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Integrated Processes for Modelling & Simulation Health Care Systems

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Director DEXHELPP – Decision Support for Health Policy and Planning, Vienna

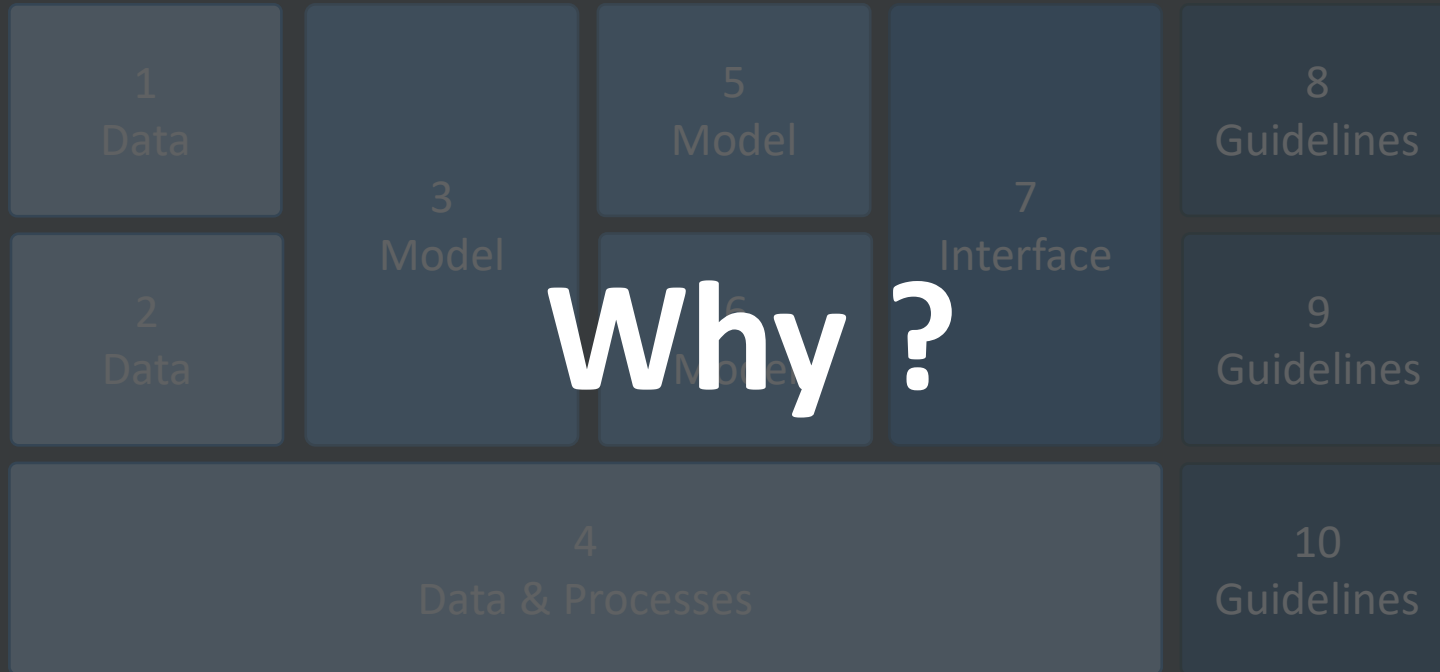
Coordinator Centre for Computational Complex Systems, TU Wien

CSO dwh GmbH – Simulation Services and Technical Solutions, Vienna

deXhelpp

Decision Support for Health Policy and Planning





10 Concepts to Integrate

Why?

- > Dynamics & Complexity
- > Economics
- > Big Data

How?

- > 4th Paradigm
- > Modelling Dynamics
- > Data vs. Models
- > Processes

Integrated Processes for **Modelling & Simulation**



Man schützt das Klima am besten, indem man kinderlos bleibt, 58,6 Tonnen CO₂ pro Jahr lassen sich so einsparen, heißt es. So ein Unsinn!
by Gregor Walter-Drop, DIE ZEIT, 21. März 2019, Nr. 13/2019

Strange Decision Support

“We recommend four widely applicable high-impact (i.e. low emissions) actions with the potential to contribute to systemic change and substantially reduce annual personal emissions: having one fewer child (an average for developed countries of 58.6 tonnes CO₂-equivalent (tCO₂e) emission reductions per year), living car-free (2.4 tCO₂e saved per year), avoiding airplane travel (1.6 tCO₂e saved per roundtrip transatlantic flight) and eating a plant-based diet (0.8 tCO₂e saved per year). “

Seth Wynes and Kimberly A Nicholas 2017 Environ. Res. Lett. 12 074024

Kritik: „...Modellannahmen gehen von falschen Annahmen aus, stellen Fakten absurd zusammen um auf die Ergebnisse zu kommen“



The Seven Tools of Causal Inference, with Reflections on Machine Learning by Judea Pearl, Communications of the ACM, March 2019, Vol. 62 No. 3, Pages 54-60, 10.1145/3241036

Causal Inference refl. ML

„...Machine learning researchers have noted current systems lack the ability to recognize or react to new circumstances they have not been specifically programmed or trained for.... „

„...Another obstacle is "explainability," or that "machine learning models remain mostly black boxes" unable to explain the reasons behind their predictions or recommendations, thus eroding users' trust..."

Classification of causal information in terms of the kind of questions each class is capable of answering:

Level (Symbol)	Typical Activity	Typical Questions	Examples
1. Association $P(y x)$	Seeing	What is? How would seeing X change my belief in Y?	What does a symptom tell me about a disease? What does a survey tell us about the election results?
2. Intervention $P(y do(x), z)$	Doing, Intervening	What if? What if I do X?	What if I take aspirin, will my headache be cured? What if we ban cigarettes?
3. Counterfactuals $P(y_x x', y')$	Imagining, Retrospection	Why? Was it X that caused Y? What if I had acted differently?	Was it the aspirin that stopped my headache? Would Kennedy be alive had Oswald not shot him? What if I had not been smoking the past two years?

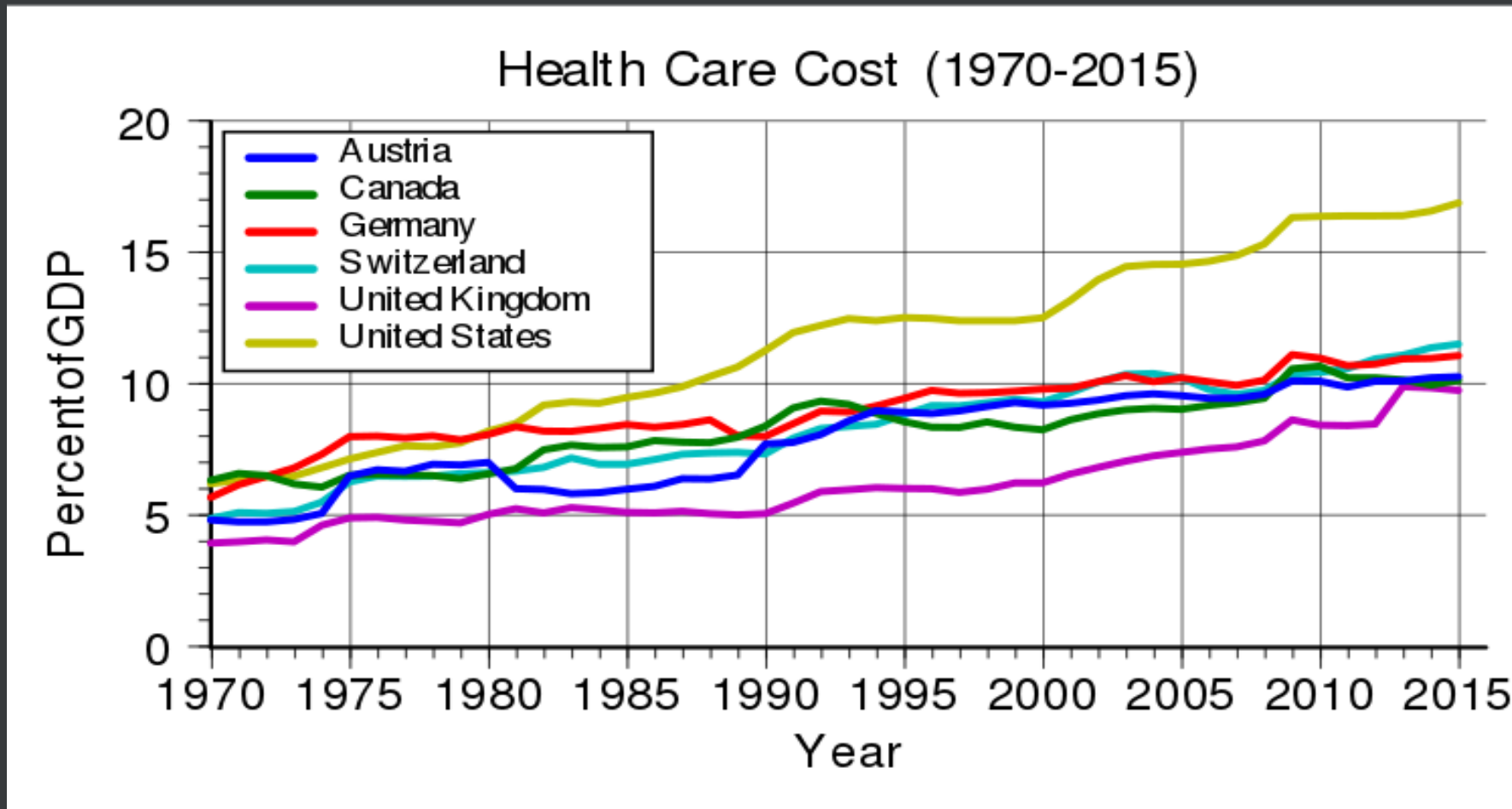
Economics Example Health Systems

Dissociation of Prooved Benefit of Costs for Medical Care

Pharma- ceutical	Application	Discovery	Costs € (Patient/Yea r)	Effects (Survival)
Insulin	Typ 1 Diabetes	1920	500	Decades
Statins	Cardiology	1990	5.000	Years
Monoclonal Antibodies	Onkology	2000	50.000	Month/Weeks
Encyme Alternation	Metabolism	2010	500.000	???

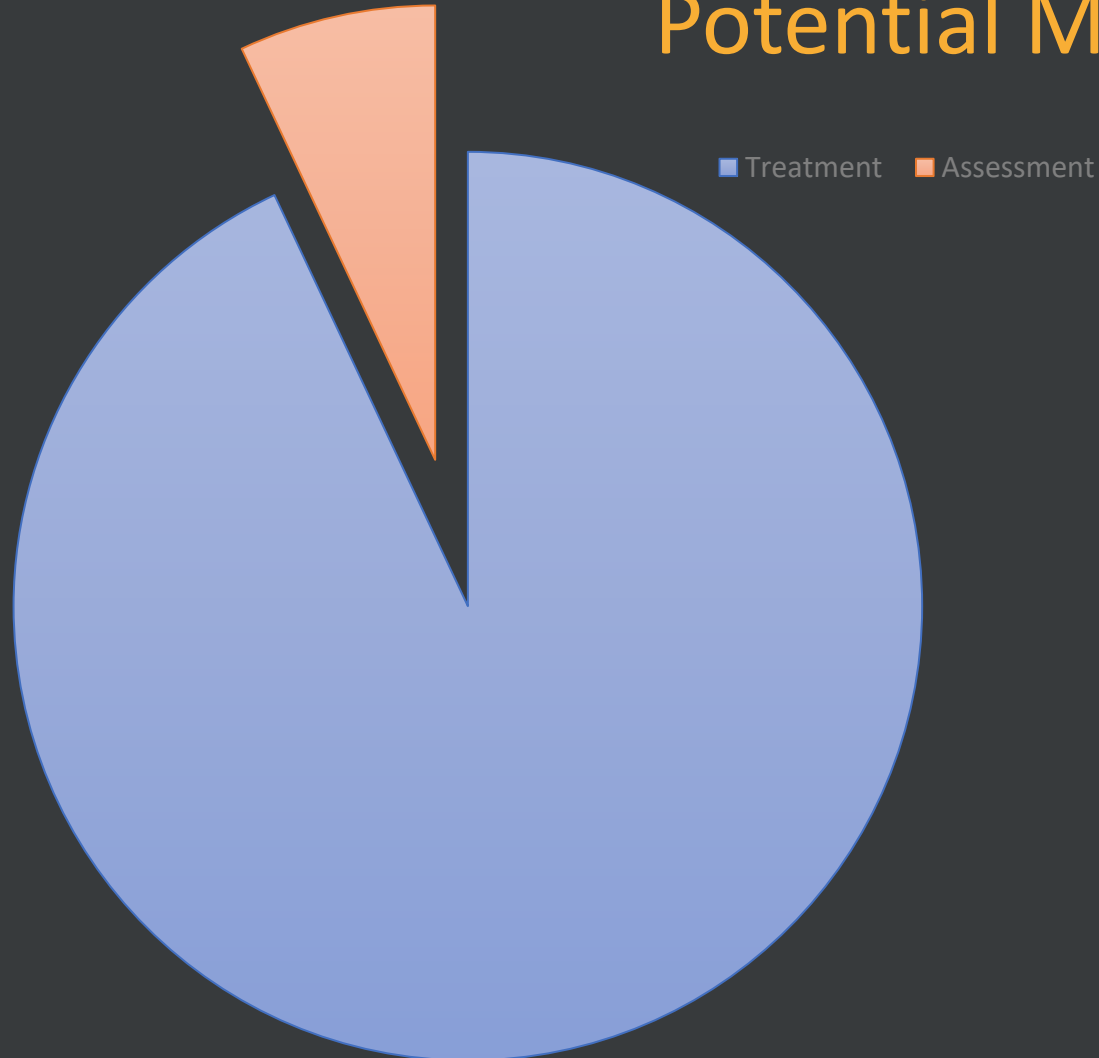
Source: C. Wild, Ludwig Boltzmann Institute for Health Technology Assessment, Vienna

Economics Example Health Systems



https://en.wikipedia.org/wiki/Healthcare_in_Austria

Health Care Assessment of Interventions Potential Market



Austria – 2-3 Billion €

UK – 13-20 Billion €

Germany – 17-27 Billion €

Development vs. Measuring

Advanced Treatment

Dealing with Increasing Costs

New Medication

Estimating Reachable Outcomes

Cutting Edge Technology

Measuring of Adherence

Innovative Diagnosis

Managing # of Patients

Fair Distribution of Health Services

Developing Registers

Disease Prevention

Benchmarking Interventions

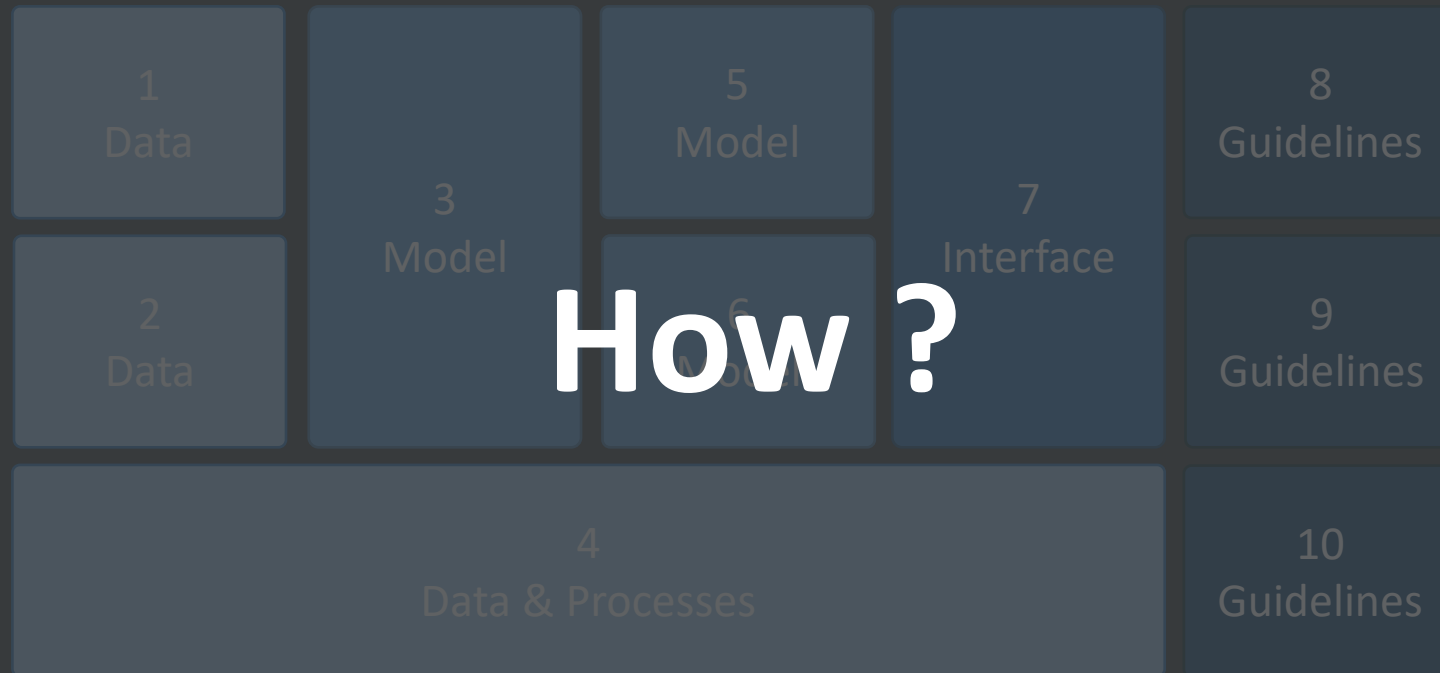
Economical Need

Combination of Methods

Complex & Dynamic Processes

Increasing & Complex Data

Integrated Processes for **Modelling & Simulation**



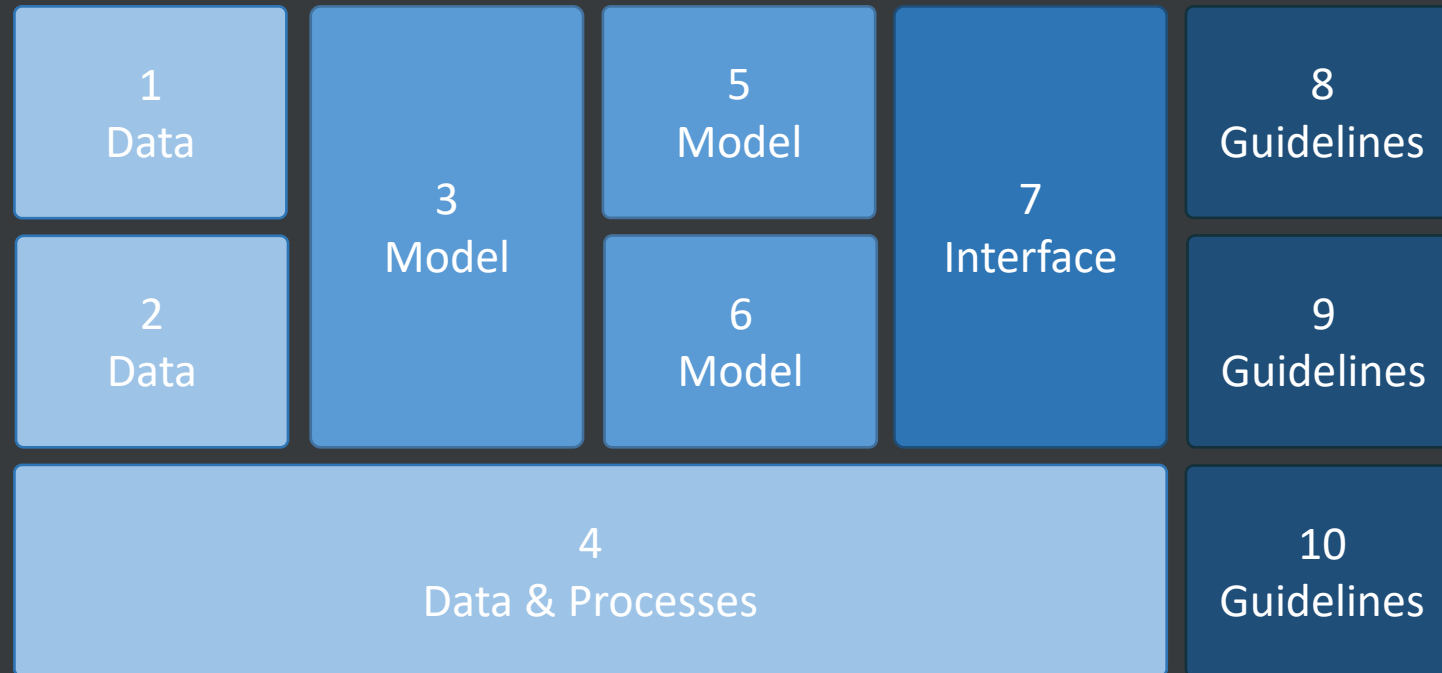
Integrated Processes for **Modelling & Simulation**

Data Driven System Simulation

10 Concepts to Integrate:
*Implementing Future Simulation
Models, including Data Processes*

<https://www.eurosim.info/tcs/tc-ddss/>

Integrated Processes for **Modelling & Simulation**



Integrated Processes for **Modelling & Simulation**

The Fourth Paradigm by Jim Gray (Microsoft)

10 Concepts to Integrate

Why?

