate the delay within the network. In Austria, there are longer distances between stops and therefore it takes longer to propagate delays. Choosing the length of the time interval to be 1 hour, most trains will “occur” in more than one time interval. Since the different data sets (i.e. delays within different time intervals) are meant to be independent, this causes some problems.

- There are stations that are not passed by any train in certain time intervals. Calculating the average delay in that period would provoke dividing by zero. This problem can be solved by considering the totaled delay instead of the average delay.

However, this still might yield meaningful results, since the delays at a station that is approached by a lot of trains, each having just a little delay, might be more relevant than the delay at a station that is only approached by a single train, having a longer delay.

- The Long Island Rail Road System is quite crowded, i.e. if a train is delayed, it is very likely that this affects other trains. In Austria, there are
also parts where the temporal distance between different trains is much longer. In this case, delays will still propagate since a delayed train probably will still be delayed at its next stop, but it does not directly affect the delays of other trains.

- By considering only the eastbound or westbound trains, a cycle-free network was obtained in [1]. In Austria, a classification in eastbound or westbound is not possible (e.g. there are also trains going from north to south and vice versa). However, considering all the trains will lead to cycles within the network. This conflicts with the attempt of modelling the network as a directed acyclic graph. A strategy to resolve this problem is to delimit the number of trains considered.

- In the Long Island Rail Road System, there are just 124 stops, so it is a relatively small network. Considering all trains in Austria leads to a gigantic model (in terms of vertex numbers, edge numbers and parameters). But when examining only a selection of trains (and its corresponding stations), there are still influences from the non-considered trains. So there is "external" disturbing.

- There are also trains not operated by ÖBB using the same tracks. In the provided data there is only information about the trains operated by ÖBB. So even if no data selection is done, there will be still disturbances by other, not observed trains.

### 3.2 Methodology

Considering Subsection 3.1, the following restriction will be made:

- Only trains going from Vienna (station Hauptbahnhof or Westbahnhof) to St. Pölten and vice versa and the corresponding operational points between those two stations will be observed. In this case each train can be classified in eastbound or westbound and we obtain cycle-free networks. This section is also quite crowded, so there will be interactions between different trains.

- Since the distance between Vienna and St. Pölten is relatively short, we can set the length of the time intervals to 1 hour, without dealing too much with the problems mentioned in Subsection 3.1.

- We will not consider the average delay but the total delay of every time interval at each station.

Considering only the stations from Vienna to St. Pölten and vice versa is of course a very strong restriction but it enables us receiving suitable results. The considered stations (“Betriebsstellen”) are sketched in Figure 2. In Vienna, there are two branches - one branch going to Wien Westbahnhof and one branch going to Wien Hauptbahnhof. Between the stations Knoten Hadersdorf and Knoten Wagrain there are also two different routes: In the north (via Tullnerfeld) there is the so-called Neue Westbahn, in the south (via Pressbaum) there is the Alte Westbahn.

Figure 2: Considered stations (operation control points) between Vienna and St. Pölten

### 3.3 Network structure

Similar to [1], we use the R-package bnlearn (and the functions hc and bn.fit) to obtain the network structure. It was also considered to use the Python-packages Pomegranate and PyMC3 for the same task, but those packages can only deal with random variables that take on unordered values. The results can be seen in Figure 3 and Figure 4. In these figures, the directions of the dependencies were not indicated.

When searching for the graph structure, we expect a graph that resembles the geographical topology of the stations, e.g. it would not make sense if a station close to St. Pölten directly affects a station close to Vienna. The topological sorting never coincides with the actual geographical order of the stations. However, we did expect such a behavior, since phenomena like backward propagation (i.e. dependencies going in the opposite direction) and others can cause a mismatch with the geographical order of the stations. Furthermore, some stations are very close and therefore have almost the same delays. In this case, the order of those stations in the obtained network is almost arbitrary.

When observing the plots visually, we detect that there are also connections between stations that are far apart. There are even connections between stations from the Alte Westbahn and Neue Westbahn, whereat these kind of connections are relatively rare.
3.4 Metrics

The metrics are inspired by the metrics used in [1], but slightly modified.

The Inducer score indicates how much a delay at a station \( s \) influences the delays at all the other stations. Therefor, the delay at the examined station \( s \) is set to the median delay of this station (based on the data) and the delays at all the other stations are set to 0. Given this initial condition, the expected delay at every other station is estimated by the Bayesian network. Those values are averaged over all stations. The score of station \( s \) can be written as

\[
\text{Score}_I(s) = \frac{1}{n-1} \sum_{i=1, i \neq s}^n \mathbb{E}(X_i | X_s = x_s, X \setminus (X_s \cup X_i) = 0)
\]

where \( n \) is the number of random variables, \( x_s \) is the median (observed) delay at station \( s \) and \( X = (X_i)_{i \in \{1,...,n\}} \) is the ensemble of all random variables.