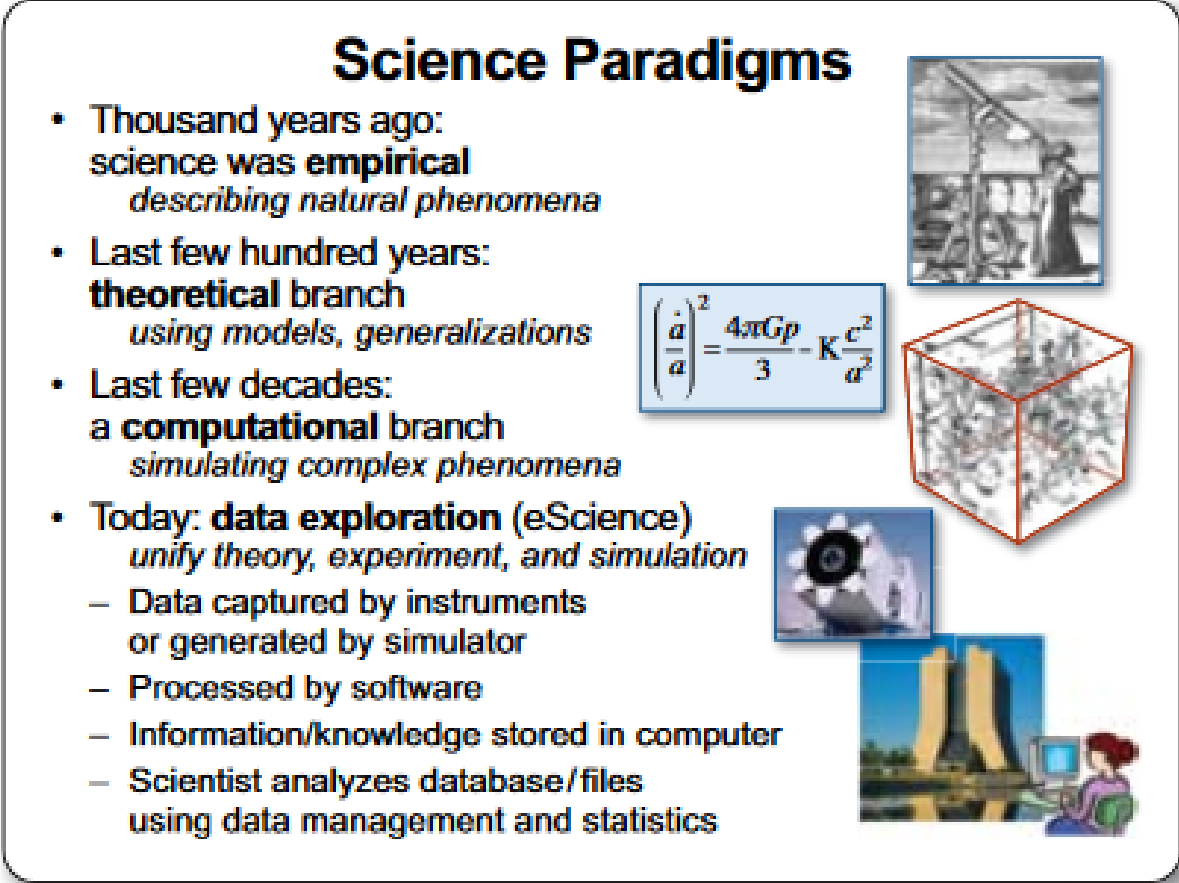
- > Dynamics & Complexity
- > Economics
- > Big Data

How?

- > 4th Paradigm
- > Modelling Dynamics
- > Data vs. Models
- > Processes

Science Paradigms

- **Thousand years ago:**
science was **empirical**
describing natural phenomena
- **Last few hundred years:**
theoretical branch
using models, generalizations
- **Last few decades:**
a computational branch
simulating complex phenomena
- **Today: data exploration (eScience)**
unify theory, experiment, and simulation
 - Data captured by instruments or generated by simulator
 - Processed by software
 - Information/knowledge stored in computer
 - Scientist analyzes database/files using data management and statistics



Modelling Dynamics

10 Concepts to Integrate

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#

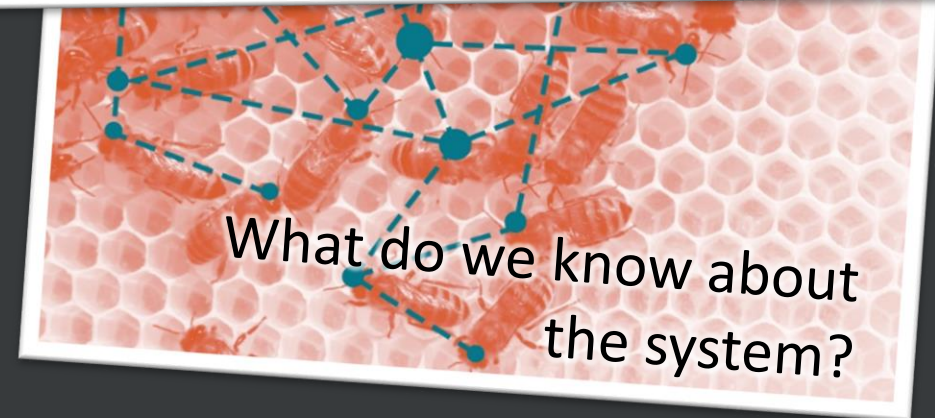
Which **Questions**
are to be answered?

#

Which **Data**
Ressources are
available?

#

Which **Systems &**
Processes are
described ?



Modelling Dynamics

10 Concepts to Integrate

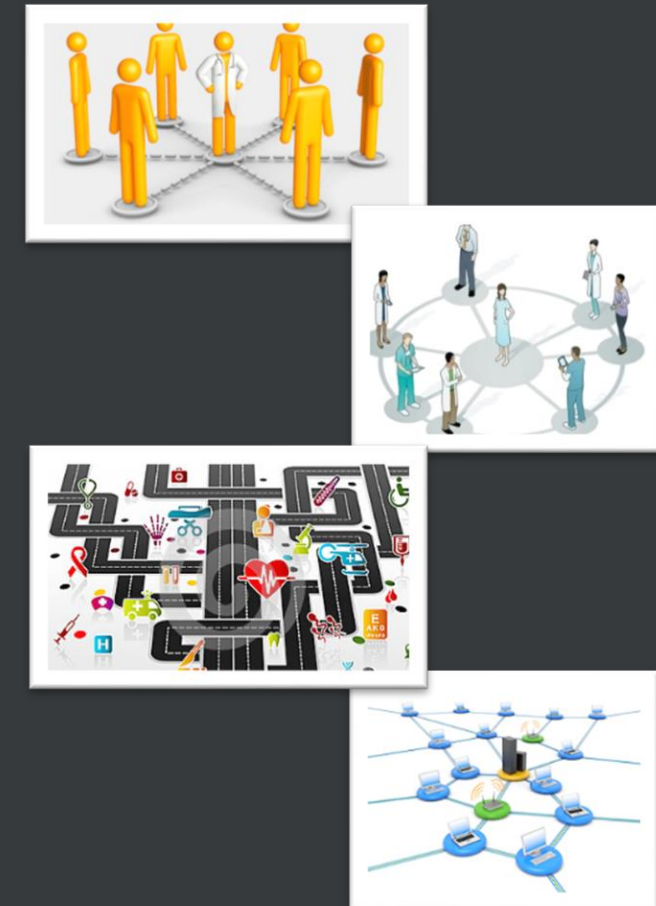
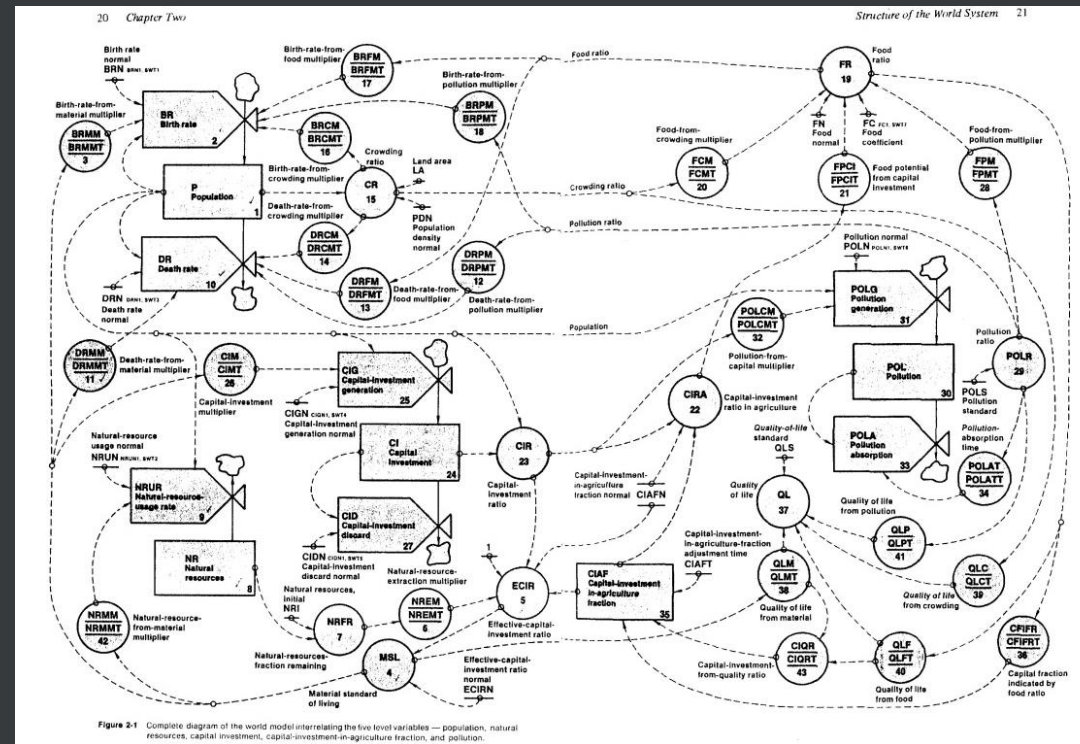
Why?

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How?

- > 4th Paradigm
- > Modelling Dynamics
- > Data vs. Models
- > Processes

Structures & Knowledge



Forrester, Jay W., *World Dynamics*. 1973 second ed. 1971, Waltham, MA: Pegasus Communications. P. 144, Reproduced by permission of Jay W. Forrester.

Data vs. Models in Health System Research

10 Concepts to Integrate

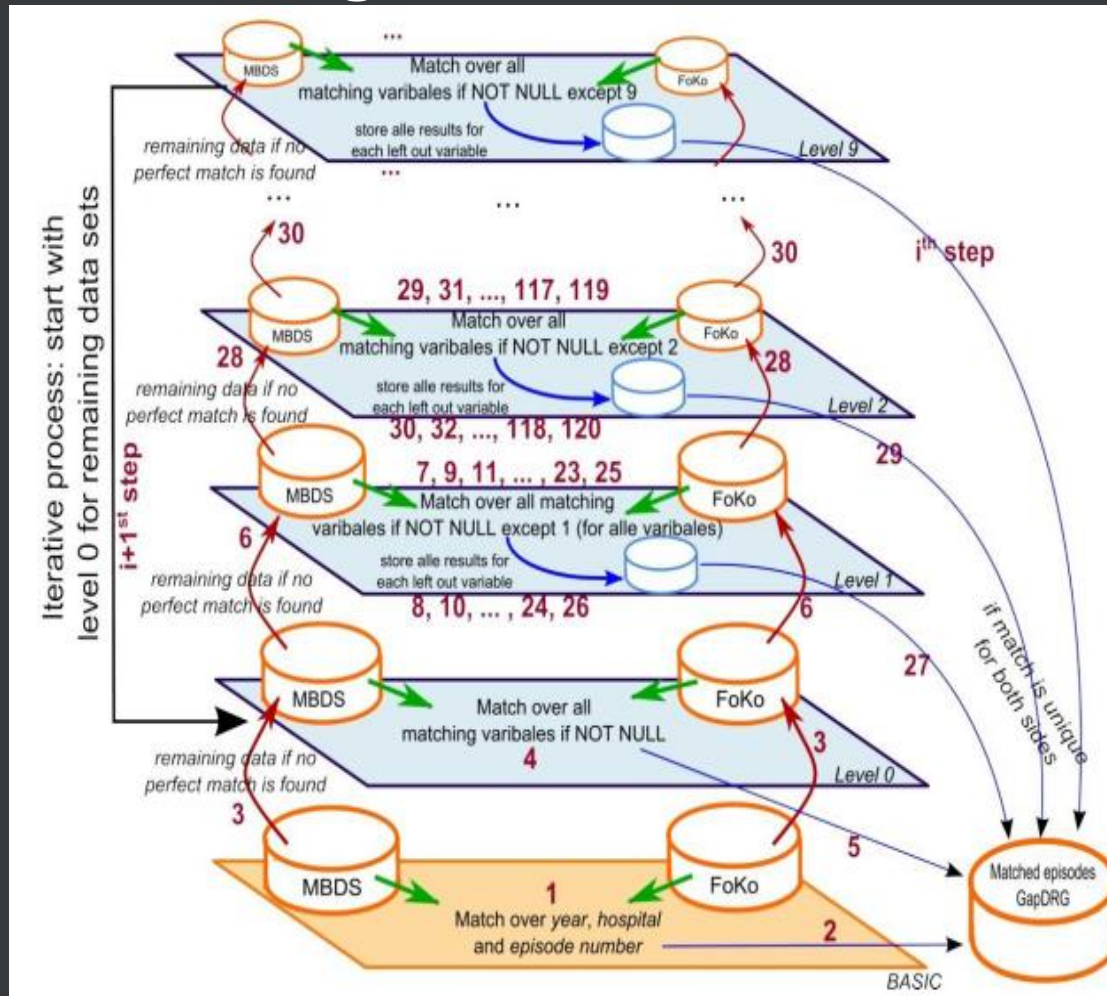
Why?

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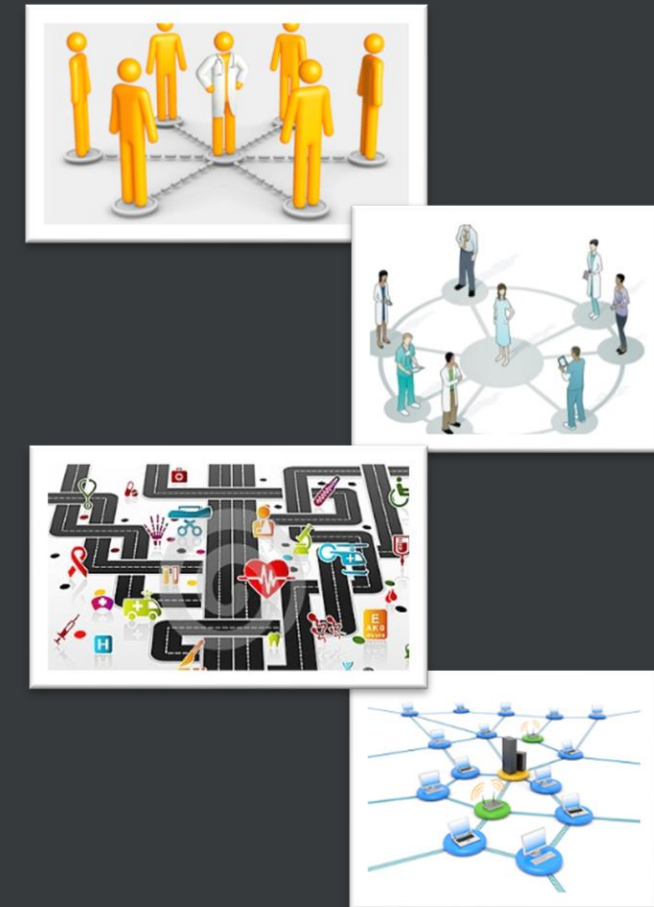
How?

- > 4th Paradigm
- > Modelling Dynamics
- > Data vs. Models
- > Processes

Big Data Sets



Structures & Knowledge



Modelling and Simulation Processes

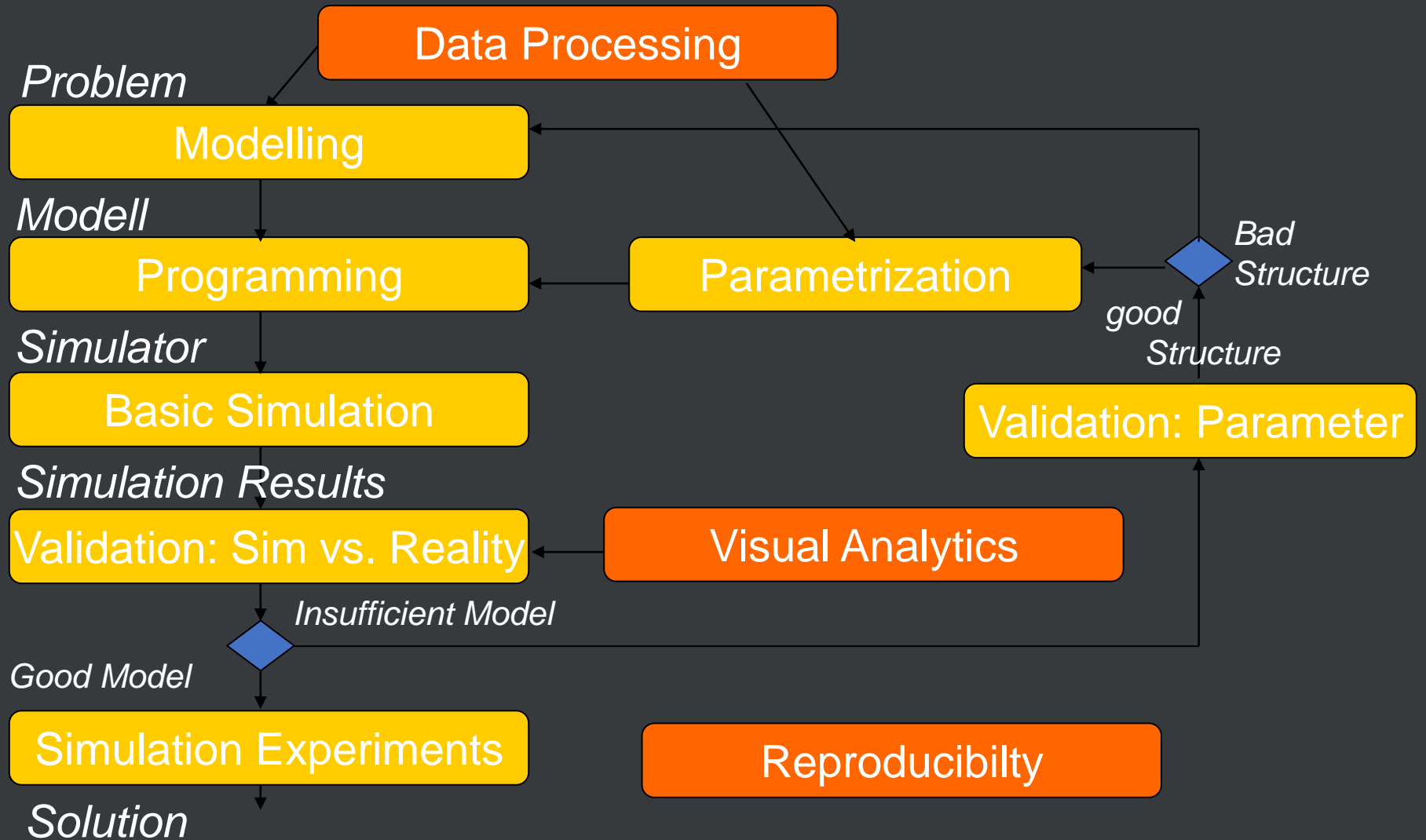
10 Concepts to Integrate

Why?

- > Dynamics & Complexity
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How?