The Susceptible score indicates how much the delay at a station \( s \) is influenced by the delays at all the other stations. Therefor, the delay at one station \( i \) (with \( i \neq s \)) is set to its median delay (based on the data) and the delays at all the other stations (including the examined station \( s \)) are set to 0. Given this initial condition, the expected delay at the examined station is estimated by the Bayesian network. This procedure is repeated and averaged over all stations \( i \neq s \). Using the same notation as above, the score of station \( s \) can be written as

\[
\text{Score}_S(s) = \frac{1}{n-1} \sum_{i=1, i \neq s}^n \mathbb{E}(X_i | X_s = x_s, X \setminus (X_i \cup X_s) = 0)
\]

When considering the metrics, we observe that almost all stations with the highest (top 4) scores are either stations that are passed by all trains or stations of the \textit{Alte Westbahn}. Only the station \textit{Tfo} (close to Tullnerfeld) has the highest inducer score (westbound), although it is located at the \textit{Neue Westbahn}. So the \textit{Neue Westbahn} acts well in reducing delay propagation. Going in the westbound direction, two stations at the very beginning of the route reach high susceptible scores (\textit{Wien Hauptbahnhof} and \textit{Wien Westbahnhof Frachtenbahnhof}). Those stations are very susceptible to delays at other stations (although almost all stations are passed \textit{after} those two stations). We also observe that some stations reach negative susceptible scores. According to our model, this means that delays at certain other stations reduce the delay at this observed station.

3.5 Conclusion

When applying the methods introduced in [1] to the Austrian railway system, we need to make restrictions. In order to obtain the inevitable property of an acyclic directed graph, we can only consider an area where every train can be classified into a certain (cycle-free) direction. \( \mathbb{R} \) yields a network structure that seems partly reasonable. There are also connections between stations that are far apart and connections between \textit{Alte Westbahn} and \textit{Neue Westbahn}. However, this does not necessarily need to conflict with the real behaviour.

The inducer score and susceptible score indicate which stations play a big role in delay propagation. It seems reasonable that the \textit{Neue Westbahn} acts well w.r.t. reducing delay propagation, since this section was constructed quite recently. The fact that stations at the beginning of the route have quite high susceptible scores fits to the idea of back-propagation. However, there are also stations with negative susceptible scores. This does not seem to be reasonable.

References


Abstract. The ARGESIM benchmark CP2 consists of three different tasks to study current technologies for the parallelization of simulation programs, two of which have been addressed in a previous publication. The third one is the study of the fluid flow in a special geometry using the Lattice Boltzmann method. The task is studied with two methods for up to 256 cores, again using the MPI message passing library Open MPI and Matlab from The MathWorks in combination with the Parallel Computing Toolbox. Solutions with different grid sizes are compared with each other in terms of runtime and speedup. The Open MPI version generally shows good speedups even for large core numbers, while the Matlab version has poor results for 32 cores or more. On the other hand, the scalar Matlab version is several times faster than the Open MPI version, leading to a smaller runtime for up to 32 cores.

Introduction

In this article the solution of one task of the ARGESIM benchmark CP2 [1] will be presented. The benchmark contains three tasks, two of which have already been solved and presented in [2]. The remaining task, the Lattice Boltzmann simulation, is solved in this article. The solution is again implemented with the two technologies Open Message Passing Interface (Open MPI) [3] and Matlab Parallel Computing Toolbox (Matlab PCT)[4].

Since the description of the Lattice Boltzmann task in [1] is very sketchy, the basic method is explained in more detail in the first section, followed by a description of the scalar algorithm. Then the strategy for parallelization of this algorithm is presented. Finally, implementations in MPI and PCT are examined and compared.

The investigations were performed on an HPC system of the PHWT Institute called Seneca. The specifications of the system are shown in Table 1. All programs and scripts necessary to reproduce the results presented here are available from [5].

1 Introduction to Lattice Boltzmann Simulation

The Lattice Boltzmann Method (LBM) is a computational fluid dynamics method that has gained popularity, since it is especially well suited for parallel computers. In the following its basic ideas will be presented and the CP2 benchmark task described in full detail. This should make the task accessible to non-experts. A thorough introduction into the method and its connection to the familiar Navier-Stokes description can be found in [6].

LBM is based on the kinetic theory, which describes a fluid on a mesoscopic scale between the macroscopic description of a continuous fluid and the microscopic tracking of individual particles.
Its fundamental variable is the distribution function \( f(\vec{x}, \vec{\xi}, t) \), which represents the mass density of particles at position \( \vec{x} \), velocity \( \vec{\xi} \) and time \( t \). It is connected to the macroscopic mass density \( \rho(\vec{x}, t) \) and the macroscopic velocity \( \vec{u}(\vec{x}, t) \) by

\[
\rho(\vec{x}, t) = \int f(\vec{x}, \vec{\xi}, t) d^3\xi
\]

\[
\rho(x,t)\vec{u}(x,t) = \int \vec{\xi} f(\vec{x}, \vec{\xi}, t) d^3\xi.
\]

Using standard assumptions one can show that the distribution function will reach an equilibrium distribution given by

\[
f_{eq}(\vec{x}, \vec{\xi}, t) = \rho \left( \frac{1}{2\pi R_T} \right)^\frac{3}{2} e^{-\vec{\xi}^2/(2 R_T)}
\]

where the relative velocity \( \vec{\xi} := \vec{\xi} - \vec{u} \) is the deviation of the particle velocity from the macroscopic velocity. \( f \) satisfies the Boltzmann equation

\[
\frac{\partial f}{\partial t} + \vec{\xi} \cdot \frac{\partial f}{\partial \vec{x}} + \vec{F}_i \frac{\partial f}{\rho \partial \xi_i} = \Omega(f).
\]

Here, \( \vec{F} \) describes external forces, while \( \Omega \), the so called collision operator, is given by the forces between the particles during collisions. In LBM it is usually replaced by the BGK collision operator (named after its inventors Bhatnagar, Gross and Krook)

\[
\Omega(f) = -\frac{1}{\tau}(f - f_{eq}).
\]

which simply describes the evolution of \( f \) towards \( f_{eq} \). The corresponding relaxation time \( \tau \) is related to the macroscopic viscosity.

For the computation of \( f \) with LBM one discretizes space, time and velocity, choosing one of several schemes according to the dimension of the problem. Applying standard lattice units, the CP2 task uses a fixed time step \( \Delta t = 1 \) for \( t \), a two-dimensional square grid with \( \Delta x = \Delta y = 1 \) for \( x \) and a set \( \vec{c}_i, i = 0, \ldots, 8 \) of velocities \( \vec{\xi} \) (cf. Fig. 1), e. g. \( \vec{c}_0 = (0,0)^T \), \( \vec{c}_1 = (1,0)^T \), \( \vec{c}_5 = (1,1)^T \) etc. .

For simplicity one replaces the argument \( \vec{\xi} \) with an index corresponding to one of the nine possible velocities, writing

\[
f_i(\vec{x}, t) := f(\vec{x}, \vec{c}_i, t), \quad i = 0, \ldots, 8
\]

The macroscopic quantities are then given simply by

\[
\rho(\vec{x}, t) = \sum_i f_i(\vec{x}, t)
\]

\[
\rho(\vec{x}, t)\vec{u}(\vec{x}, t) = \sum_i \vec{c}_i f_i(\vec{x}, t).
\]

The equilibrium distribution can now be simplified to

\[
f_{eq}^i(\vec{x}, t) = \omega_i \rho \left( 1 + \frac{\vec{u} \cdot \vec{c}_i}{c_i^2} + \frac{(\vec{u} \cdot \vec{c}_i)^2}{2 c_i^4} - \frac{\vec{u} \cdot \vec{u}}{2 c_i^2} \right),
\]

\[
\text{Figure 1: Velocities used in discretization scheme.}
\]

\[ 
\begin{array}{cccccccc}
1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\
\end{array} 
\]
with the speed of sound \( c_s = 1/\sqrt{3} \) and the weights

\[
\mathbf{w}_i = \left(\frac{4}{9}, \frac{1}{9}, \frac{1}{9}, \frac{1}{9}, \frac{1}{9}, \frac{1}{9}, \frac{1}{9}\right)^T.
\]

The corresponding Lattice Boltzmann Equation (LBE) in the force-free case \( \mathbf{F} = \mathbf{0} \) is

\[
f_i(\vec{x} + \vec{c}_i, t + 1) = f_i(\vec{x}, t) + \mathbf{F}_i(\vec{x}, t),
\]

again using the BGK collision operator

\[
\mathbf{F}_i(f) = -\frac{1}{\tau}(f_i - f_i^{eq}),
\]

where the relaxation time is given by the viscosity \( \nu \):

\[
\tau = \frac{\nu}{c_s^2} + \frac{1}{2}.
\]

Finally one needs initial and boundary values and a few parameters. The CP2 task requests a quadratic lattice of size \( n_x \times n_y \) with \( n_x = 257 \) and boundary values describing a cavity flow that has been studied in [7]: The upper grid line has a constant velocity \( \vec{u}_0 = (0,1,0)' \), while all other boundary points are fixed walls (cf. Fig. 2).