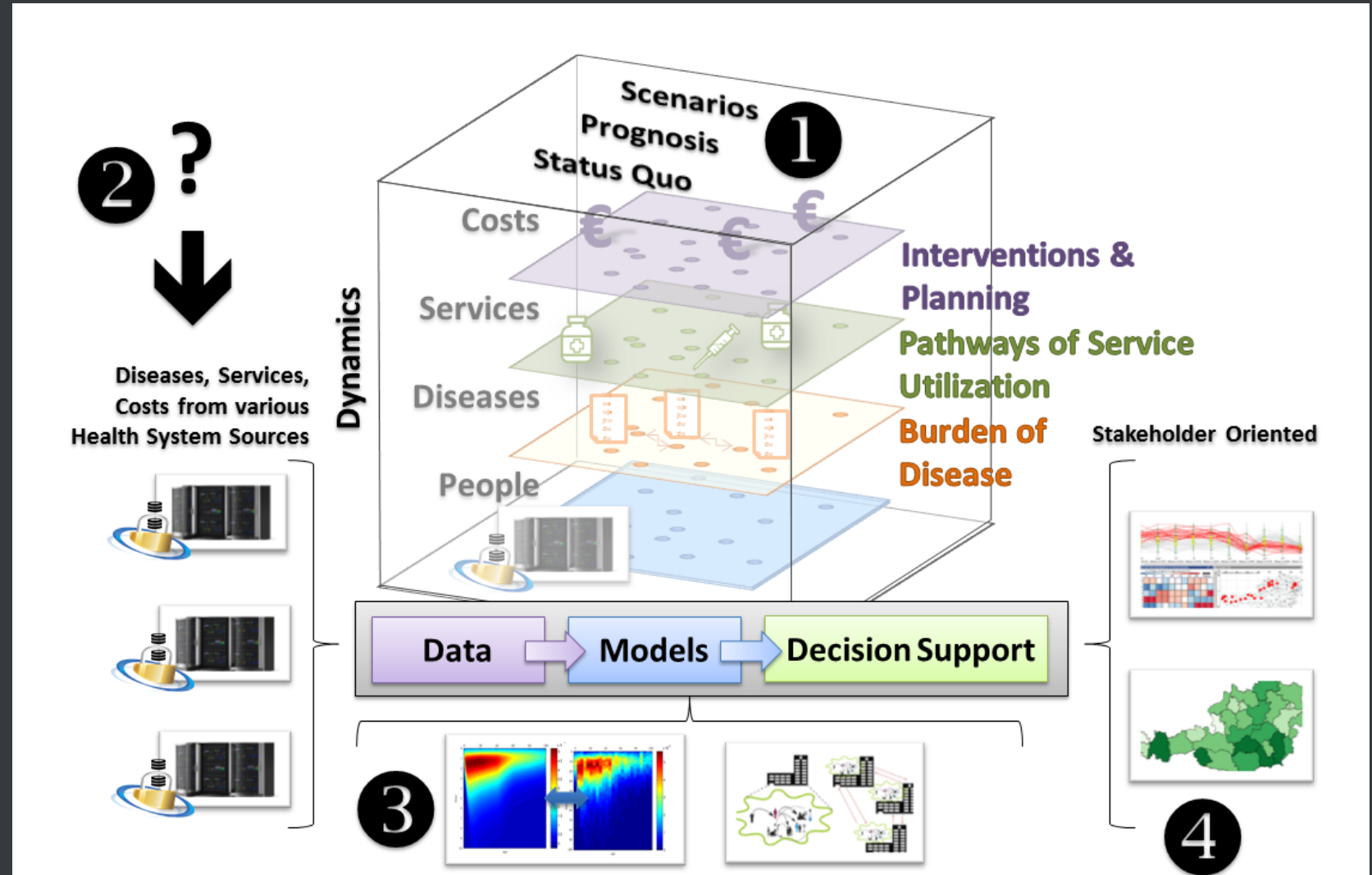
- > 4th Paradigm
- > Modelling Dynamics
- > Data vs. Models
- > Processes



Example Health System: DEXHELPP

DEXHELPP

Decision Support for Health Policy and Planning: Methods, Models and Technologies based on existing health care data

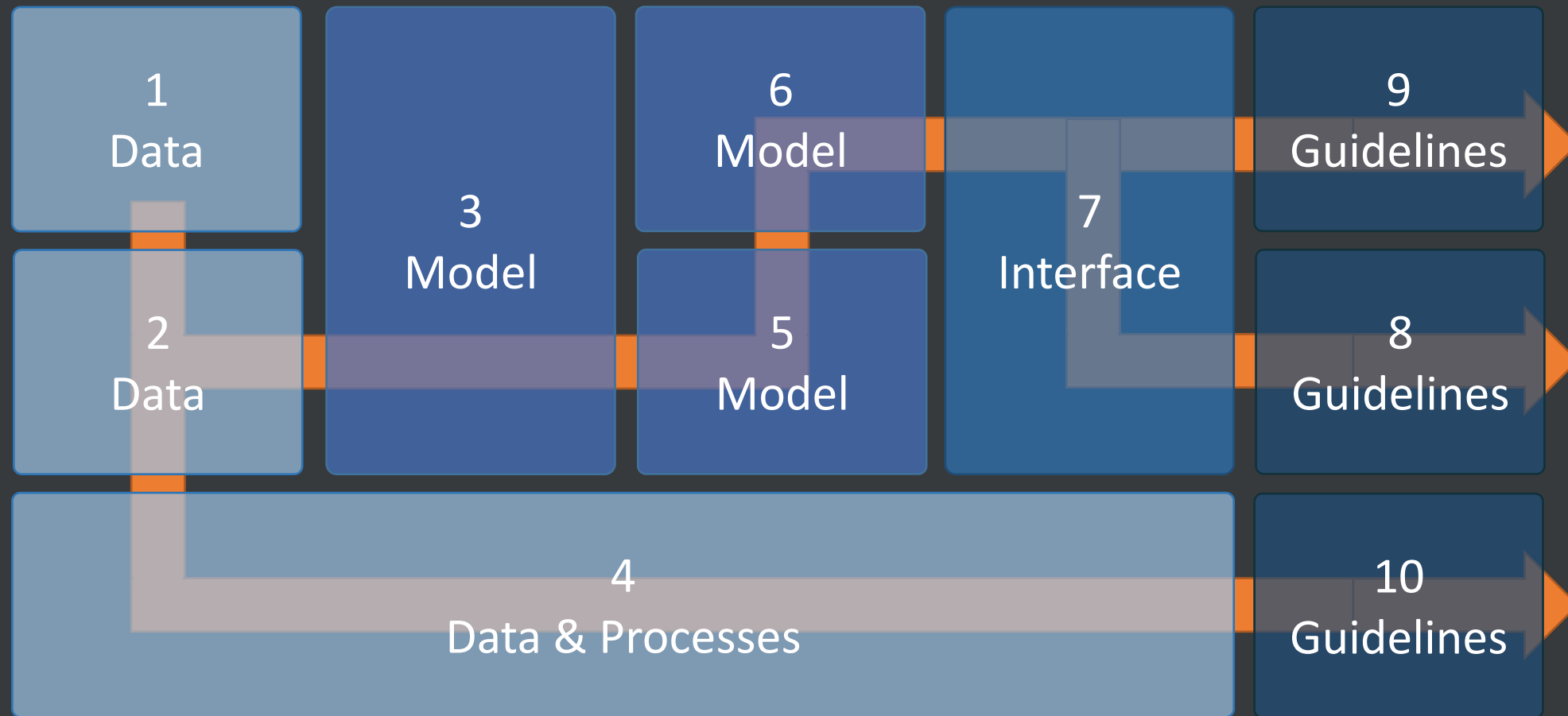


SV Hauptverband der österreichischen Sozialversicherungsträger
Gesundheit Österreich GmbH OBIG BIOG FGO
TU WIEN TECHNISCHE UNIVERSITÄT WIEN Vienna University of Technology
vrvis
dwh simulation services
IMEHPS Improving mental health pathways
SYNTHESIS Forschung
UMIT the health & life sciences university
secure sba-research.org
wirtschftsagentur wien Ein Fonds der Stadt Wien
bm vti Bundesministerium für Verkehr, Innovation und Technologie
bmwfw Bundesministerium für Wissenschaft, Forschung und Wirtschaft
COMET
FFG Competence Centers for Excellent Technologies

10 Concepts to Integrate

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results
7. Make it Understandable
8. Open and Independent Solutions
9. Priority for Data Security and Stake Holder Interests
10. Broad Applications (Health System, Energy, Industry, Energy, Mobility, Infrastructure)



Concept 1

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data

Problems of Collected Data

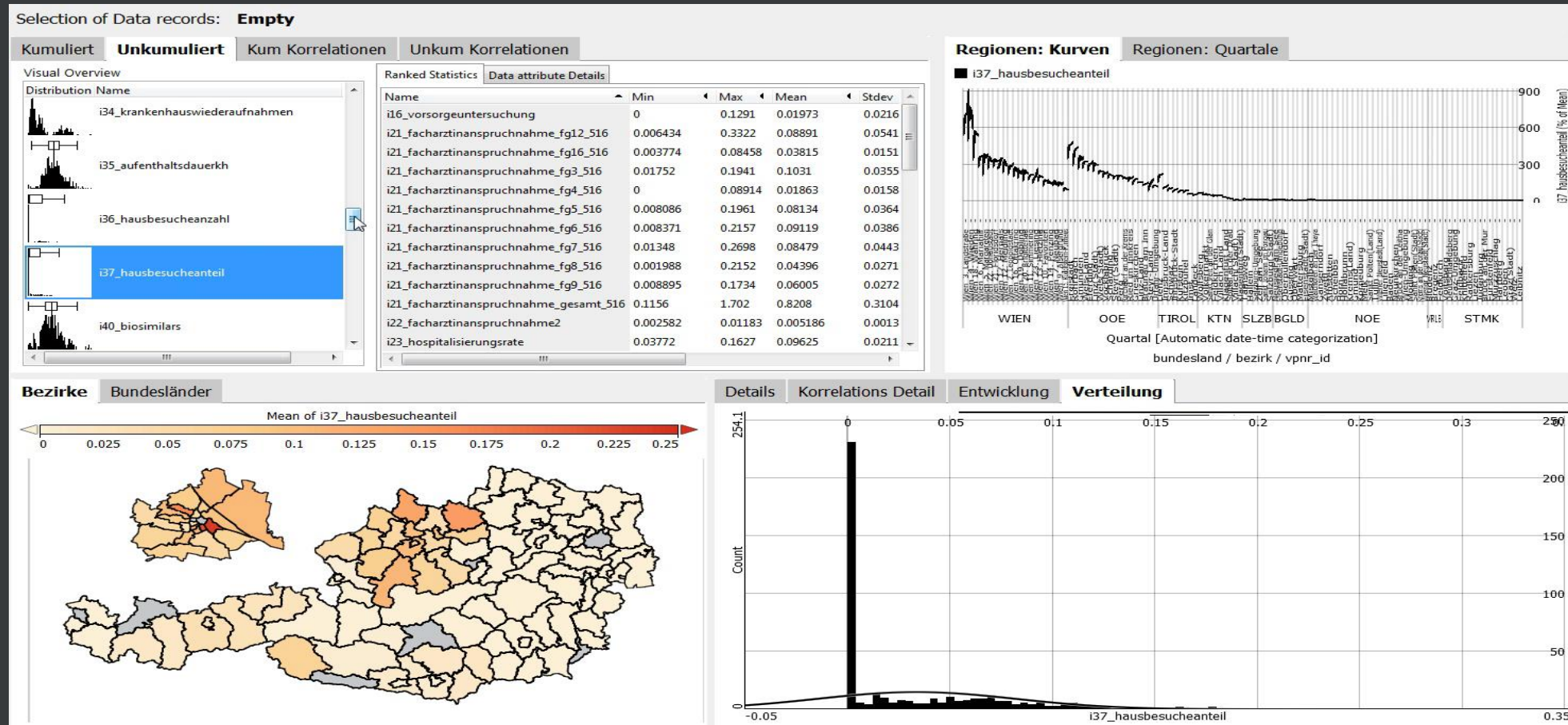
- 1
 - **Biased Collection of Data e.g. Sensor Data or Reimbursement Data is available**
 - **Pre-processing at Various Stakeholders**
 - **Privacy Demands**

METHOD: Explorative Visual Computing – Visual Analytics and Statistics

Interactive Dashboards

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data



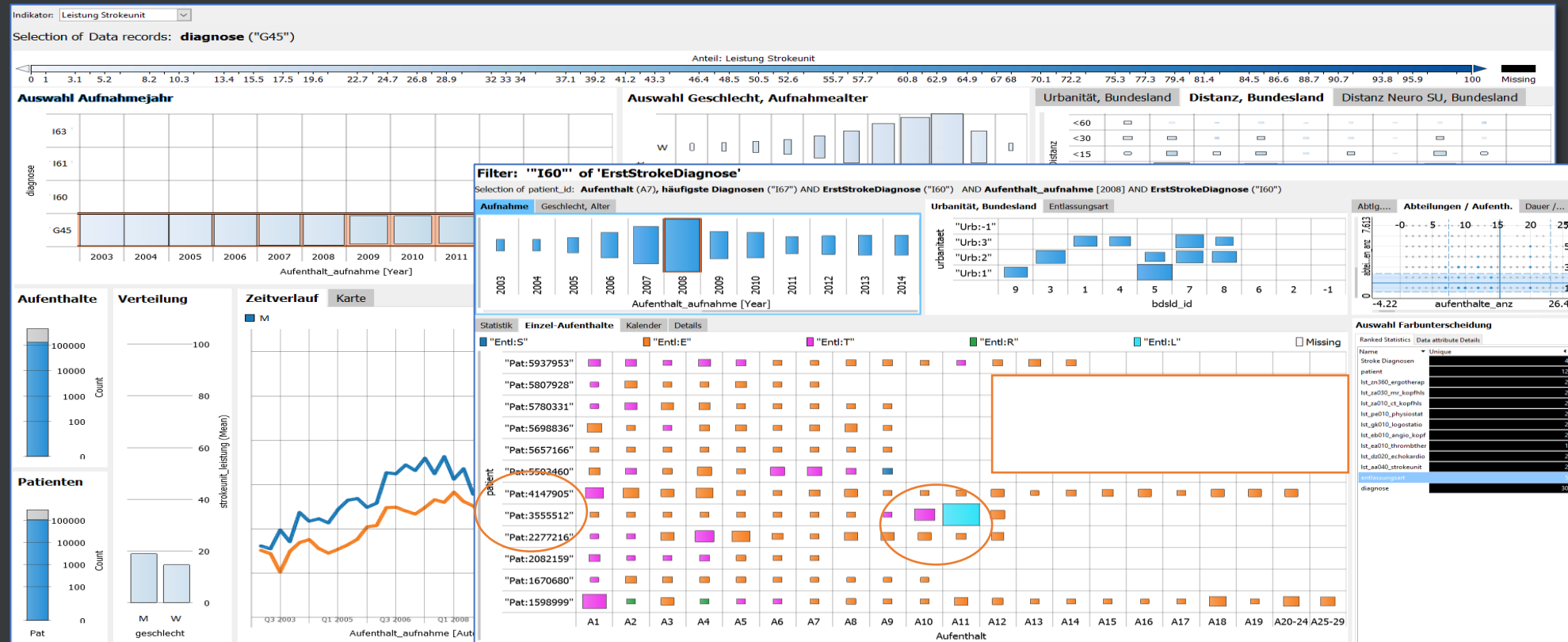
Data on Stroke Treatment

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data

Applied to: Data and Trend analysis of stroke treatment

- Percentage of patients being treated in stroke units
- Inspection of patient histories for outlier detection and hypothesis generation



Concept 2

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data

Integration of Different Data Sets

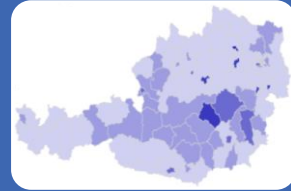
- ¹ Data Pre-processing at Various Stakeholders
- Unstructured and Different Structures
- ² Data Privacy Demands

METHOD: Data Processes (Integration & Linkage) & Modelling Tools (Parametrization & Calibration)

Data Levels

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Austrian Health System Data

DEXHELPP



Provincial Data

DEXHELPP
2018 PLUS



Clinical Data



OMICS
(excluded at the moment)

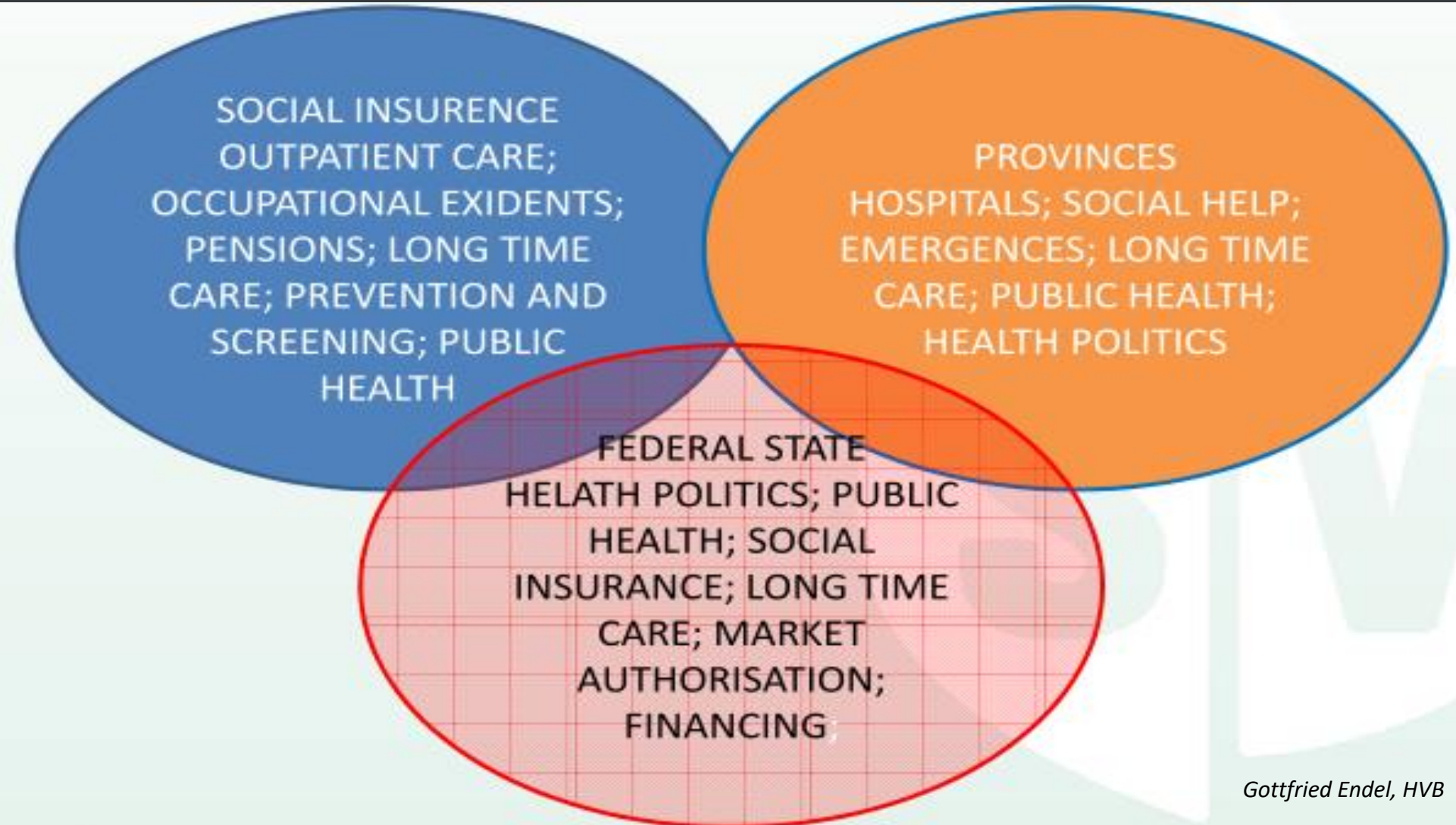
DEXHELPP
FUTURE

Health System Austria

3 Areas

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Data Social Insurancess

10 Concepts to Integrate

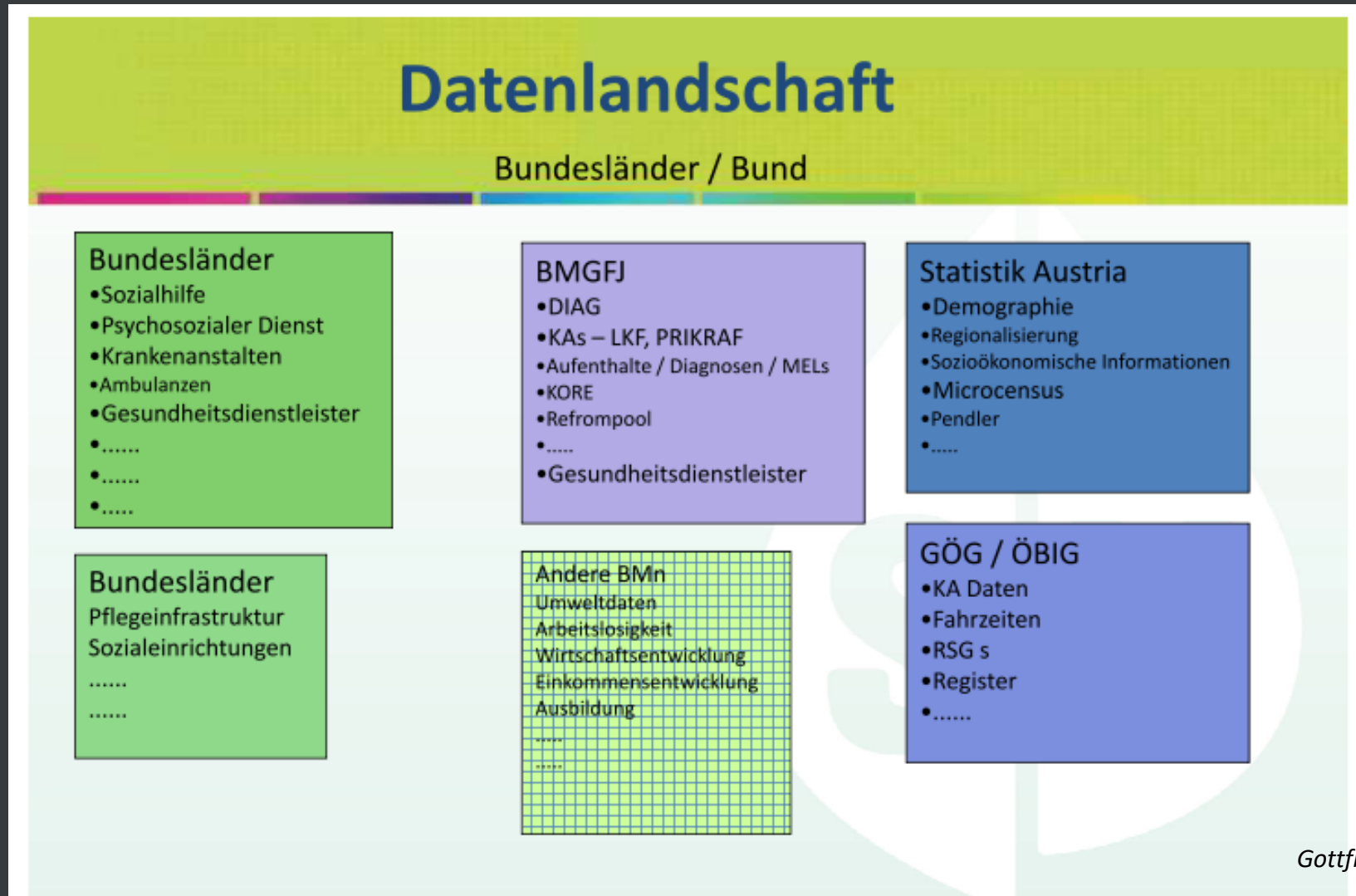
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Data Provinces & Austria

10 Concepts to Integrate

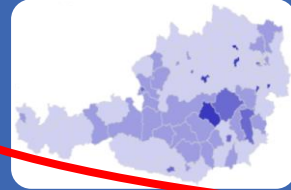
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Data Processing DEXHELPP until now....