![Figure 2: Definition of boundary values.](image)

The initial values of the macroscopic properties are defined as

\[
\begin{align*}
\rho(\vec{x},0) &= 1 \\
\vec{u}(\vec{x},0) &= \mathbf{0}.
\end{align*}
\]

The end time of the simulation is given as \( t_{end} = 350,000 \), when a steady state should have been reached. To adapt to the much larger processing power and memory size since the publication of [1], larger grids up to \( n_x = 2048 \) and end time \( t_{end} = 500,000 \) will be studied in the following.

At last, a Reynolds number of \( \text{Re} = 1000 \) is given, which leads to a viscosity of

\[
\nu = \frac{(n_x - 1)u_0,1}{\text{Re}} = 0.0256.
\]

### 2 Basic Scalar Implementation

The scalar implementation follows the lines of the basic algorithm described in [6]. The essential idea is to split the LBE in two parts, the collision step

\[
f_i^c(\vec{x},t) = f_i(\vec{x},t) - \frac{1}{\tau}(f_i(\vec{x},t) - f_i^{eq}(\vec{x},t)),
\]

which is a local operation at a point, and the propagation step

\[
f_i(\vec{x} + \vec{c}_i, t + 1) = f_i^c(\vec{x},t),
\]

which transports the distribution to neighbouring points. Initially the distribution function is set to the equilibrium values given by \( \rho(\vec{x},0) \) and \( \vec{u}(\vec{x},0) \). Then the following computations are performed in a loop over time steps:

1. computation of \( \rho \) and \( \vec{u} \) from \( f_i \),
2. computation of \( f_i^{eq} \) from \( \rho \) and \( \vec{u} \),
3. collision step to get \( f_i^c \),
4. propagation step to get \( f_i(\vec{x},t+1) \).

During the collision step the boundary behaviour is implemented: At the top line \( f \) is set to the equilibrium value defined by the given \( \vec{u}_0 \); at the wall points the full-way bounce-back method is used, which reverses the velocity of an incoming distribution, i.e. it sets

\[
f_i^c(\vec{x},t) = f_i(\vec{x},-\vec{c}_i,t).
\]

After the time loop the relative magnitude \( |\vec{u}|/|\vec{u}_0| \) of the macroscopic velocity is computed and plotted, as is requested in the benchmark. Fig. 3 shows the result for the standard benchmark parameters.

### 3 Parallel Lattice Boltzmann Simulation

In this chapter, a parallelization strategy for the Lattice Boltzmann simulation is introduced and investigated on the HPC system Seneca using two technologies.
3.1 Parallelization Strategy

To process the Lattice Boltzmann simulation in parallel, the set of cores is virtually arranged in a two-dimensional cartesian coordinate system as shown in Figure 4. The data grid $f$ is distributed equally along this grid, so that each core contains a section of the entire grid according to its coordinates. Except for the propagation step, all calculations on such a section can be done independently by each core.

To simplify the decomposition, the data grid size $n_x$ is restricted to a power of two in the implementation, e.g. 32x32, 64x64, etc. The number of cores $n_C$ is subject to the same condition so that the entire grid can be divided equally.

Therefore, $n_C$ was varied from 1, 2, 4, ..., 256 in the following measurements. Of course, an uneven distribution could be implemented easily. But a fine-grained study of the effects of the total core number would reveal more about the memory architecture and the interconnect structure of the hardware than about the properties of the high-level tools we are interested in.

In the propagation step, displacements are performed in the north, east, south and west directions, as well as the four diagonal directions. In these shifts, data must be communicated with neighboring cores. For horizontal or vertical shifts, there is at most one neighbor to send and one to receive and complete edges of the local matrix are sent or received. Data points at the global edges, which do not receive values through a shift, retain their current values. The diagonal shifts are a bit more complicated because there are more communication partners here (cf. Fig. 5). At first sight, these operation can be simplified by combining two of the simple shift operations. For example, a shift to the north-east can be realized by shifting first to the east, then to the north. But actually, this leads to wrong values at the outer edge points, as can be seen in Fig. 6.

A different approach is used by the ScaLAPACK library [8]: There the entire grid is divided into smaller matrices of fixed blocksize, which are then distributed cyclically across the core grid. Since this seems to lead to a much more involved communication pattern in our application, it has not been followed here.
receive function resulting in a null operation. This value can be used
safely in a send or turn.

stricted to shifts along the cartesian axes. It simply re-
vided into eight functions, one function for each di-
rection. To identify the communication partners in
maining of the data. This allowed the grid segments to be
associated with the correct points in the global grid.