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Austrian Health System Data



Provincial Data



Clinical Data

Hospitals/Inpatients
Physicians /Outpatients



OMICS
(excluded at the moment)

Σ



Σ



Diagnosis

Treatment

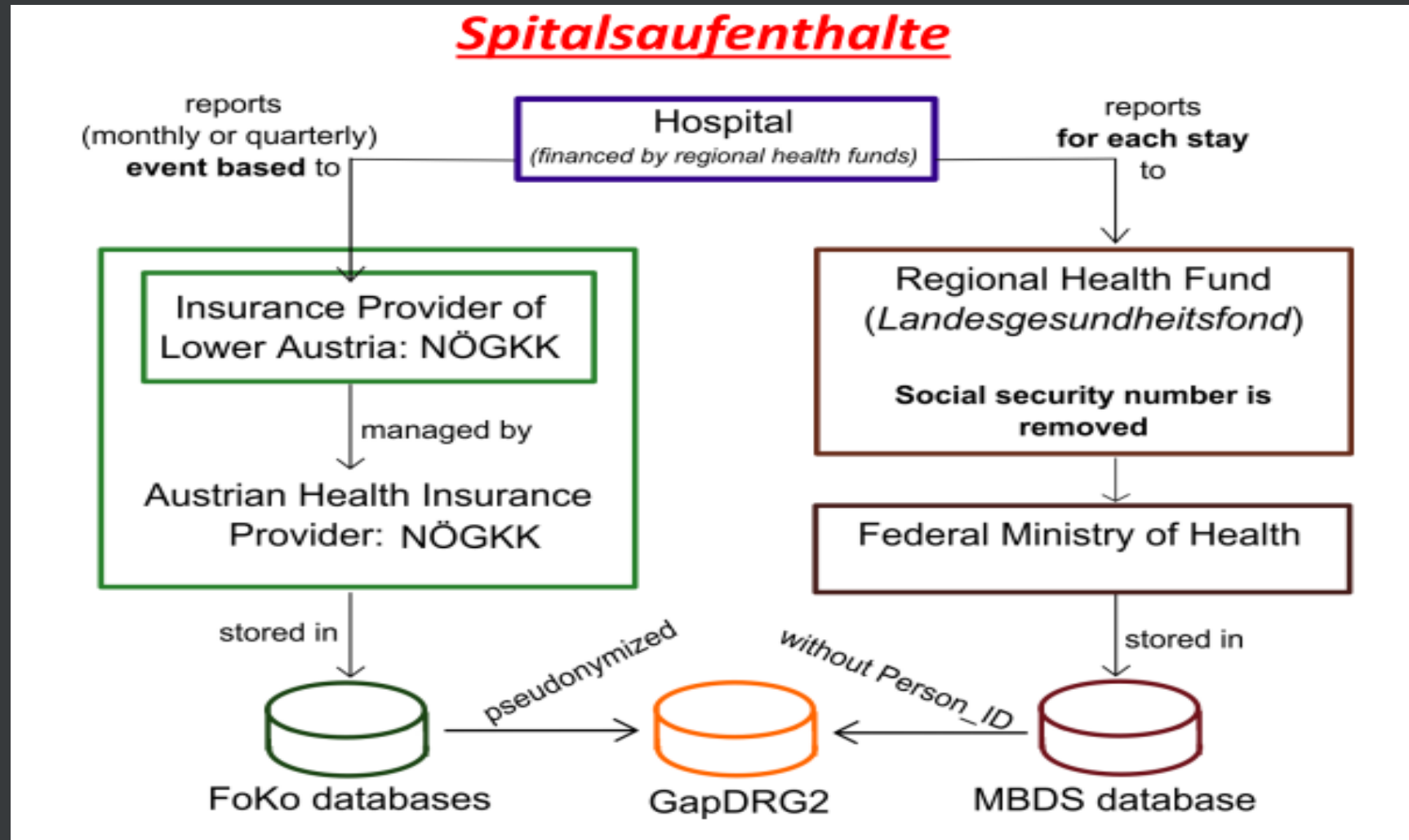
- Therapy
- Medication

Accounting

Example Record Linkage

10 Concepts to Integrate

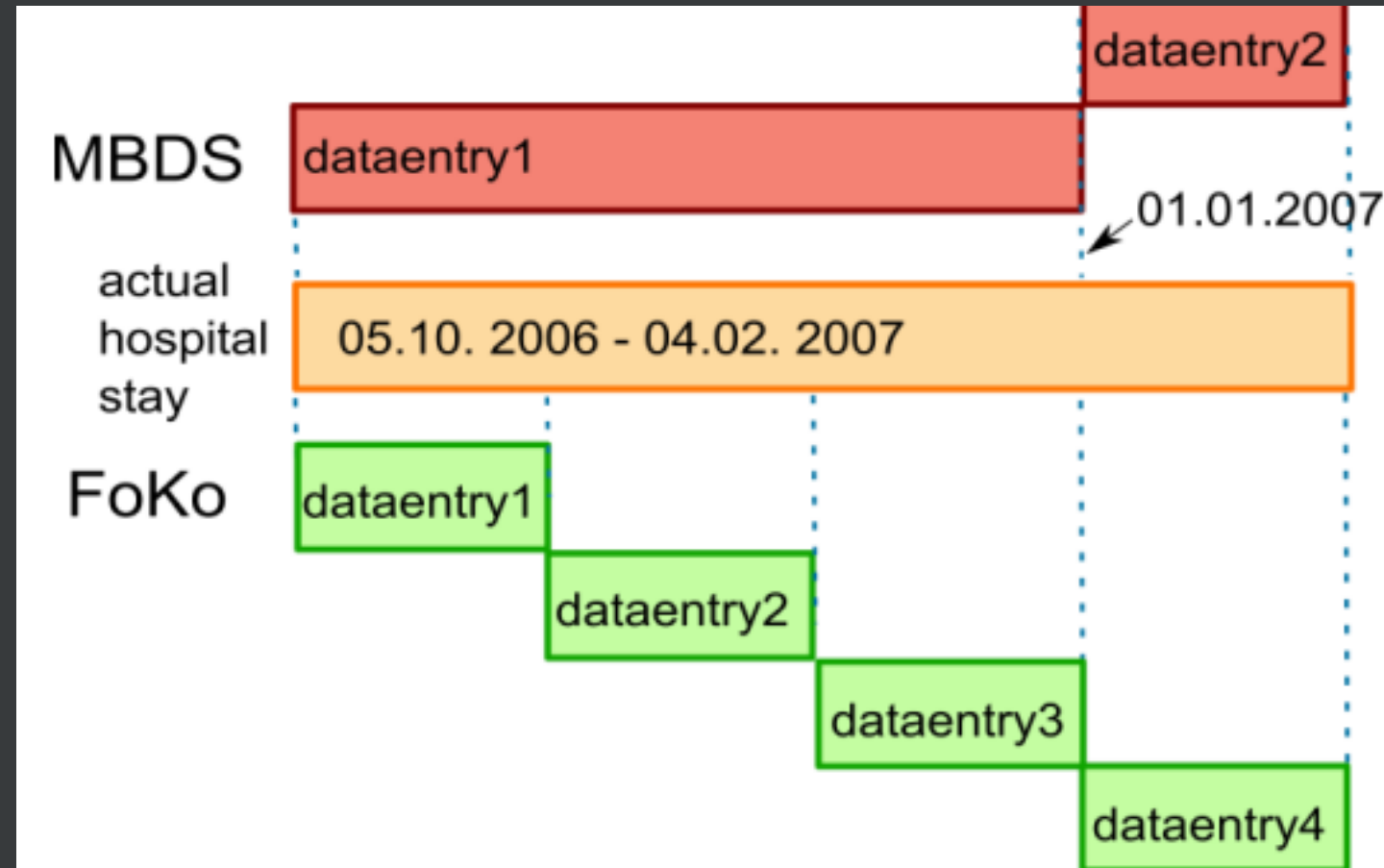
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Example Record Linkage

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data

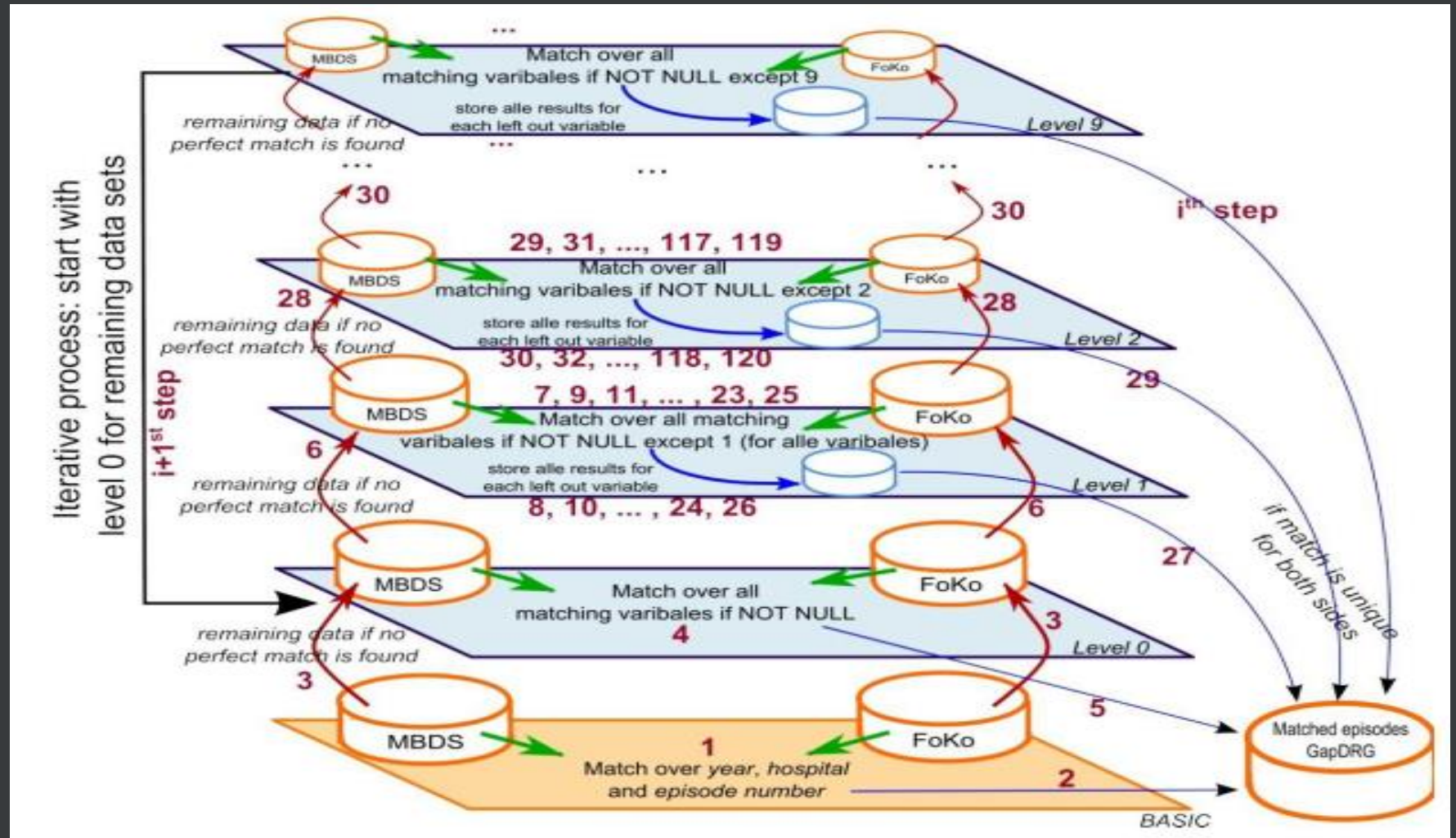


B. Glock, F. Endel et al: Challenges and Results with the Record Linkage of Austrian Health Insurance Data of Different Sources, Informatics for Health Conference 2017 (24 – 26. April, Manchester, UK)

Example Record Linkage

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



Concept 3

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions

Modular Concepts for Models

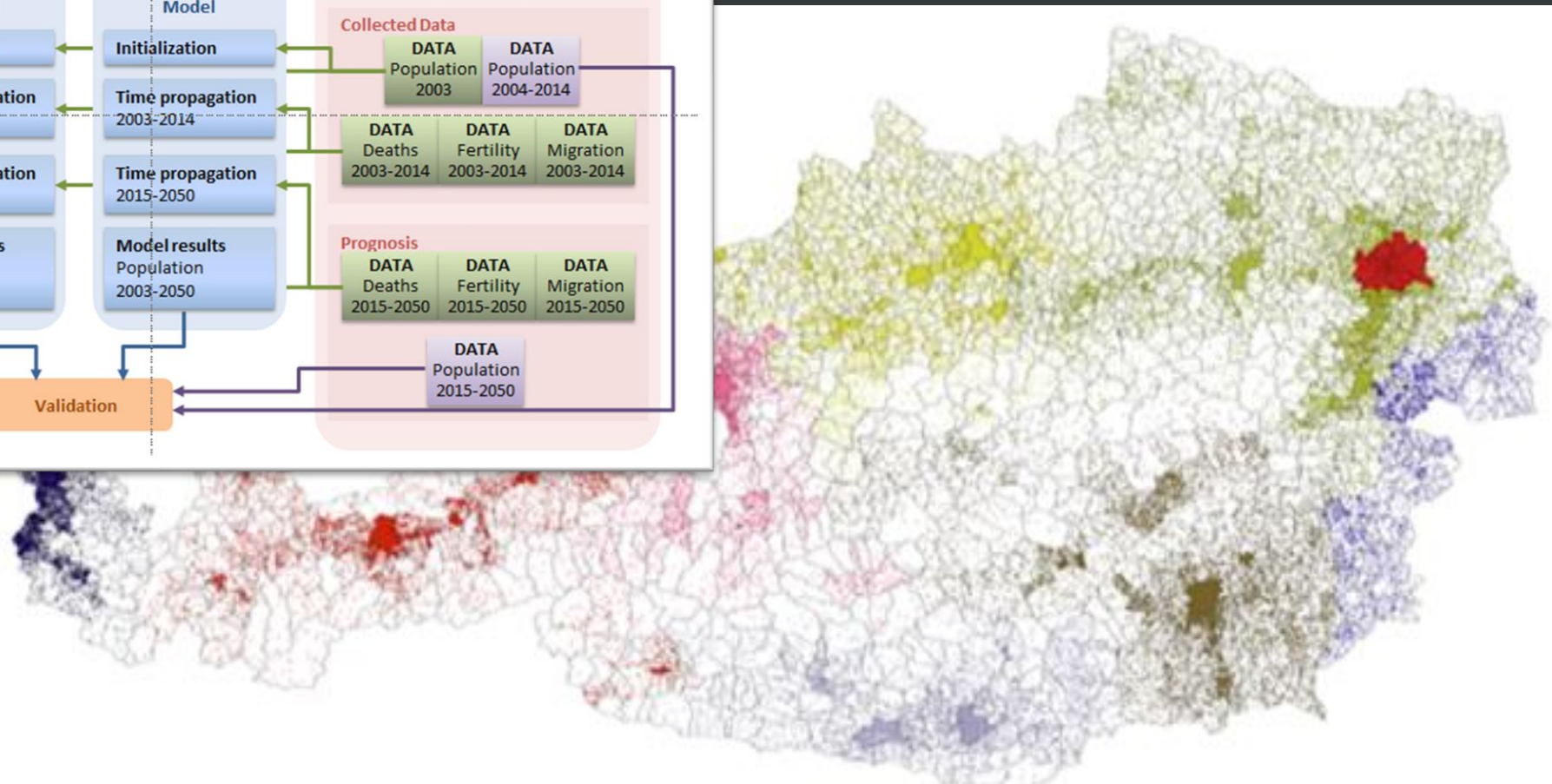
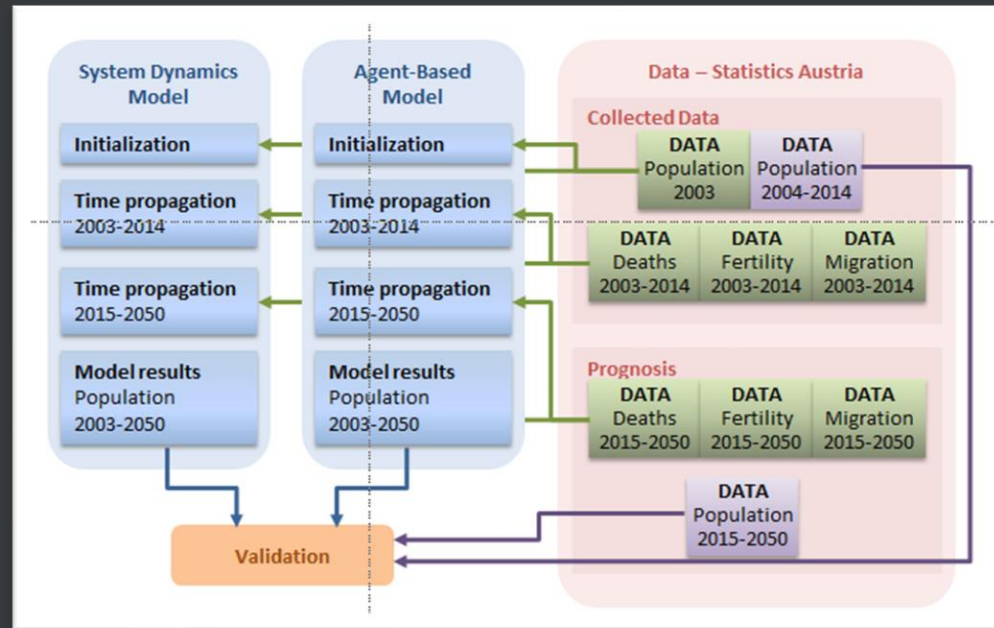
- We can find wrong Data
- We can change wrong Data ...
- ...even over time and when integrated
- We need transparent, “simple” Models

**METHOD: Modular Models,
Coupling of Models**

Virtual Population

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions



Social Network Layers

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions

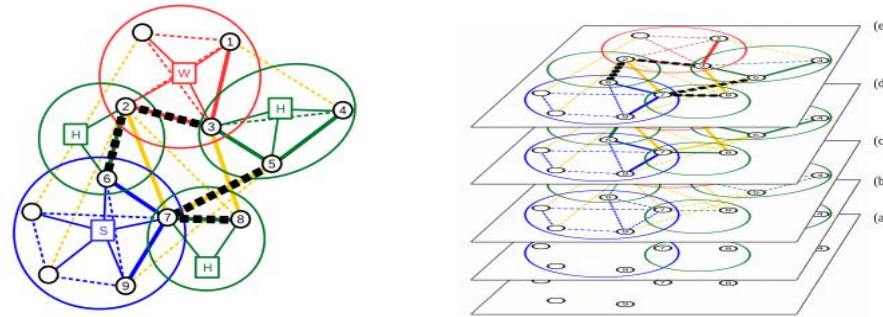


Figure 1: Visualization of the basic network layout. Households (H), workplaces (W) and school classes (S) and their memberships are depicted as green, red and blue blocks (b) of nodes (a). Social ties are indicated by dashed lines (c), derived contact patterns are shown as thick lines (d). Inter-block connections are distinguished by yellow lines. Instantiated temporal contacts are visualized as an overlaid black dashed pattern (e).

Time Enhancement

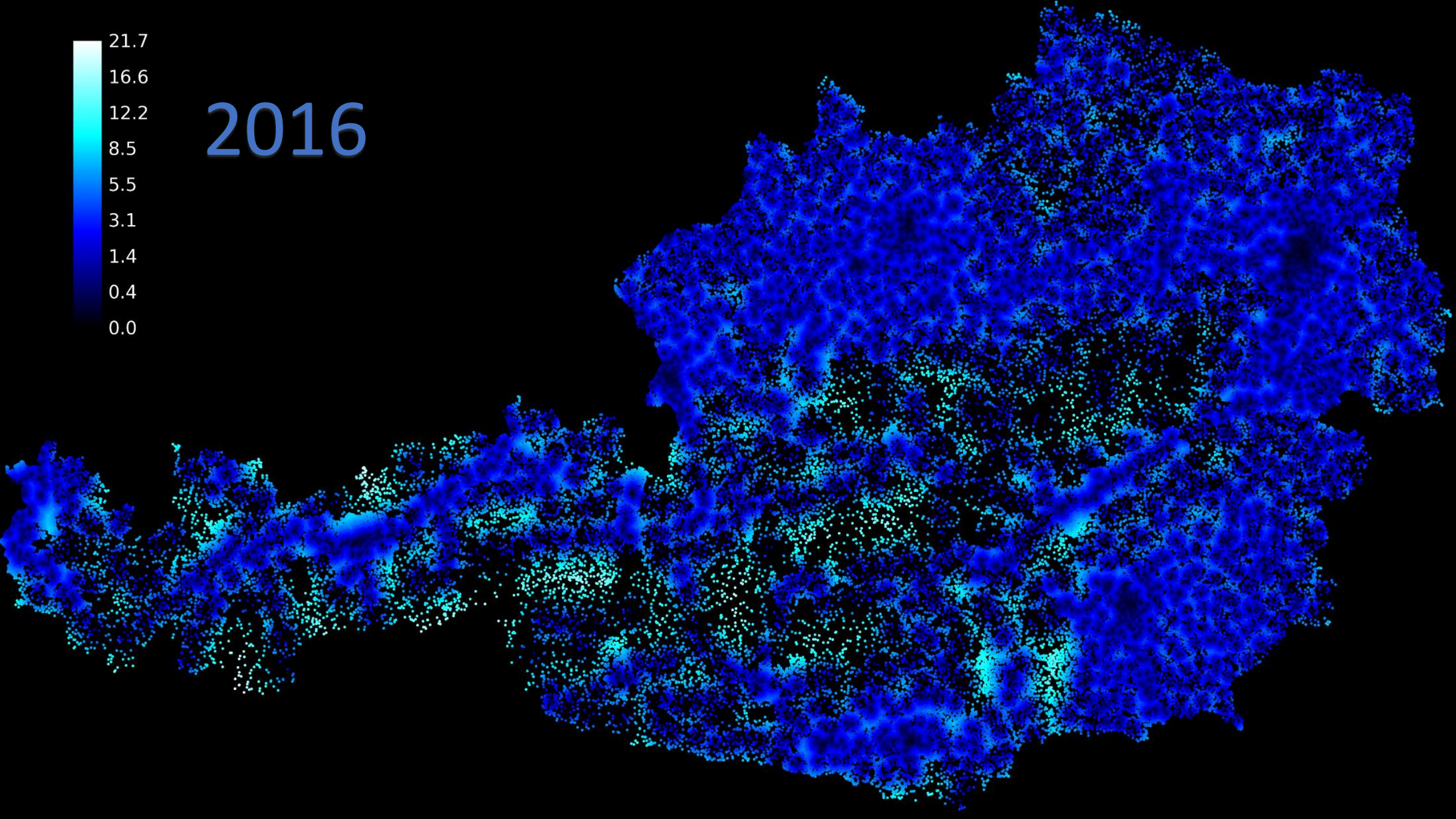
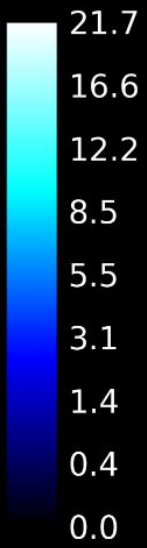
Internal Migration and Spatial Relationships

Interaction & Networks

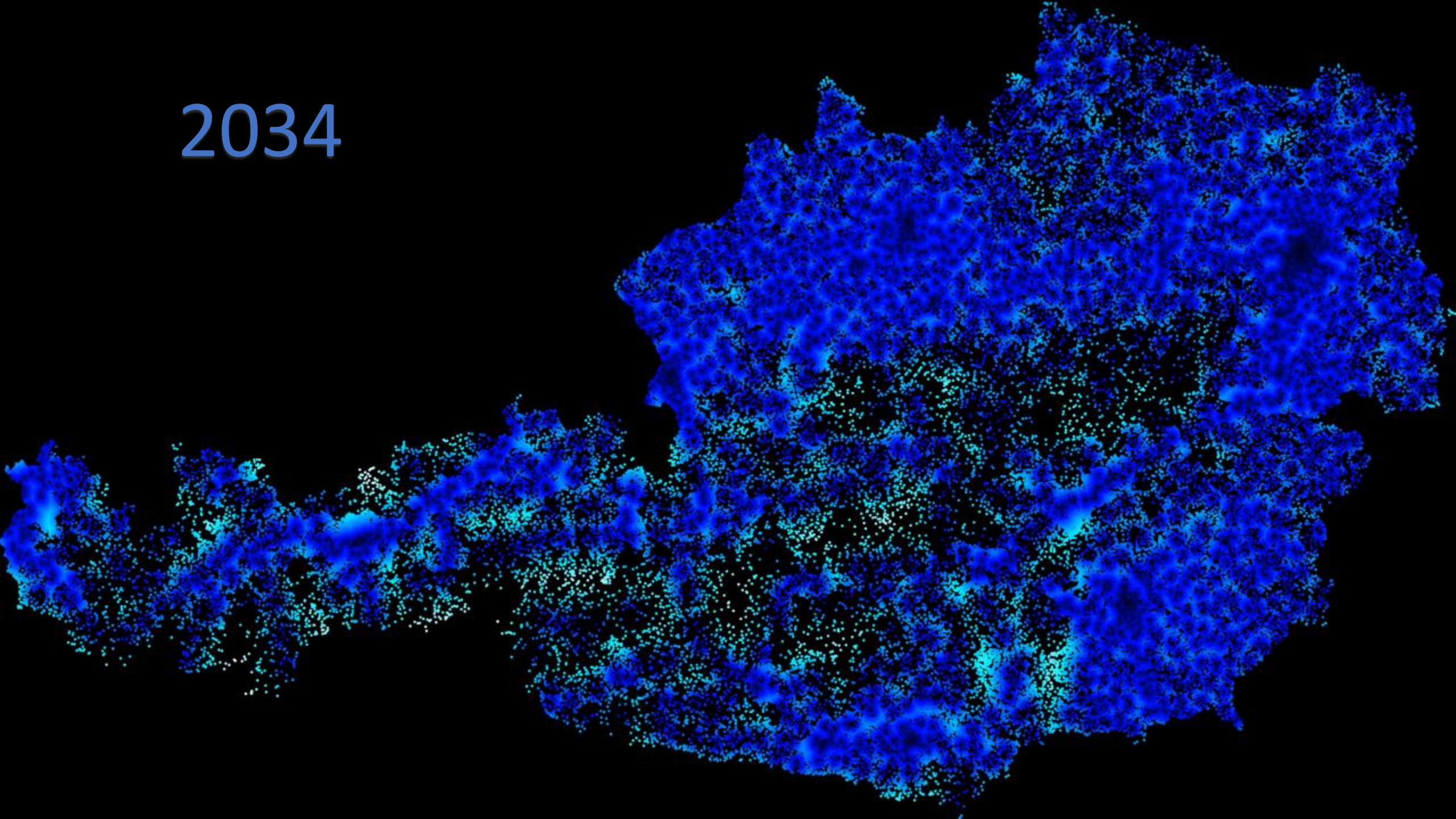
Runtime, Parallelisation

Parameter-Calibration

2016



2034



Concept 4

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes

Reproducible Processes

- No “General Model” possible, because of...
- ...different time scales or characteristics.
- Based on Provenance and Modularity
- Managing Tools for Data & Models

METHOD: Validation & Data Citation

Best Practice CEPHOS

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes

Comparative Effectiveness Research on Psychiatric Hospitalisation

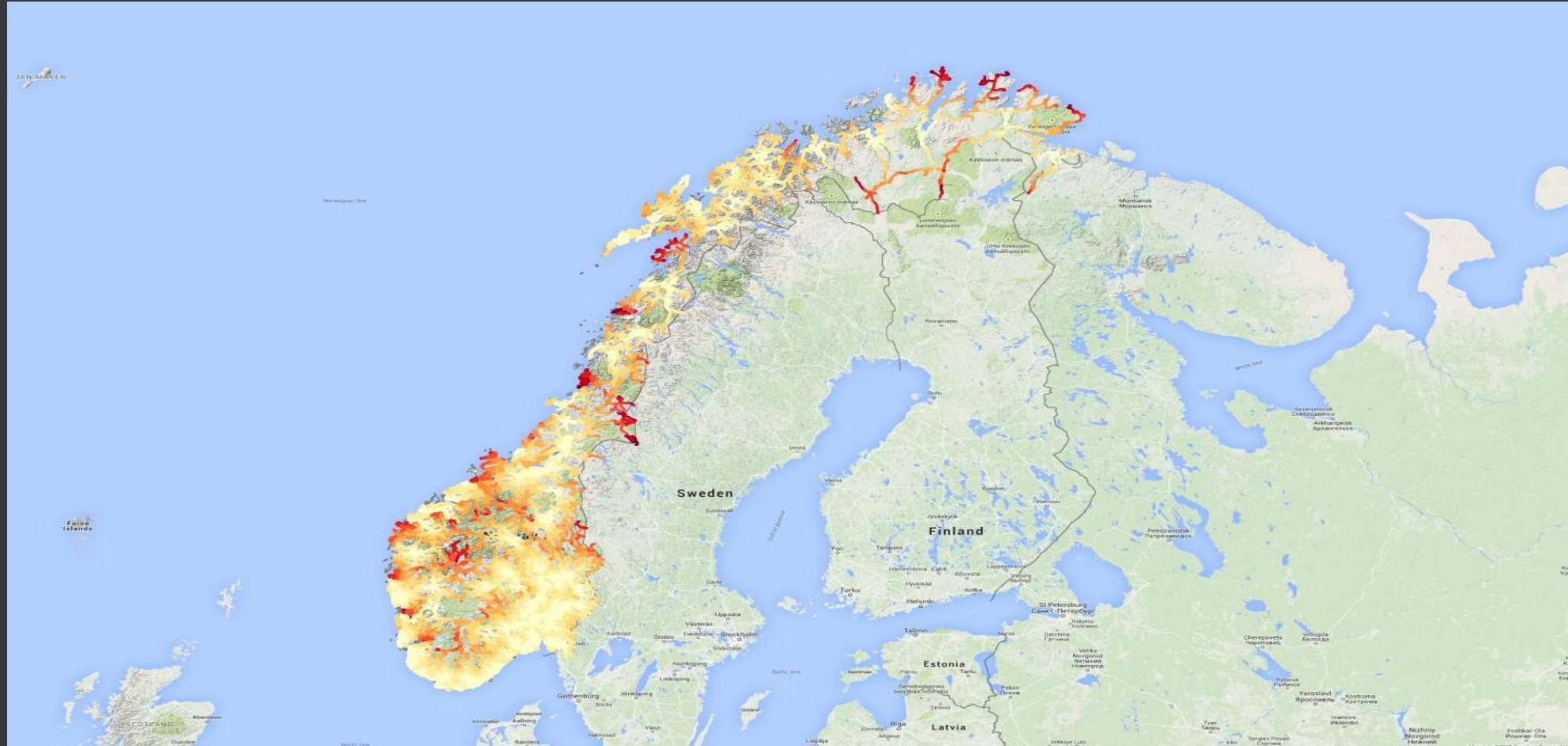
Description: Comparative Effectiveness Research on Psychiatric Hospitalisation by Record Linkage of Large Administrative Data Sets

CEPHOS-LINK is a comparative European register-based study, performed by record linkage technique. It will develop a methods toolkit for conducting record linkage studies in the mental health care field and produce recommendations, guidelines and a set of decision support tools for decision makers in the field of mental health system.

Best Practice CEPHOS

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes

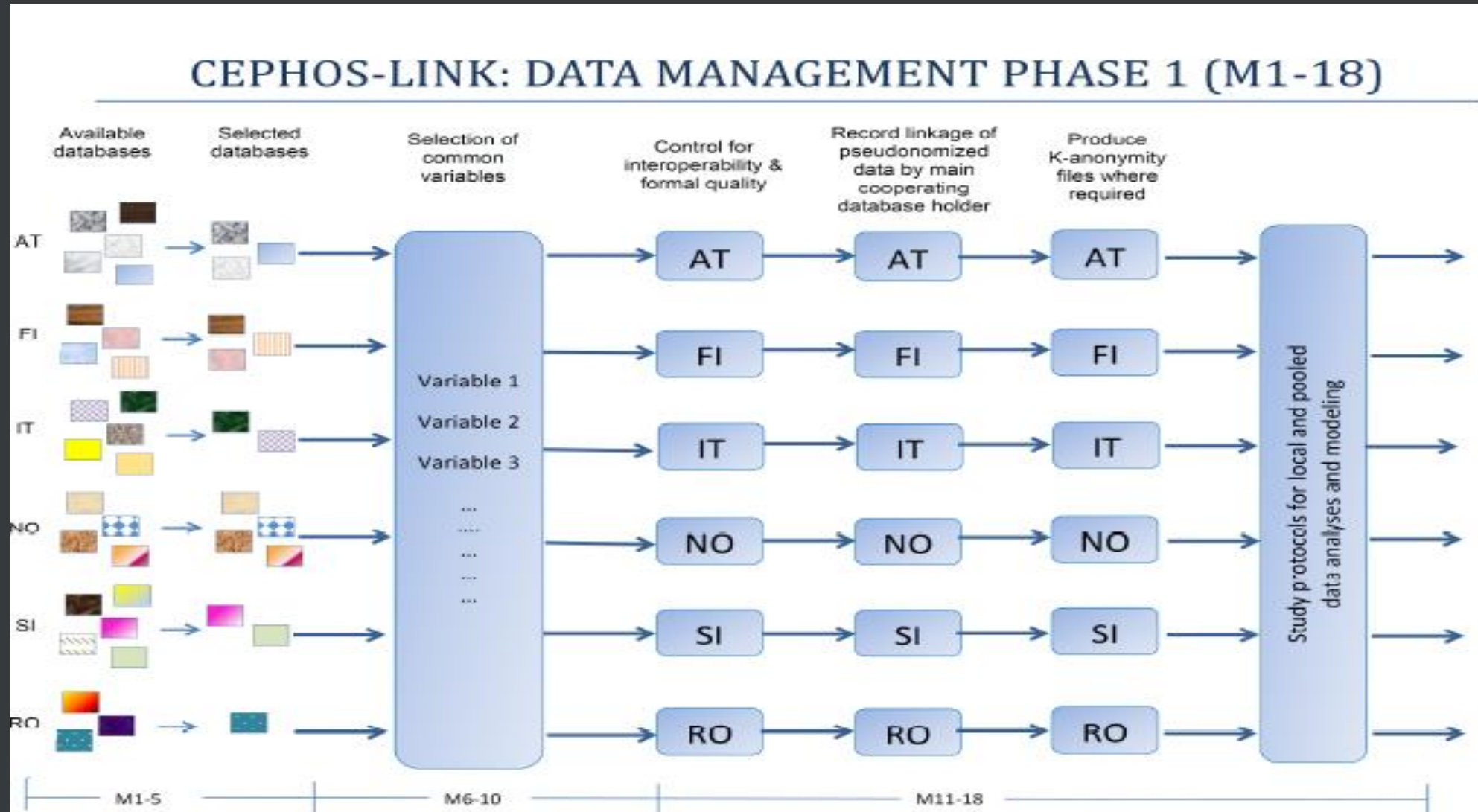


C. Urach, G. Zauner et al "Statistical methods and modelling techniques for analysing hospital readmission of discharged psychiatric patients: a systematic literature review"; BMC Psychiatry (2016)

Best Practice CEPHOS

10 Concepts to Integrate

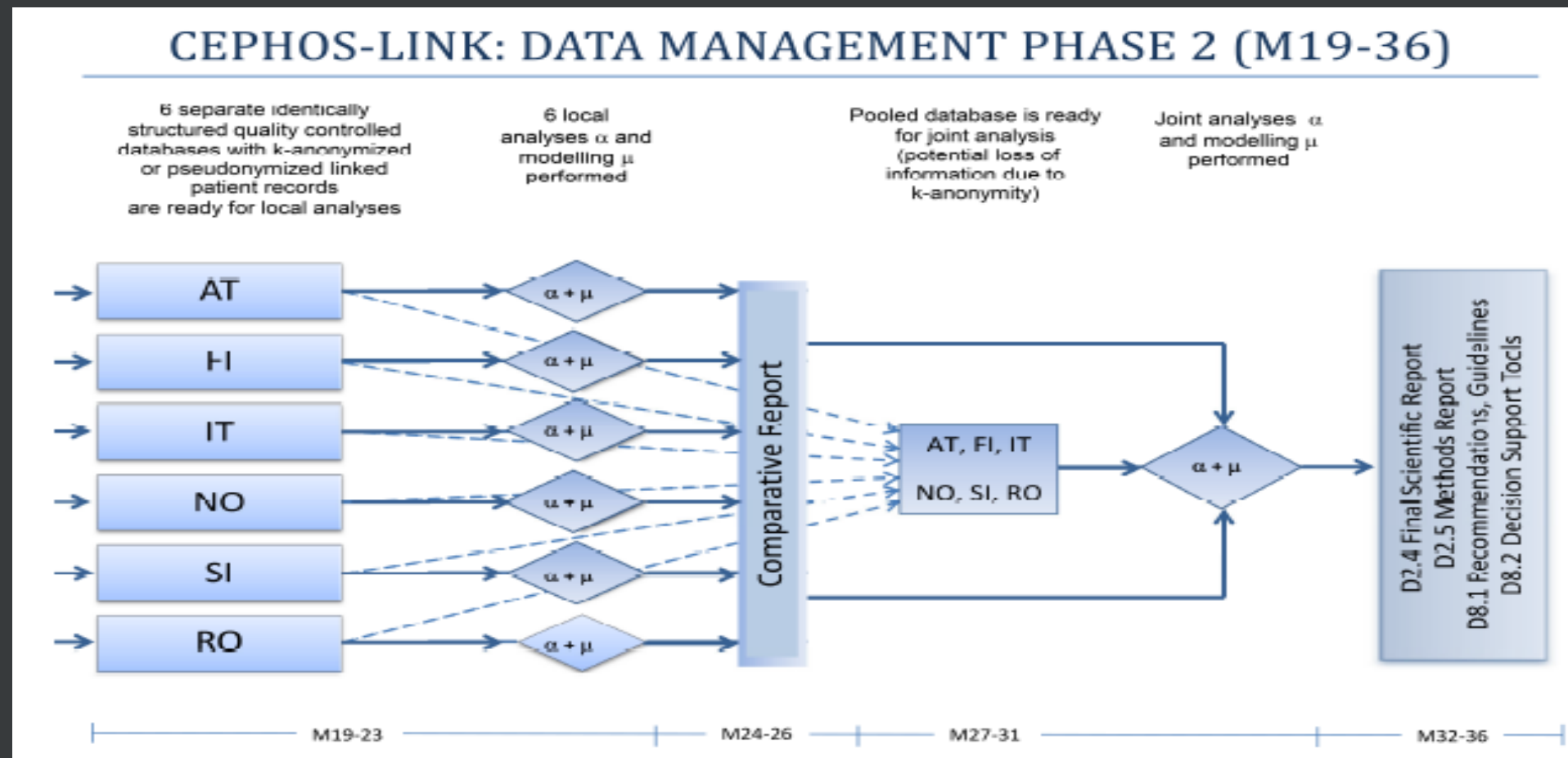
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes



Best Practice CEPHOS

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes



Concept 5

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)

Different Models for Different Questions

- Stable model, already checked
- Better Data doesn't improve
- Model changes don't improve
- Need to Integrate Complexity/Dynamics
- Different (comparable) methods needed