For the diagonal shifts, the coordinates of a
core in the core grid and the corresponding target
and source coordinates were determined with
MPI_Cart_coords. They were then used to deter-
mine the corresponding ranks for the send and receive
operations with MPI_Cart_rank. It is important to
use this function only with valid coordinates, otherwise
the program terminates with an error. To prevent this,
one can again use the MPI_Cart_get function.

To collect the data at the end of the simulation the
function MPI_Cart_coords was used to define the
core with coordinates (0, 0) as the master and for sort-
ing of the data. This allowed the grid segments to be
associated with the correct points in the global grid.

3.3 Matlab PCT

The second parallel implementation was created using
Matlab and the Parallel Computing Toolbox (PCT) [4].
A very short introduction to Matlab/PCT and the differ-
ent parallelization strategies it supplies can be found in
[2]. The underlying scalar version was taken from [9],
which already compares different earlier parallel tool-
boxes from Mathworks and other authors.

For the task studied here, the PCT concept of codis-
distributed arrays seems to be ideal: The data array is
distributed to the workers automatically, and the local
counters can still use the global index to address all array
points. If an element is not available locally, it is sent
immediately to the accessing worker. Unfortunately,
the corresponding functions are badly documented and
– more importantly – show an extremely bad perfor-
ance [2]. Therefore, the spmd construct was used in
the implementation, which creates a parallel section
that is executed by each worker. This leads to a pro-
gramming style that is similar to the MPI implementa-
tion. Since PCT does not provide functions for creat-
ing a two-dimensional core grid and to map cores to a
grid, this functionality was implemented in the appli-
cation using the functions numlabs and labindex.

For convenience, an array was supplied that contains
the positions of all labindex values in the core grid.

For the propagation step, separate functions were
created for each of the eight directions. To determine
the communication partners a function
cart2D_shift was written, which determines the
rank of the target and the source from a direction vec-
tor and a displacement. This makes it easy to deter-
mine the communication partners. The sending and re-
ceiving of the data was realized with labSend and
labReceive.
After the `spmd` section the local arrays are directly available in one cell array. Therefore the merging of the results could be realized simply by rearranging all velocity data in a global array.

3.4 Comparison of solutions

The two implementations were tested on Seneca and the runtime was analyzed depending on the grid size and the number of cores. The grid size was selected from 256x256 to 2048x2048, the number of cores from 1 to 256. Since Seneca consists of three nodes with 96 cores each, runs with up to 64 cores were computed on one node using local memory only, while for 128 and 256 cores two and three nodes were utilised, and data had to be sent across the high-speed network. For small numbers of cores the number of iterations has been reduced to make the computing times feasible. The timing results have been scaled afterwards to the total number of iterations for easier comparison. The scalar overhead of the computations was negligible in these cases.

Figure 7 shows the runtime and speedup results of the Open MPI implementation. The most basic model would assume runtimes that fall with \(1/n_C\) and rise with \(n_x^2\), which in a log-log representation would lead to parallel decreasing straight lines with constant separation. The runtime plots very roughly show this behaviour, especially for larger grid size. The deviations can be better analyzed in the speedup plot, in particular when comparing the results with the ideal case of linear speedup. Here the occurring superlinear speedup can be seen clearly, which is usually due to the memory architecture of the system: With increasing core number, the local memory decreases, until it fits in a fast cache memory. This interpretation is supported by the observation that the point, where superlinear speedup begins, rises with grid size and core number in such a way that the local memory size stays constant. Another expected behaviour is the drop of speedup for large core numbers, when more than one node is used. This effect gets smaller with rising grid size, because the amount of computation dominates the rising communication times. The results suggest that for large grid sizes an even higher speedup can be reached with more cores. In Figure 8 the results of the Matlab implementation are presented.

Looking at the speedups, Matlab PCT reaches good to reasonable values for up to 16 cores, but very poor results for more cores. The drop in performance at 128 cores can be seen again, except for a very peculiar rise at the grid size \(n_x = 2048\). But when comparing the runtimes, the image changes completely: The table 2, which displays the runtimes for the scalar case \(n_C = 1\), shows the impressive performance of Matlab compared to the direct C implementation. This is probably due to the efficient matrix operations Matlab is based on.