METHOD: Methods for Choosing Models

Example Comparison Vaccination

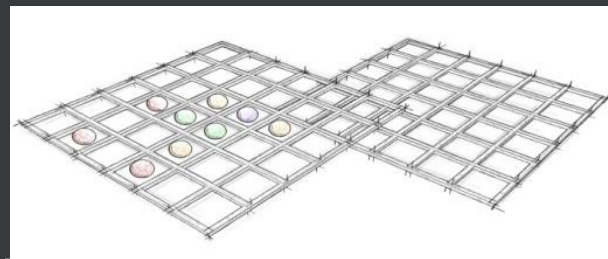
10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
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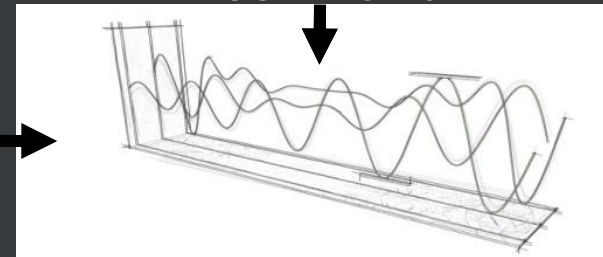
Modelling of infectious diseases: Pneumococcal modelling systems



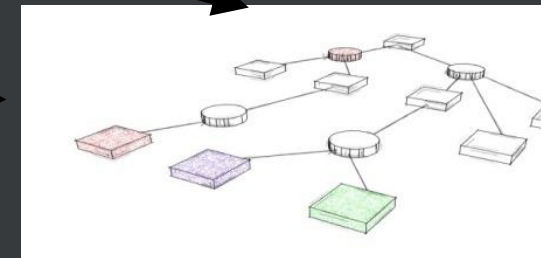
Real world



AB with const. population

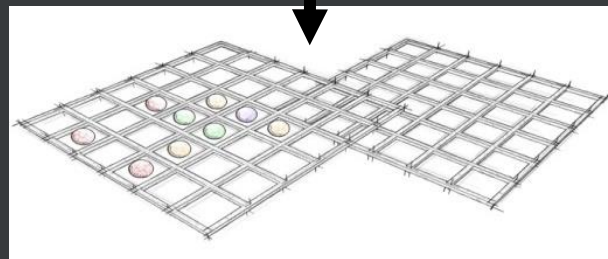


ODE with const. population

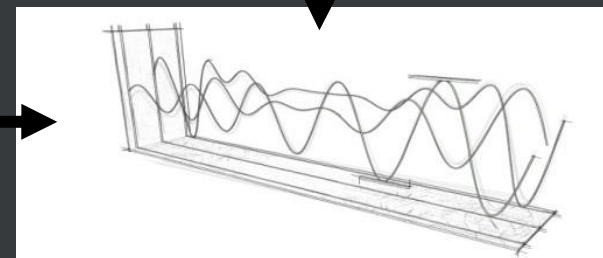


Markov Model

- Decision tree
- Statistics



AB with populationdynamics



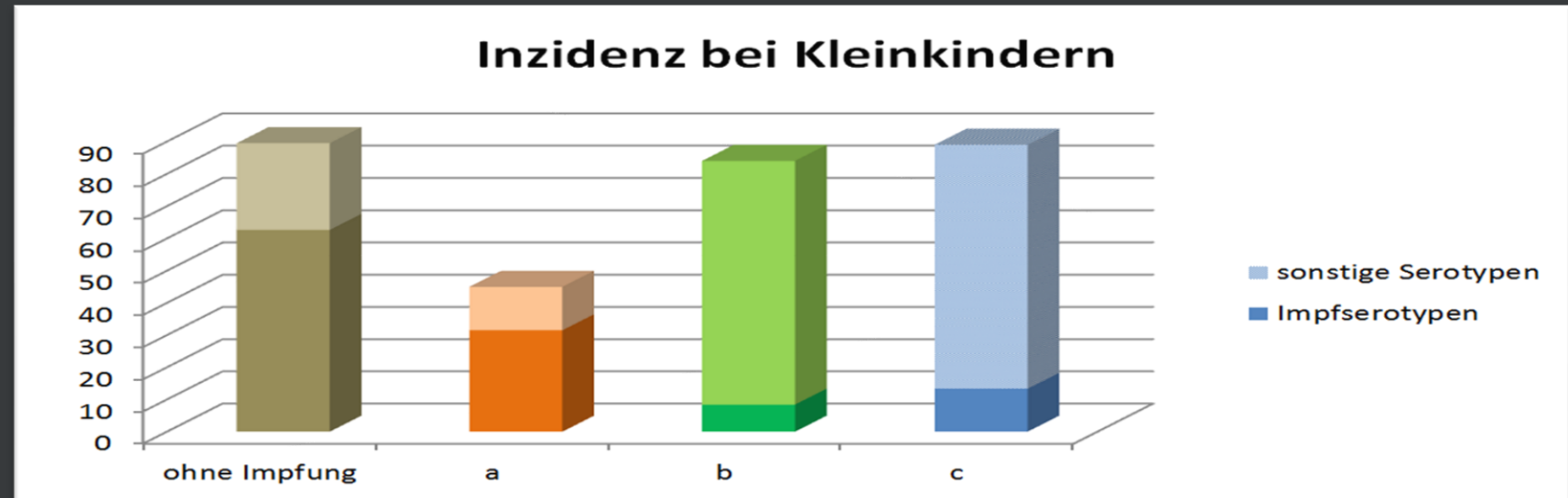
ODE with populationdynamics



Example Comparison Vaccination

10 Concepts to Integrate

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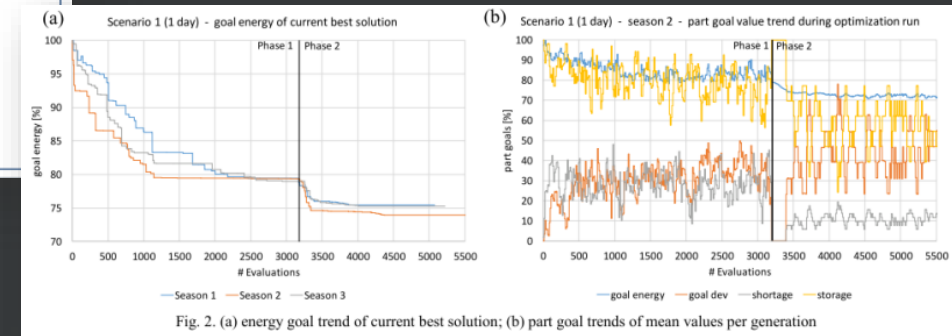
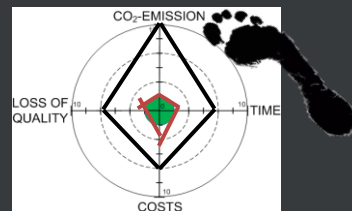
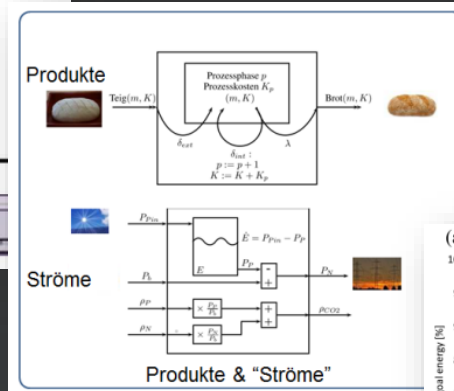
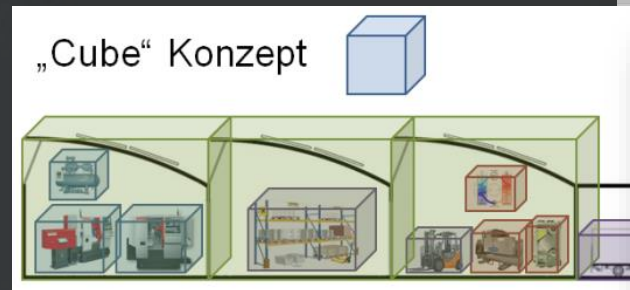
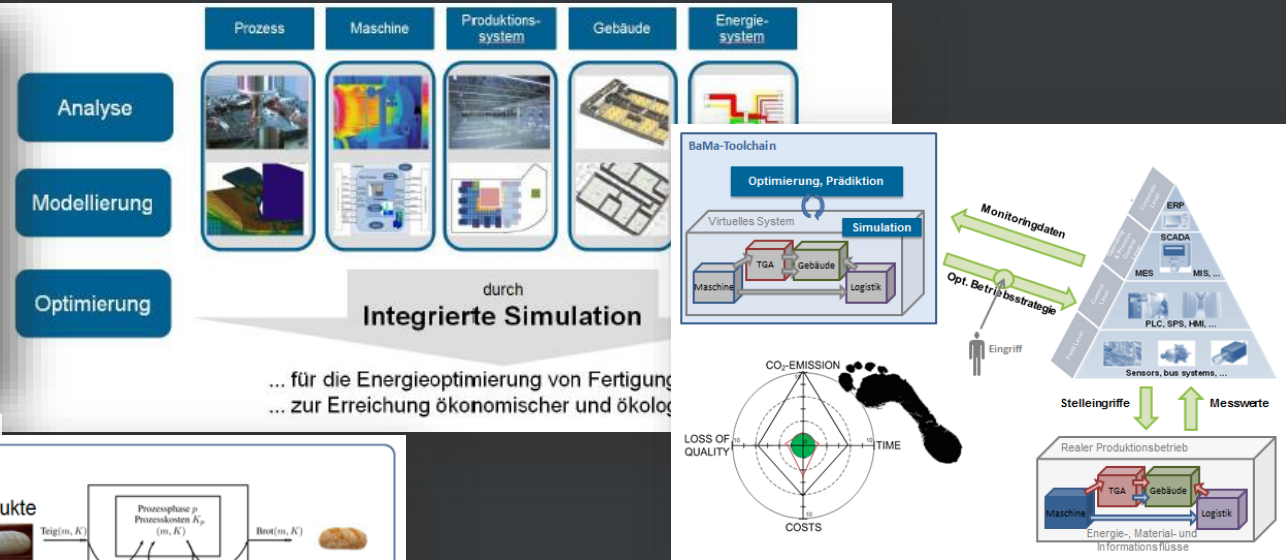
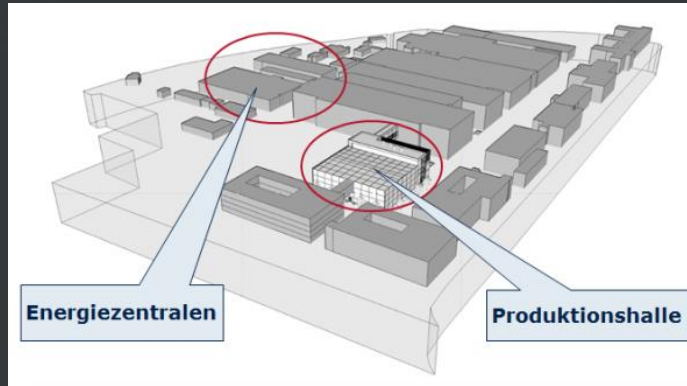


- a) **Gebräuchliche Prognosemodelle (Markov-Modelle, auf Österreich umgerechnet):**
E.D.G. McIntosh, P. Conway, J. Willingham, R. Hollingsworth, and A. Lloyd. The cost-burden of paediatric pneumococcal disease in the UK and the potential cost-effectiveness of prevention using 7-valent pneumococcal conjugate vaccine. *Vaccine*, 2003 Jun 2,21(19-20):2564-72
- b) **Dynamisches Pneumokokkenmodell, 2009 – Simulationsergebnisse:**
u.a.: C. Urach, "Modellierung und Simulation von Impfstrategien gegen Pneumokokkenerkrankungen: Markov- und Differentialgleichungsmodelle im Vergleich" (Diploma Thesis, Inst. f. Analysis und Scientific Computing, Vienna University of Technology, 2009).
- c) **Erhebung aus den USA, 2010 (auf Österreich umgerechnet):**
Hsu KK et al. Changing serotypes causing childhood invasive pneumococcal disease: Massachusetts, 2001–2007. *Pediatr Infect Dis J* 2010 Apr; 29:289

Example Combination Balanced Manufacturing

10 Concepts to Integrate

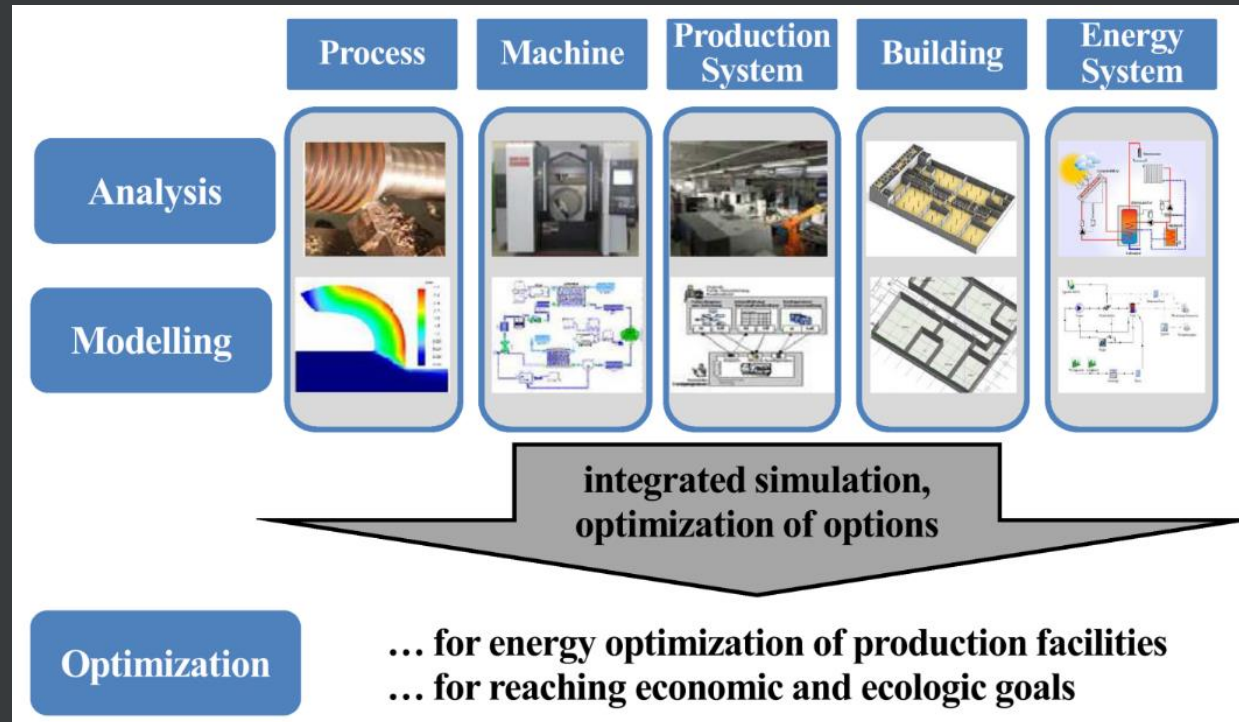
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)



Example Combination Balanced Manufacturing

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- **Coupled simulation** of the overall system enables holistic view of the energy distribution throughout the system.
- Find **optimization** approaches by simulating and comparing different scenarios.

Concept 6

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
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4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results

Comparability of Models and Results

- Qualitative Comparison
- Quantitative Comparison including Parameter Transformation
- Showing Limitations of Modelling Approaches and Implementation

METHOD: Comparative Modelling

Example: Calibration

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Calibration algorithms require thousands of simulation runs

Problematic



Each simulation run contains random elements and takes some time

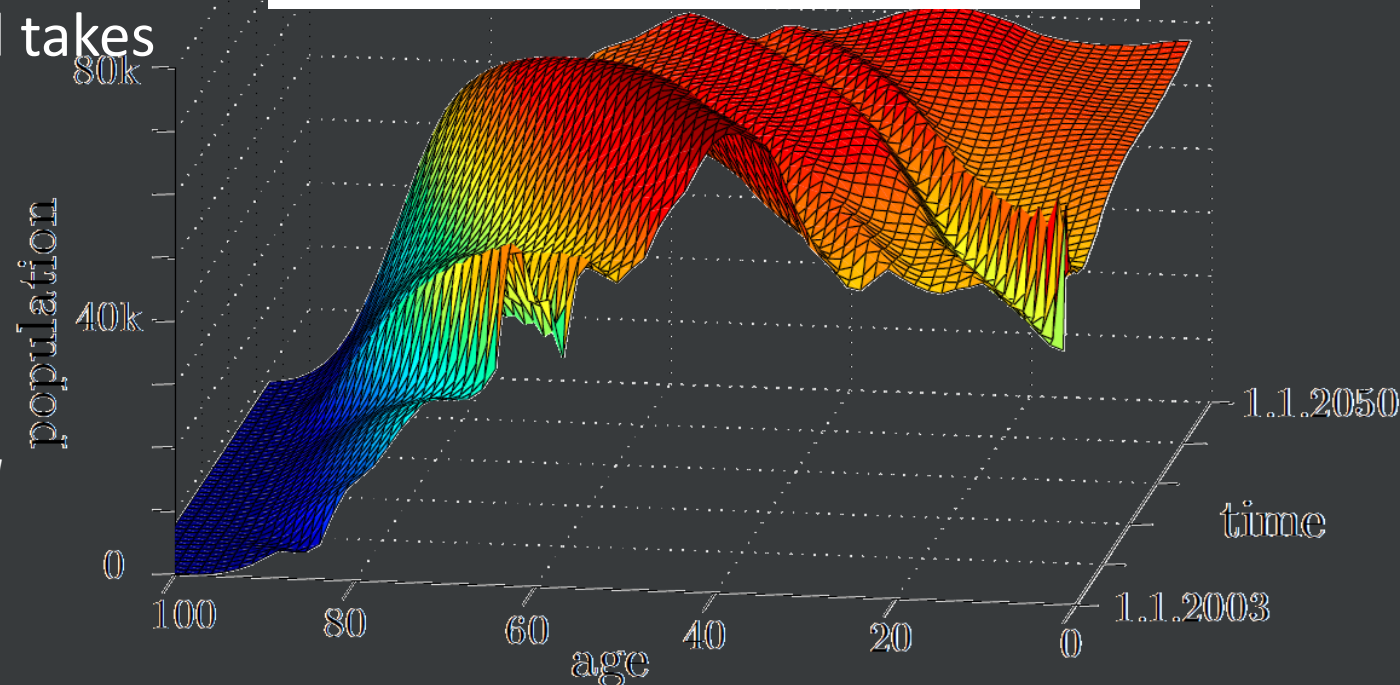
M. Bicher, N. Popper, „Mean-Field Approximation of a Microscopic Population Model for Austria “ published Eurosima 2016



Formal model analysis

$$\frac{\partial M}{\partial t} - \frac{\partial M}{\partial a} = \alpha_1 + \gamma_1 \mathbb{1}_{[0,dt)}(a) \Psi(a, t) - M \delta_1$$

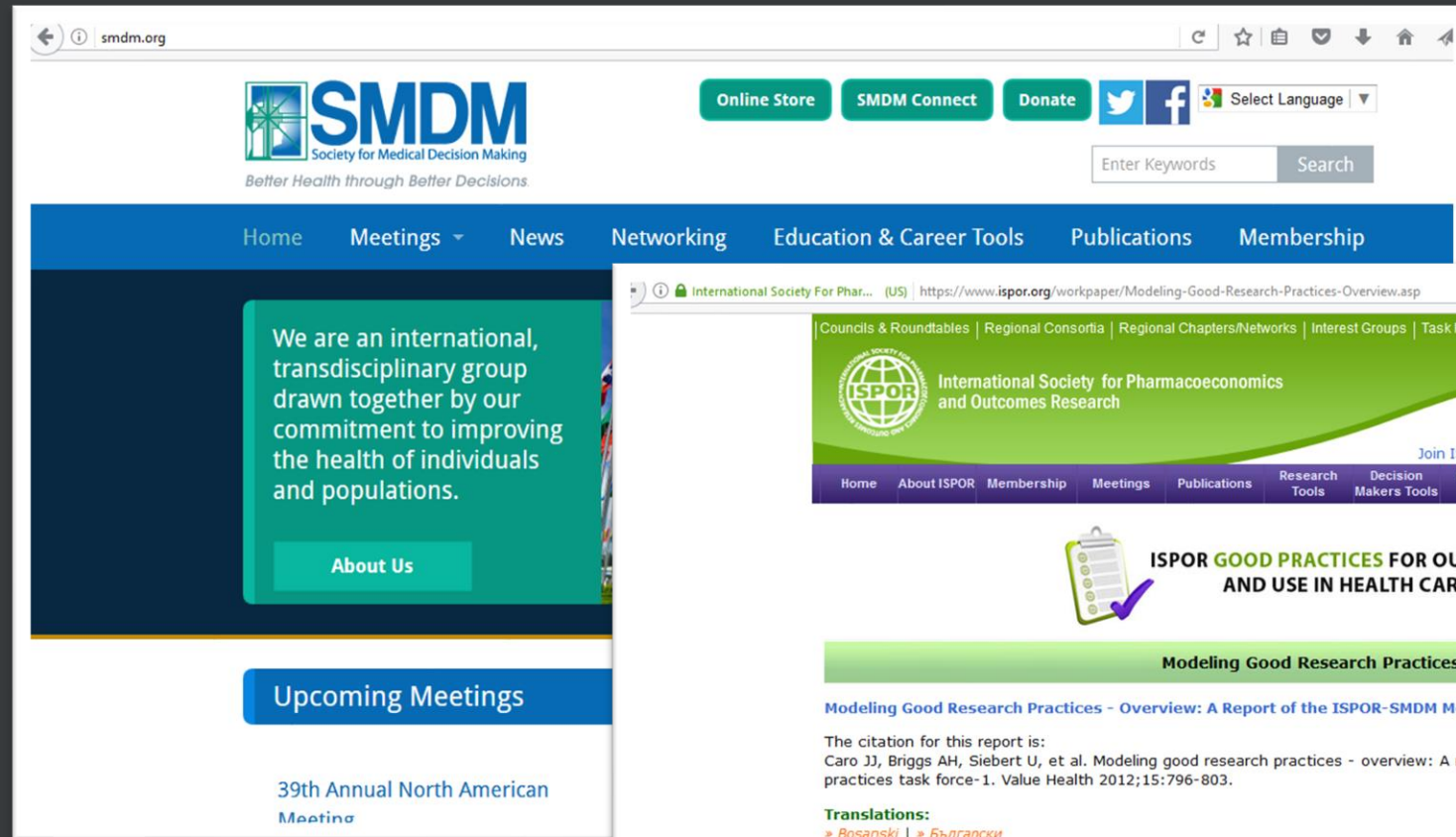
$$\frac{\partial F}{\partial t} - \frac{\partial F}{\partial a} = \alpha_2 + \gamma_2 \mathbb{1}_{[0,dt)}(a) \Psi(a, t) - F \delta_2$$