<table>
<thead>
<tr>
<th>grid size</th>
<th>runtime [min]</th>
<th>runtime [min]</th>
<th>ratio</th>
</tr>
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<tbody>
<tr>
<td></td>
<td>Open MPI</td>
<td>Matlab PCT</td>
<td></td>
</tr>
<tr>
<td>256</td>
<td>274.4</td>
<td>36.1</td>
<td>7.59</td>
</tr>
<tr>
<td>512</td>
<td>906.3</td>
<td>175.6</td>
<td>5.16</td>
</tr>
<tr>
<td>1024</td>
<td>4297.3</td>
<td>1173.1</td>
<td>3.66</td>
</tr>
<tr>
<td>2048</td>
<td>14757.2</td>
<td>4347.6</td>
<td>3.39</td>
</tr>
</tbody>
</table>

Table 2: Comparison of runtimes for \(n_C = 1\).
Figure 8: Runtime and speedup of the Matlab PCT implementation for different grid sizes.

Figure 9 shows the runtime ratios of the parallel Open MPI and Matlab PCT versions, compared for the same number of cores. Apart from the case of small grid size, Matlab is faster for up to 32 cores, and for the larger grids even up to 64 cores. It takes a lot of cores, until the better speedup of Open MPI catches up with the better scalar performance of Matlab PCT.

4 Conclusion

The goal of the CP2 benchmark cannot be to compare latest parallelization strategies for current hardware architectures – this would require much more complex algorithms such as those studied in the HPCG benchmark [10]. Instead, the underlying question of the benchmark is: How can non-experts use tools and computing environments to take advantage of modern multi-core CPUs? Correspondingly, the MPI implementations of the three CP2 tasks that have been presented here and in [2] mainly serve as points of reference and use well known techniques, whereas the parallelization methods and outcomes provided by a scientific computing environment such as Matlab are the main point of interest.

In this respect Matlab PCT is a very interesting approach, but the results presented here show that it doesn’t quite live up to its prospects. On the one hand the spmd construct and standard message-passing methods make it possible to reach reasonable speedups for small core numbers, but require knowledge of standard parallelization techniques. Furthermore, the parfor command provides a convenient tool for very simple cases like Monte Carlo studies. On the other hand, the promising concept of codistributed arrays provides a very elegant way to cope with the distribution and parallel processing of large matrices, but it has shown very poor performance even in the simple PDE task.

For very compute-intensive applications, which benefit most from larger core numbers, one typically would resort to standard languages like C or C++, together with optimizing compilers. But as we have seen in the PDE and LBM tasks, Matlab can be compatible or even better in the scalar case, due to its highly optimized mathematical libraries. Therefore it would be very desirable, if Matlab PCT could sustain good speedups to larger numbers of cores.

In spite of its interesting topic, the ARGESIM CP2 benchmark has not been addressed before. To some extent, this might be due to the rather short description of the Lattice Boltzmann task, which needs background reading to understand the algorithm at all. Hopefully, the short introduction provided here can help as a starting point for further studies.
Another problem of CP2 is the small problem size. To make the tasks challenging on modern architectures, one has to scale up the problem sizes considerably. Maybe, one could modernize this benchmark by using scaling methods instead of fixed sizes, and reissue it as a new entry in the standard ARGESIM benchmark suite, giving up on the idea of a special “parallel benchmarks” series.

References


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KA-SIM Kosovo Simulation Society

The Kosova Association for Modeling and Simulation (KA-SIM) is closely connected to the University for Business and Technology (UBT) in Kosovo.

President Edmond Hajrizi, ehajrizi@ubt-uni.net
Vice President Muzaffer Shala, info@ka-sim.com

LIOPHANT Simulation

LIOPHANT Simulation is a non-profit association born in order to be a trait-d’union among simulation developers and users; LIOPHANT is devoted to promote and diffuse the simulation techniques and methodologies; the Association promotes exchange of students, sabbatical years, organization of International Conferences, courses and internships focused on M&S applications.

President A.G. Bruzzone, agostino@itim.unige.it
Director E. Bocca, enrico.bocca@liophilant.org

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LSS – Latvian Simulation Society

The Latvian Simulation Society (LSS) has been founded in 1990 as the first professional simulation organisation in the field of Modelling and simulation in the post-Soviet area.

President Yuri Merkuryev, merkur@itl.rtu.lv
Vice President Egils Ginters, egils.ginters@rtu.lv

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Prof. Yuri Merkuryev, Dept. of Modelling and Simulation, Riga Technical University, Kalku street 1, Riga, LV-1658, Latvia

NSSM – National Society for Simulation Modelling (Russia)

NSSM – The National Society for Simulation Modelling (Национальное Общество Имитационного Моделирования – НОИМ) was officially registered in Russia in 2011.

President R. M. Yusupov, yusupov@iias.spb.su
Chairman A. Plotnikov, plotnikov@sstc.spb.ru

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NSSM / R. M. Yusupov, St. Petersburg Institute of Informatics and Automation RAS, 199178, St. Petersburg, 14th line, h. 39

PTSK – Polish Society for Computer Simulation

PTSK is a scientific, non-profit association of members from universities, research institutes and industry in Poland with common interests in variety of methods of computer simulations and its applications.

President Tadeusz Nowicki, Tadeusz.Nowicki@wat.edu.pl
Vice President Leon Bobrowski, leon@ibib.waw.pl

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PTSK, 00-908 Warszawa 49, ul. Gen. Witolda Urbanowicza 2, pok. 222
SIMS – Scandinavian Simulation Society

SIMS is the Scandinavian Simulation Society with members from the five Nordic countries Denmark, Finland, Norway, Sweden and Iceland. The SIMS history goes back to 1959.

<table>
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<tr>
<th>President</th>
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vadime@wolfram.com

Vadim Engelson, Wolfram MathCore AB, Teknikringen 1E, 58330, Linköping, Sweden

SLOSIM – Slovenian Society for Simulation and Modelling

The Slovenian Society for Simulation and Modelling was established in 1994. It promotes modelling and simulation approaches to problem solving in industrial and academic environments by establishing communication and cooperation among corresponding teams.

<table>
<thead>
<tr>
<th>President</th>
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SLOSIM, Fakulteta za elektrotehniko, Tržaška 25, SI-1000, Ljubljana, Slovenija

UKSIM – United Kingdom Simulation Society

The UK Modelling & Simulation Society (UKSim) is the national UK society for all aspects of modelling and simulation, including continuous, discrete event, software and hardware.

<table>
<thead>
<tr>
<th>President</th>
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david.al-dabass@ntu.ac.uk

- UKSIM / Prof. David Al-Dabass, Computing & Informatics, Nottingham Trent University, Clifton lane, Nottingham, NG11 8NS, United Kingdom

Observer Members

ROMSIM – Romanian Modelling and Simulation Society

ROMSIM has been founded in 1990 as a non-profit society, devoted to theoretical and applied aspects of modelling and simulation of systems.

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ROMSIM / Florin Hartescu, National Institute for Research in Informatics, Averescu Av. 8 – 10, 011455 Bucharest, Romania

ALBSIM – Albanian Simulation Society

The Albanian Simulation Society has been initiated at the Department of Statistics and Applied Informatics, Faculty of Economy at the University of Tirana, by Prof. Dr. Kozeta Sevrani.

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goizeta.sevrani@unitir.edu.al

Albanian Simulation Group, attn. Kozeta Sevrani, University of Tirana, Faculty of Economy, rr. Elbasanit, Tirana 355, Albania

Former Societies / Societies in Re-organisation

- CROSSIM – Croatian Society for Simulation Modelling
  Contact: Tarzan Legović, Tarzan.Legovic@irb.hr
- FrancoSim – Société Francophone de Simulation
- HSS – Hungarian Simulation Society
  Contact: A. Gábor, andrasi.gabor@uni-bge.hu
- ISCS – Italian Society for Computer Simulation

The following societies have been formally terminated:

ARGESIM is a non-profit association generally aiming for dissemination of information on system simulation – from research via development to applications of system simulation. ARGESIM is closely co-operating with EUROSIM, the Federation of European Simulation Societies, and with ASIM, the German Simulation Society. ARGESIM is an ‘outsourced’ activity from the Mathematical Modelling and Simulation Group of TU Wien, there is also close co-operation with TU Wien (organisationally and personally).

→ www.argesim.org
→ office@argesim.org

→ ARGESIM/Math. Modelling & Simulation Group, Inst. of Analysis and Scientific Computing, TU Wien Wiedner Hauptstrasse 8-10, 1040 Vienna, Austria Attn. Prof. Dr. Felix Breitenecker

ARGESIM is following its aims and scope by the following activities and projects:

- Publication of the scientific journal SNE – Simulation Notes Europe (membership journal of EUROSIM, the Federation of European Simulation Societies) – www.sne-journal.org
- Organisation and Publication of the ARGESIM Benchmarks for Modelling Approaches and Simulation Implementations
- Publication of the series ARGESIM Reports for monographs in system simulation, and proceedings of simulation conferences and workshops
- Publication of the special series FBS Simulation – Advances in Simulation / Fortschrittsberichte Simulation - monographs in co-operation with ASIM, the German Simulation Society
- Support of the Conference Series MATHMOD Vienna (triennial, in co-operation with EUROSIM, ASIM, and TU Wien) – www.mathmod.at
- Administration of ASIM (German Simulation Society) and administrative support for EUROSIM www.eurosim.info
- Simulation activities for TU Wien

ARGESIM is a registered non-profit association and a registered publisher: ARGESIM Publisher Vienna, root ISBN 978-3-901608-xx-y and 978-3-903347-xx-y, root DOI 10.11128/z...zz.zz. Publication is open for ASIM and for EUROSIM Member Societies.