Cross Model Validation

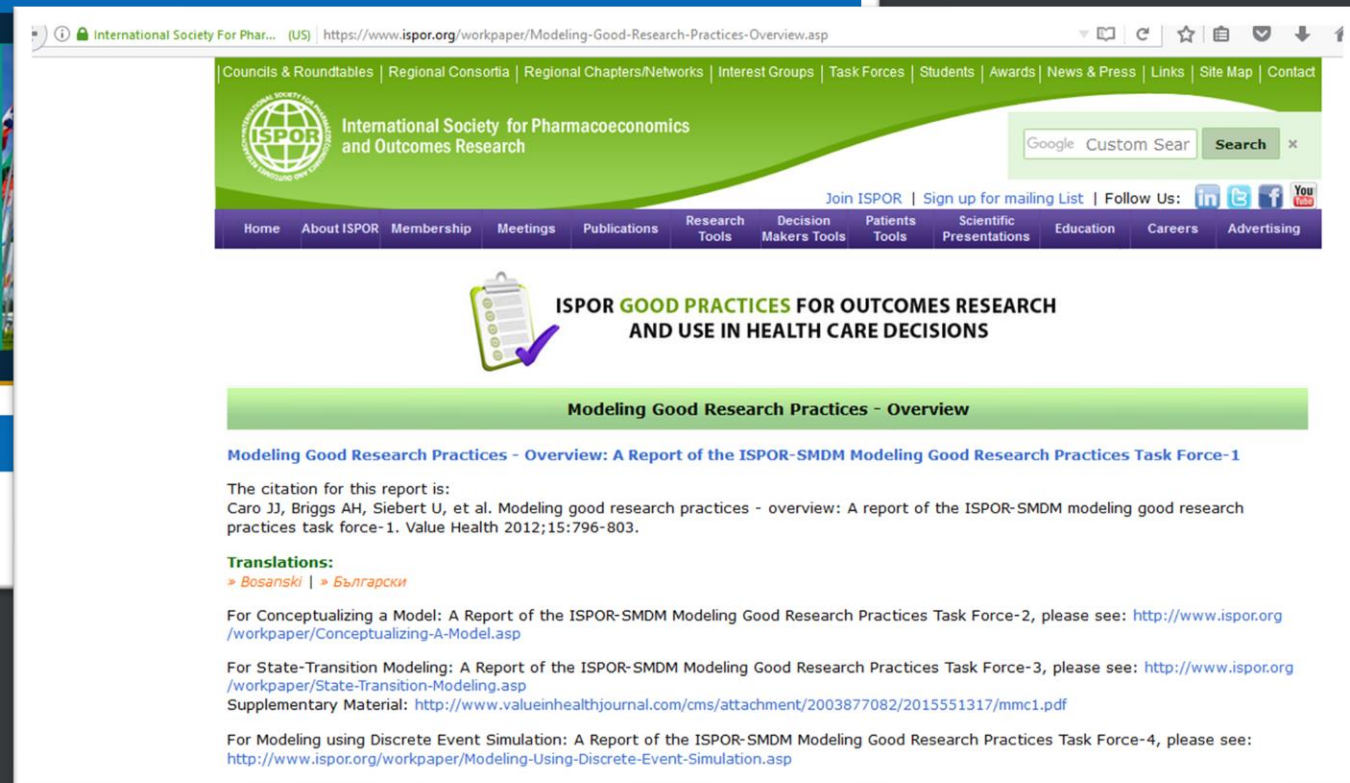
10 Concepts to Integrate

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The screenshot shows the SMDM website (smdm.org) with the following content:

- Header: SMDM Society for Medical Decision Making, Better Health through Better Decisions. Navigation buttons: Online Store, SMDM Connect, Donate, Social media icons, Select Language dropdown, Search bar.
- Menu: Home, Meetings, News, Networking, Education & Career Tools, Publications, Membership.
- Main Content: A green box with the text: "We are an international, transdisciplinary group drawn together by our commitment to improving the health of individuals and populations." Below it is an "About Us" button.
- Footer: "Upcoming Meetings" section with "39th Annual North American Meeting".



The screenshot shows the ISPOR website (www.ispor.org) with the following content:

- Header: International Society for Pharmacoeconomics and Outcomes Research. Navigation: Councils & Roundtables, Regional Consortia, Regional Chapters/Networks, Interest Groups, Task Forces, Students, Awards, News & Press, Links, Site Map, Contact. Search bar: Google Custom Search.
- Menu: Home, About ISPOR, Membership, Meetings, Publications, Research Tools, Decision Makers Tools, Patients Tools, Scientific Presentations, Education, Careers, Advertising.
- Main Content: "ISPOR GOOD PRACTICES FOR OUTCOMES RESEARCH AND USE IN HEALTH CARE DECISIONS" with a checklist icon. Below it is a green bar: "Modeling Good Research Practices - Overview".
- Text: "Modeling Good Research Practices - Overview: A Report of the ISPOR-SMDM Modeling Good Research Practices Task Force-1".
- Citation: "The citation for this report is: Caro JJ, Briggs AH, Siebert U, et al. Modeling good research practices - overview: A report of the ISPOR-SMDM modeling good research practices task force-1. Value Health 2012;15:796-803."
- Translations: "Bosanski" | "Български".
- Links: "For Conceptualizing a Model: A Report of the ISPOR-SMDM Modeling Good Research Practices Task Force-2, please see: http://www.ispor.org/workpaper/Conceptualizing-A-Model.asp", "For State-Transition Modeling: A Report of the ISPOR-SMDM Modeling Good Research Practices Task Force-3, please see: http://www.ispor.org/workpaper/State-Transition-Modeling.asp", "Supplementary Material: http://www.valueinhealthjournal.com/cms/attachment/2003877082/2015551317/mmc1.pdf", "For Modeling using Discrete Event Simulation: A Report of the ISPOR-SMDM Modeling Good Research Practices Task Force-4, please see: http://www.ispor.org/workpaper/Modeling-Using-Discrete-Event-Simulation.asp".

Exact Comparable Definitions of Models. E.g. Formal Definition of a Cellular Automaton

Cell

Definition
The set of all cells will be denoted M . Consequently a cell is an unique (A set contains by definition unique elements.) element $m \in M$.

There is no limitation in the number of cells. However if only one cell is observed, we call the cellular automaton a trivial cellular automaton.

Special Case
We may call a cellular automaton with a finite number of cells **finite** or otherwise **infinite**.

Neighbours

Let $k \in \mathbb{N} \setminus \{0\}$.

Definition
A tuple $J := (j_1, \dots, j_k) \in (\mathbb{Z}^d)^k$ where $j_\alpha \neq j_\beta$ is called a **relative index tuple** and for $i \in \mathbb{Z}^d$ the addition respectively subtraction $J \pm i := (j_1 \pm i, \dots, j_k \pm i)$ is well-defined.

Definition
Given a relative index tuple J we define the **index translation** T_J of an index i by $T_J(i) := i + J$ and call the result an **absolute index tuple**.

Note that $i + J$ is not necessarily a subset of I .

Theorem
 I is a vector space? (TODO)

Definition
For a cell m_i from an indexed set of cells with index set I and relative index tuple J we use the resulting absolute index tuple $T_J(i) = (i_1, \dots, i_k) \in (\mathbb{Z}^d)^k$, to define the **neighbourhood** of m_i as $N_{m_i, J} := (n_1, \dots, n_k) \in (M \cup \{\emptyset\})^k$ where

$$n_\alpha := \begin{cases} m_{i_\alpha} = T^{-1}(i_\alpha) & i_\alpha \in I \\ \emptyset & i_\alpha \notin I \end{cases} \quad \alpha \in \{1, \dots, k\}.$$

Furthermore we call k the **neighbourhood dimension**.

The *non-existent cell* \emptyset is required in order to maintain the original tuple structure of the neighbourhood and to be able to indicate that indices which are outside the index set do not refer to a cell.

Special Case

- A cell lies in its own neighbourhood (**reflexive**) if and only if $0 \in \mathbb{Z}^d$ is part of the relative index tuple.
- An index tuple respectively neighbourhood is neither necessarily **symmetric**, **bidirectional** nor **local**.

Theorem
It is not unusual that neighbourhoods are of **local character**, which means that the neighbourhood relation is defined by the distance between cells. In this case, the index translation can be replaced by the appropriate function T_{metric} . It is however always possible to find an equivalent relative index tuple J and use T_J .

Definition
For an indexed set of cells (M, I, T, T^{-1}) and an index translation T the **neighbourhood mapping** is defined by $\mathcal{N} := T^{-1} \circ T \circ \mathcal{I} : M \rightarrow I \rightarrow \mathbb{Z}^k \rightarrow (M \cup \{\emptyset\})^k : m_i \mapsto i \mapsto (i_1, \dots, i_k) \mapsto (n_1, \dots, n_k)$.

Border

Two types of "special cells" can be distinguished:

Definition
A **border-cell** is a cell, which is located at the boundary of the lattice.

Definition
If the absolute index tuple of a cell m_i does not lie completely within the index set $(T_J(i) \not\subseteq I)$, we talk of (a cell with) a **degraded neighbourhood**.

Optimal boundary conditions must be applied to cells with a degraded neighbourhood!

In order to manipulate the geometry of the lattice (periodic boundary conditions for example) we modify the index translation:

Definition
Given a relative index tuple J , the **generalised index translation** is defined by $T_J : I \rightarrow I^k : i \mapsto (i_1, \dots, i_k)$ where

$$i_\alpha := \begin{cases} i + j_\alpha & i + j_\alpha \in I \\ \tau(i + j_\alpha) & i + j_\alpha \notin I \end{cases} \quad \alpha \in \{1, \dots, k\}$$

and $\tau : \mathbb{Z}^d \setminus I \rightarrow I$.

Special Case
What kind of preconditions to or characteristics of τ generate which type geometry? (TODO)

- a toroid geometry for a two-dimensional cellular automaton (coll. "periodic boundary condition") can be achieved by using the modulus function...
- the same is true for a cylindrical geometry...
- what about a spherical surface?

*actually in this situation there is no boundary!

State

Let $k \in \mathbb{N} \setminus \{0\}$ be the neighbourhood dimension.

Definition
There exists a (temporary) state mapping from the set of all cells M to the set of all possible states S , which assigns a state to each cell $S : M \rightarrow S : m \mapsto S(m) = s$. We also use S as $S : (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k : (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k)$ where

$$s_\alpha := \begin{cases} S(m_\alpha) & m_\alpha \in M \\ \emptyset & m_\alpha \notin M \iff m_\alpha = \emptyset \end{cases} \quad \alpha \in \{1, \dots, k\}$$

The *non-existent state* \emptyset is required to maintain the tuple structure and to indicate a degraded neighbourhood.

Update

Let $k \in \mathbb{N} \setminus \{0\}$.

Definition
An **update rule (also update rules)** is a mapping $\mathcal{F} : (S \cup \{\emptyset\})^k \rightarrow S : (s_1, \dots, s_k) \mapsto s$

To calculate a new state for a cell:
 $\mathcal{F} \circ \mathcal{S} \circ \mathcal{N} : M \rightarrow (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k \rightarrow S : m \mapsto (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k) \mapsto s$

actually, in detail:
 $\mathcal{F} \circ \mathcal{S} \circ T^{-1} \circ T \circ \mathcal{I} : M \rightarrow I \rightarrow \mathbb{Z}^k \rightarrow (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k \rightarrow S : m \mapsto i \mapsto (i_1, \dots, i_k) \mapsto (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k) \mapsto s$

Remark

- An update rule (update rule set) can be the explicit definition of a mapping but also a (continuous) function or a combination of functions.
- Since degraded neighbourhoods contain non-existent cells respectively states \emptyset , an update rule must react on a degraded neighbourhood and implement (arbitrary) "boundary conditions".

Warning

- An update rule never defines the geometry of the lattice!
- We exclude stochastic update rules from a basic definition since the necessary introduction of a probability space would be an extension to our formal definition (TODO).

Definition
An update rule must be defined for every possible neighbourhood configuration. Otherwise we deal with **undefined behaviour**.

- An update rule must be **compatible with the index set**: All occurring degradations of neighbourhoods must be taken into account.
- An update rule must be **self-contained** since all possible state-configurations⁹ arise from the update rule.

⁹except for the initial condition, see later.

Special Case

- The set of all possible states S may contain a **finite** or **infinite** number of states.
- A "state-space" (coll.) can be a vector space, a ring or any other algebraic structure.
- By introducing a partitioning on the set of all possible states, different cell types can be distinguished.

A non-trivial cellular automata features more than one different element in S .

Global State

Definition
The state of all cells is accumulated in the temporary state mapping \mathcal{S} which can be identified with an element of $\mathfrak{S} := S^M$. We then also call \mathcal{S} the (temporary) **global state**.

Definition
Given a neighbourhood mapping \mathcal{N} , a temporary state mapping \mathcal{S} and an update rule \mathcal{F} we define the **local evolution operator** $\mathcal{S} := \mathcal{F} \circ \mathcal{S} \circ \mathcal{N} : M \rightarrow (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k \rightarrow S : m \mapsto (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k) \mapsto s$.

We can see that a local evolution operator is a state mapping and a global state.

Iteration

Definition
A (global) **evolution operator** is a mapping $\mathcal{E} : \mathfrak{S} \rightarrow \mathfrak{S} : \mathcal{S} \mapsto \mathcal{S}' := \mathcal{F} \circ \mathcal{S} \circ \mathcal{N}$.

Definition
An **iterative process** can be obtained by defining $\mathcal{S}_{t+1} := \mathcal{E}(\mathcal{S}_t) = \mathcal{F} \circ \mathcal{S}_t \circ \mathcal{N}$ where $n \in \mathbb{N}$.

Definition
For an iteration, an **initial state** or **initial condition** \mathcal{S}_0 must be given. It is necessary that the initial condition is **compatible with the update rules**.

Only the states of the cells and accordingly the state mapping may change during iteration!

Automaton

Definition
A **cellular automaton** comprises ...

- an indexed set of cells
- an initial state mapping \mathcal{S}_0
- a relative index tuple J
- a generalised index translation (only if the geometry shall be "manipulated")
- an update rule \mathcal{F}
- the iterative application of an evolution operator

Lattice

Definition
If M is a set of cells, we call M **indexed** or **regularly arranged** if there exists a bijective mapping $\mathcal{I} : M \rightarrow I : m \mapsto \mathcal{I}(m) = i$ between M and an index set I . We call \mathcal{I} an **index mapping** and also use \mathcal{I} for mapping tuples of cells onto tuples of indices $\mathcal{I} : M^k \rightarrow I^k : (m_1, \dots, m_k) \mapsto (\mathcal{I}(m_1), \dots, \mathcal{I}(m_k)) = (i_1, \dots, i_k)$ where $k \in \mathbb{N} \setminus \{0\}$.

$(m_{\mathcal{I}(m)})_{m \in M}$ is the natural indexing for a set of indexed cells.

Definition
Since by now all cells are "arranged" or indexed using an index set $I \subseteq \mathbb{Z}^d$, we call d the **dimension** of the cellular automaton.

Definition
A subset I of \mathbb{Z}^d is called **connected** if for each two elements $a, b \in I$ there exists a series of elements $(z_\alpha)_{\alpha \in \mathbb{N}} \subset I$ for which $\|z_\alpha - z_{\alpha+1}\| = 1 \forall \alpha \in \mathbb{N}$ and for which $a, b \in (z_\alpha)_{\alpha \in \mathbb{N}}$.

Definition
We call a connected subset $I \subseteq \mathbb{Z}^d$, $d \in \mathbb{N} \setminus \{0\}$ an **index set**.

Special Case
Despite requiring a connected index set, this definition does not exclude exotic or even absurd index sets (i.e. lattices). (TODO)

G. Schneckreither, Thesis: "Developing Mathematical Formalisms for Cellular Automata in Modelling and Simulation"; 2014

G. Schneckreither, N. Popper, F. Breitenecker: "Methods for Cellular Automata and Evolution Systems in Modelling and Simulation"; IFAC PapersOnLine, 48 (2015), 1; S. 1 - 944.

Exact Comparable Definitions of Models. E.g. Formal Definition of a Cellular Automaton

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
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5. Different Methods for Different Questions (Complexity)
6. Comparability of Results

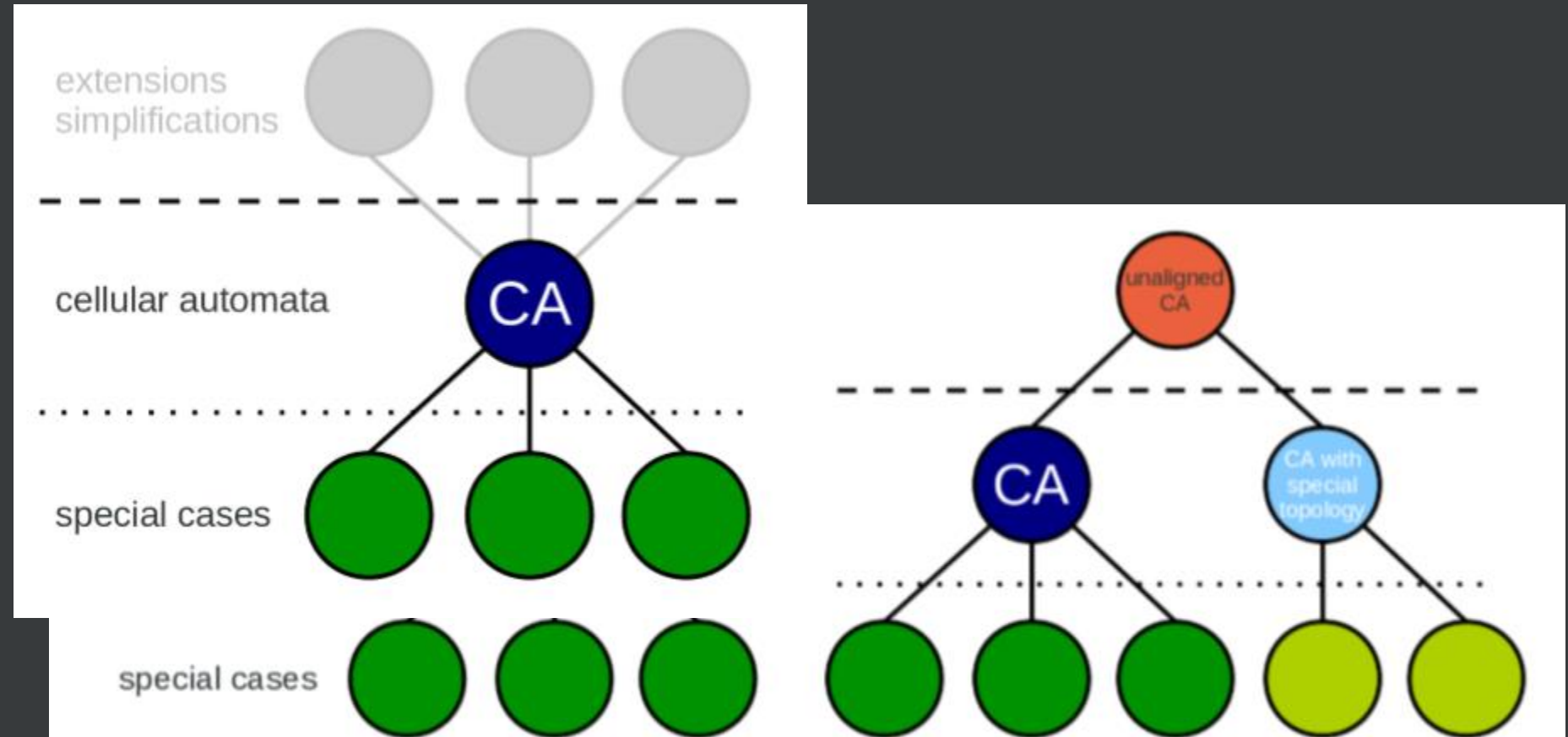


Figure 1: Different hierarchies for the classification of cellular automata.