SNE – Simulation Notes Europe

The scientific journal SNE – Simulation Notes Europe provides an international, high-quality forum for presentation of new ideas and approaches in simulation – from modelling to experiment analysis, from implementation to verification, from validation to identification, from numerics to visualisation – in context of the simulation process. SNE puts special emphasis on the overall view in simulation, and on comparative investigations. Furthermore, SNE welcomes contributions on education in/for/with simulation.

SNE is also the forum for the ARGESIM Benchmarks on Modelling Approaches and Simulation Implementations publishing benchmarks definitions, solutions, reports and studies – including model sources via web.

SNE Editorial Office /ARGESIM
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Irmgard Husinsky (Web, Electronic Publishing)
Felix Breitenecker EiC (Organisation, Authors)
ARGESIM/Math. Modelling & Simulation Group, Inst. of Analysis and Scientific Computing, TU Wien Wiedner Hauptstrasse 8-10, 1040 Vienna, Austria

SNE, primarily an electronic journal, follows an open access strategy, with free download in a basic version (B/W, low resolution graphics). SNE is the official membership journal of EUROSIM, the Federation of European Simulation Societies. Members of (most) EUROSIM Societies are entitled to download the full version of e-SNE (colour, high-resolution graphics), and to access additional sources of benchmark publications, model sources, etc. (group login for the ‘publication-active’ societies; please contact your society). Furthermore, SNE offers EUROSIM Societies a publication forum for post-conference publication of the society’s international conferences, and the possibility to compile thematic or event-based SNE Special Issues.

Simulationists are invited to submit contributions of any type – Technical Note, Short Note, Project Note, Educational Note, Benchmark Note, etc. via SNE’s website:
Proceedings Langbeiträge ASIM Workshop 2023 STS/GMMS/EDU - ASIM Fachgruppenworkshop 2023
Univ. Magdeburg, März 2023; C. Krull; W. Commerell, U. Durak, A. Köerner, T. Pawletta (Hrsg.).

Kurzbeiträge & Abstract-Beiträge ASIM Workshop 2023 STS/GMMS/EDU - ASIM Fachgruppenworkshop 2023
Univ. Magdeburg, März 2023; C. Krull; W. Commerell, U. Durak, A. Köerner, T. Pawletta (Hrsg.).

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Mathematical Modeling for New Insights into Epidemics by Herd Immunity and Serotype Shift. Florian Miksch, FBS 20;

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- Simulation for Digital Twins in Industry and Logistics – Chair: Francesco Longo/ Antonio Padovano, University of Calabria
- Methodology and Risk Assessment – Prof. Edmond Hajrizi, UBT
- Supply Chain Management, Logistics and Transportation – Chair M. Bagamanova, AUAS – Peter Meincke, DLR
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- Complex, Intelligent, Adaptive and Autonomous Systems (CIAAS)
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- Gaming
- Epidemiological
- PhD & MSc/Bach Award – Chairs Dr. Luis Manuel Fernández-Ahumada, U. of Córdoba – Dr Miguel Mujica Mota, Amsterdam U. of Applied Sciences
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- Simulation in Human Behaviour – Chairs: Jose Maria Ortiz-Gomez, Zayed University/ Christina Rott, Vrije Universiteit Amsterdam
- Simulation in Education – Cristina Ruiz Martin, Carleton University
- Agent-based Simulation – Hai Xiang Lin, TU Delft
- Energy Transition
- Aviation Track – Chairs: Benedikt Badanik, U. of Zilina -Miguel Mujica Mota, AUAS- Bruno Desart EUROCONTROL
- Monitoring and Control – Chairs: Dr. Igor Škrjanc, University of Ljubljana
- Healthcare Applications – Chair: Nikki Popper, TU WIEN
- Hybrid Simulation
- Manufacturing Applications – Chairs Dr. Antonella Petrillo& Dr. Fabio de Felice, U.of Napoli “Parthenope”
- Military Applications – Chair Oliver Rose, Bundeswehr University
- Advanced Tutorials/workshops
- Applications in Engineering and Architecture –
- Financial Simulation
- Extraction Industries
- Poster Session – Chair Dr. Paolo Scala, Fedex

www.eurosimm2023.eu
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*Prof. Dr. Thomas S. Spengler, Lehrstuhl für Produktion und Logistik; TU Braunschweig*

**Digital Twins – A Journey from Particle Physics at CERN to Industry 4.0 Manufacturing in Singapore**  
*Dr. Peter Lendermann, D-SIMLAB Technologies Pte Ltd, Singapore*

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