Concept 7

10 Concepts to Integrate

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6. Comparability of Results
7. Make it Understandable

Complicated Models and Various Results

- Good data & an “improvable” Model
- Right Model & good Results
- Nobody understands.

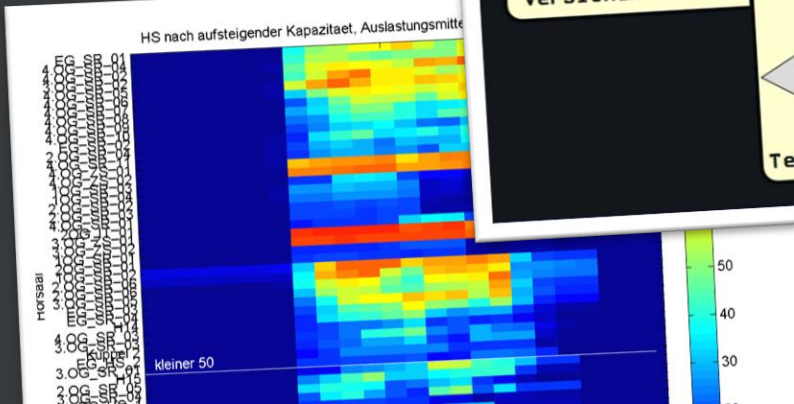
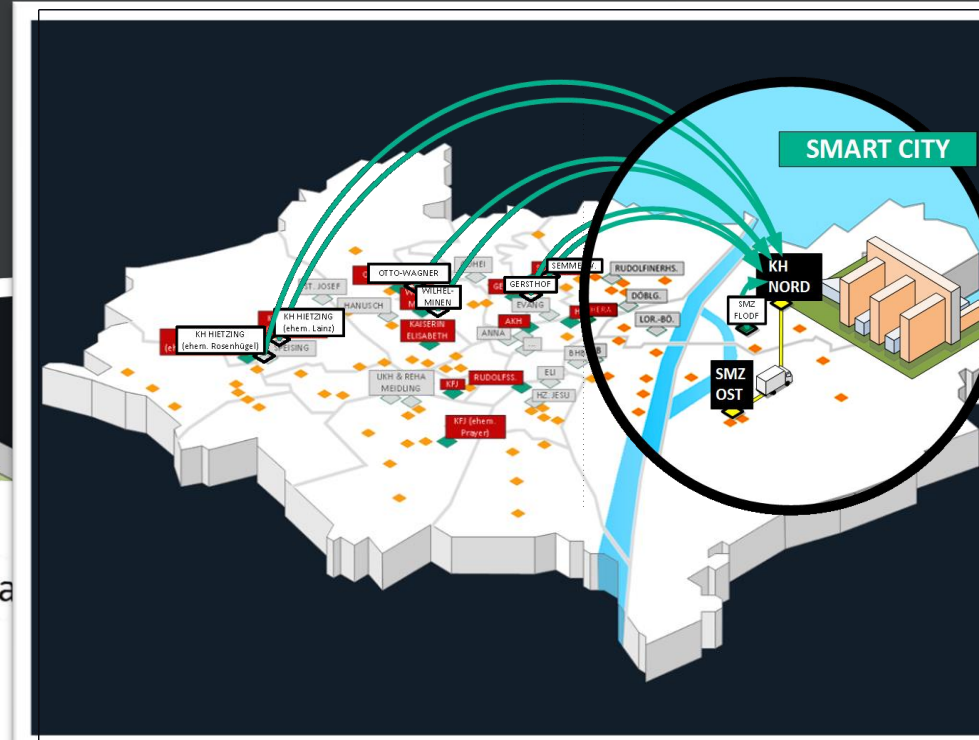
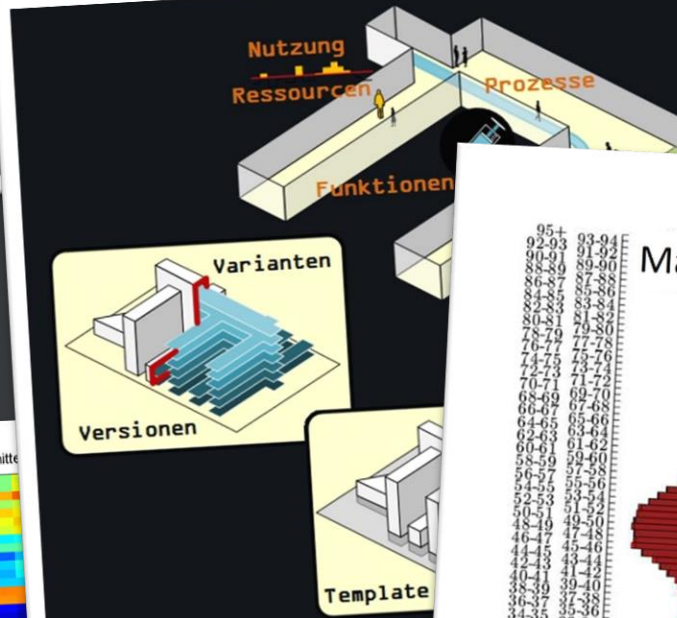
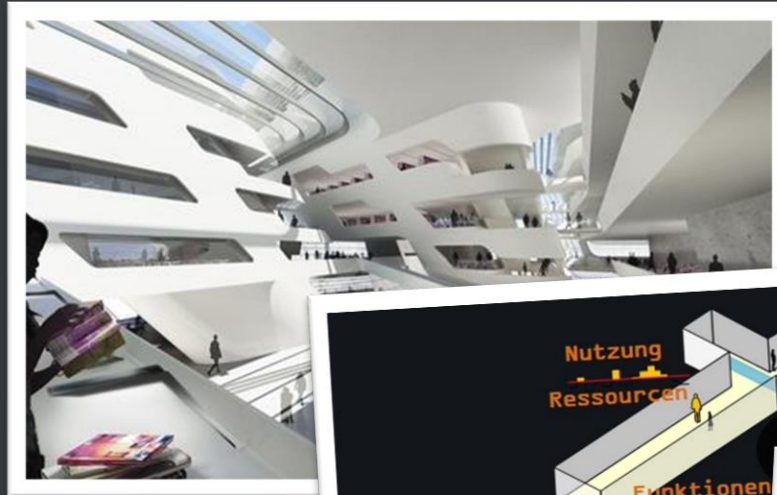
-> Change Management & Interdisciplinarity

**METHOD: Data Representation & Human
Computer Interfaces (HCI)**

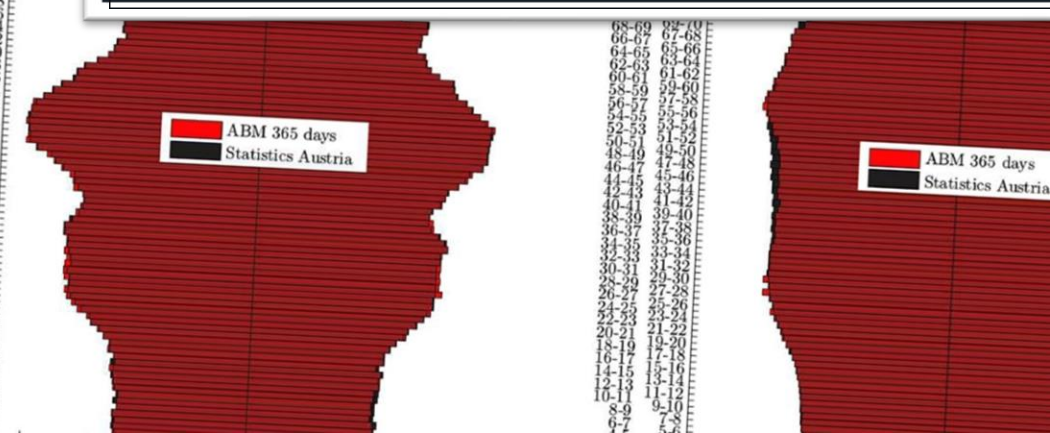
Concept 7

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95+ 93-94
92-93 91-92
90-91 89-90
88-89 87-88
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6-7 5-6
4-5 3-4
2-3 1-2
0-1



Having Fun with Epidemics and Herd Immunity...

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„On the benefits of explaining herd immunity in vaccine advocacy“, Nature Human Behaviour, 6.3.2017

Concept 8

10 Concepts to Integrate

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6. Comparability of Results
7. Make it Understandable
8. Open and Independent Solutions

Open and Independent Solutions

- Publication limited because of Economic Exploitation & Stakeholder Interests
- Lack of Comparability of Different Models
- Rules and Guidelines needed

METHOD: Open Access & Public Domain

Additional Sources: Example Web Scraping

10 Concepts to Integrate

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Integration of Data from Web Pages of Physicians – Actualisation monthly

The screenshot shows the website 'Arztstuche - Ärztekammer für Tirol'. The browser address bar displays 'www.aektirol.at/arztstuche?p_p_id=medsearchportlet_WAR_aektirmedsearchportlet&p_p_lifecycle=0&_medsearchportlet_WAR_aektirmedsearch...'. The website has a navigation menu with items: 'Kammer', 'Mitgliederservice', 'Ausbildung', 'Fortbildung', 'Wohlfahrtsfonds', and 'Kundmachungen'. The main banner features the text 'Ärztekammer für Tirol' with a red bird logo. Below the banner is a search bar with the text 'Suche...'. The search results section is titled 'Arztstuche' and shows 'Suche : Liste'. The results are filtered to show 'Ihre Ergebnisse' with '303 Treffer (259 Ärzte) in 134 Ordinationen und 138 Anstellungen'. The first result is for 'Dr. Karl Alexander Aggstein', with details on his medical specialty and location.

Anstellung	Spezialität	Adresse	Ort	Krankenhaus
Anstellung	Allgemeinchirurgie und Viszeralchirurgie	Bahnhofstraße 10	St. Johann in Tirol	A.ö. Bezirkskrankenhaus St. Johann i.T. - Abteilung Chirurgie
Anstellung	Allgemeinmedizin	Bahnhofstraße 10	St. Johann in Tirol	A.ö. Bezirkskrankenhaus St. Johann i.T. - Abteilung Chirurgie

Health Care: Physicians

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- Physicians total: 16.500
- Medical Practices: 20.219
- Medical Practices with GPS Coordinates: 13.607
- Total Hours per Week: 193.899 h
- GPs:
 - Medical Practices: 8.062
 - Contracts with Social Incurances: 3841
 - Total Hours per Week: 89.594 h

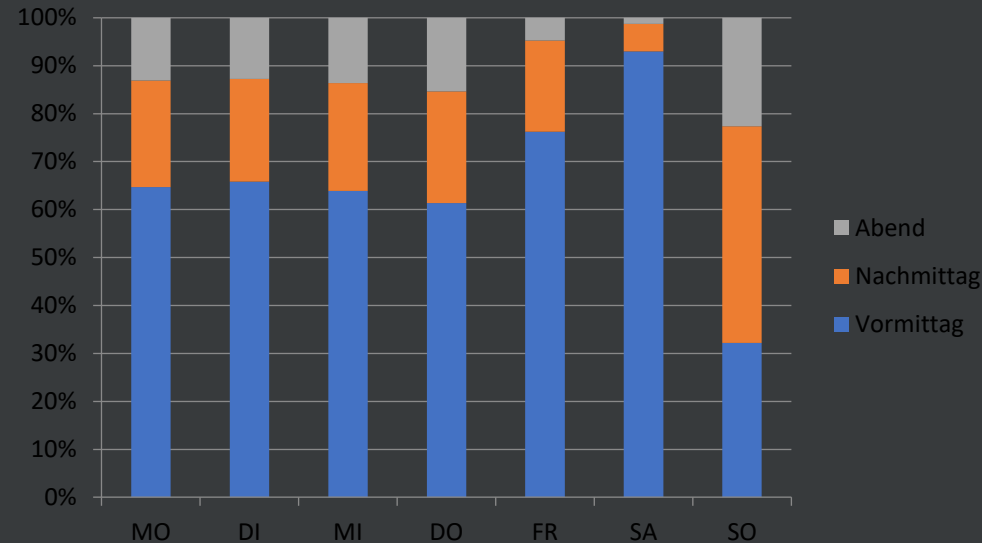
Health Care: Physicians

10 Concepts to Integrate

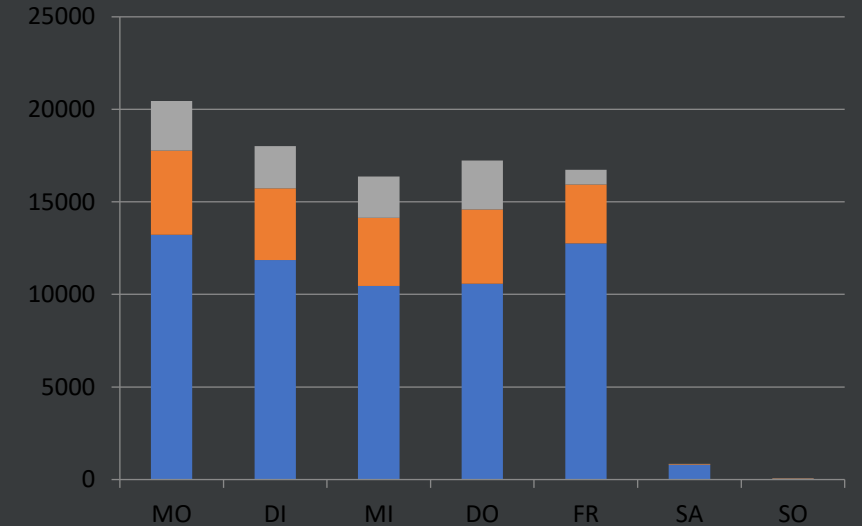
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Example Medical Practices Hours :

relative



absolut



Concept 9

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
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4. Reproducible Processes
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6. Comparability of Results
7. Make it Understandable
8. Open and Independent Solutions
9. Priority for Data Security and Stake Holder Interests

Priority for Data Security and Stakeholders

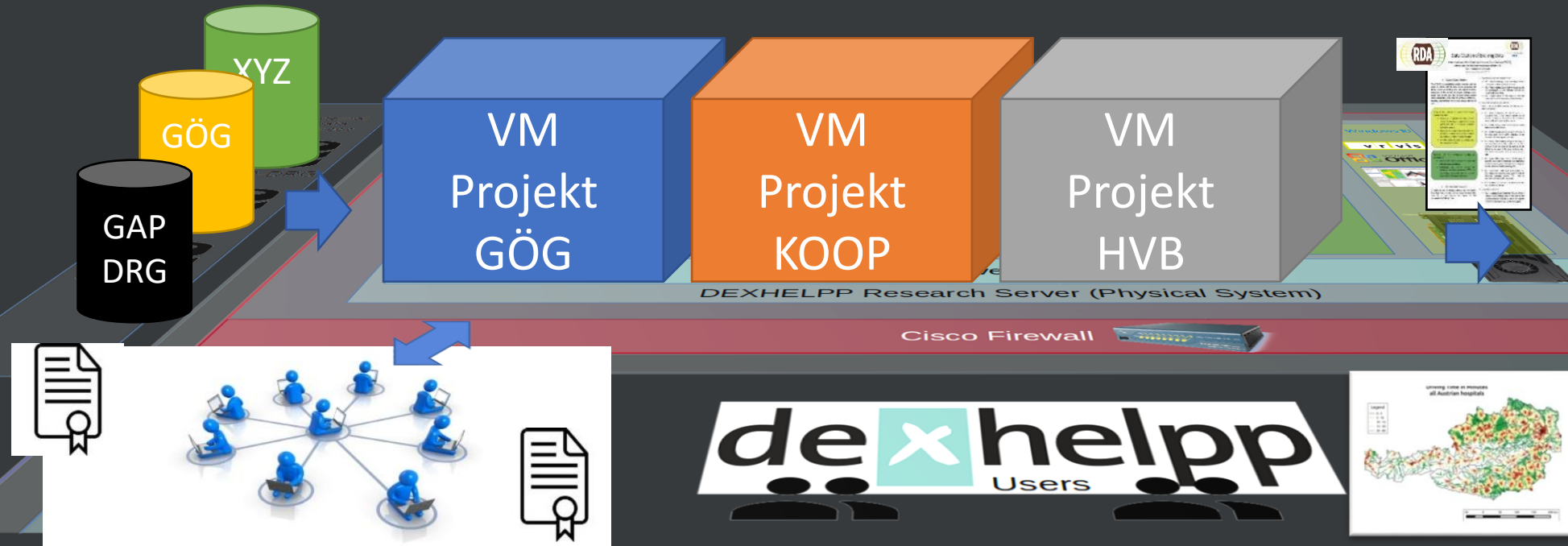
- Additional Requirements e.g. with the new EU General Data Protection Regulation
- Quality can be reduced by missing data
- Transparent Processes are Needed in advance

METHOD: Data Security & Governance

DEXHELPP Research Server

10 Concepts to Integrate

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1. **Data:** Temporary and restricted to well defined Research
2. **Access:** Restricted and Reproducible
3. **Methods:** Usage of all Modelling & Simulation Methods
4. **Export:** Well defined Rules for Export and Usage

Processes according to Guidelines of Research Data Alliance (RDA)

Concept 10

10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
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4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results
7. Make it Understandable
8. Open and Independent Solutions
9. Priority for Data Security and Stake Holder Interests
10. Broad Applications (Health System, Energy, Industry, Energy, Mobility, Infrastructure)

Broad Applications for Models

- Re-Use of Models necessary, because of...
- ...Resources & Quality
- Research Questions become interdisciplinary and “communicate”

METHOD: Co-Simulation, Multi Method Modelling

Multi-Method Airport Modelling

Exploring Advantages of Multi-Method Modelling and its applications in large socio-technical infrastructure systems

Glock et al, ASIM 2016

10 Concepts to Integrate

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Multi-Method Airport Modelling

10 Concepts to Integrate

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The airport city – A large complex socio-technical system

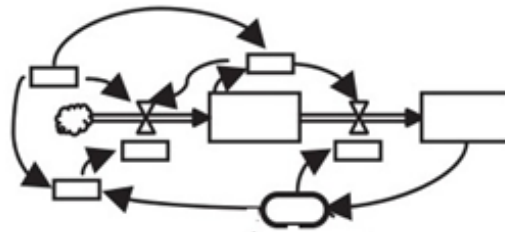


Multi-Method Airport Modelling

10 Concepts to Integrate

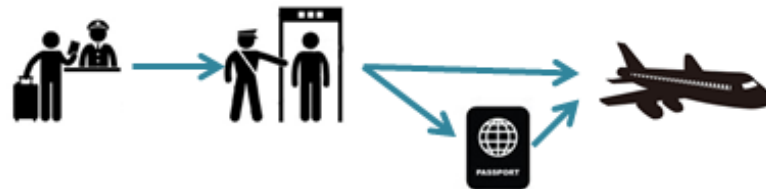
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results
7. Make it Understandable
8. Open and Independent Solutions
9. Priority for Data Security and Stake Holder Interests
10. Broad Applications (Health System, Energy, Industry, Energy, Mobility, Infrastructure)

System Dynamics



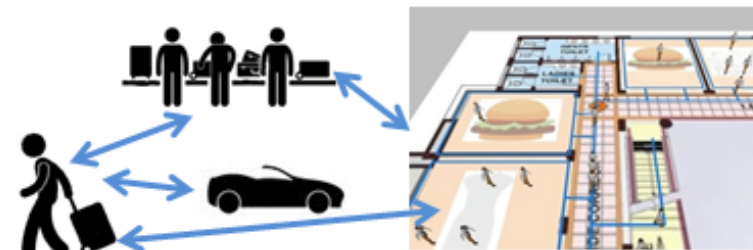
Aggregated states
Time-continuous
Causal Links & Feedback
average computing times

Discrete Events Simulation



Individuals (Entities) & Resources
Changes in discrete time (Events)
Given processes
average computing times

Agent-Based Models



Individuals
Time-continuous
Spatial aspects
Interaction (agents, environment)
High computing times

Multi-Method Airport Modelling

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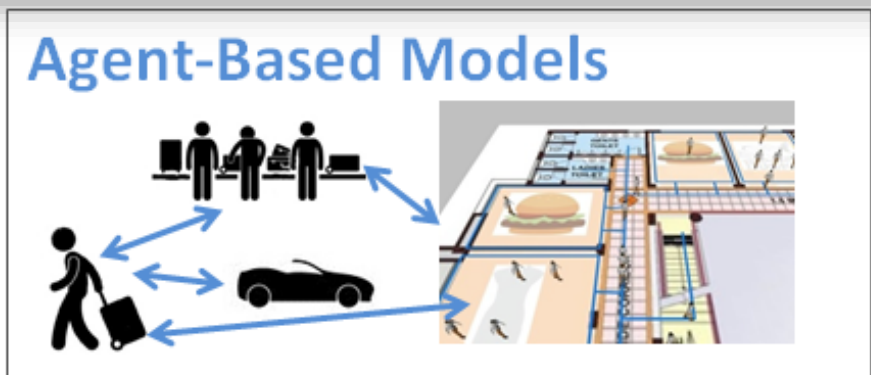
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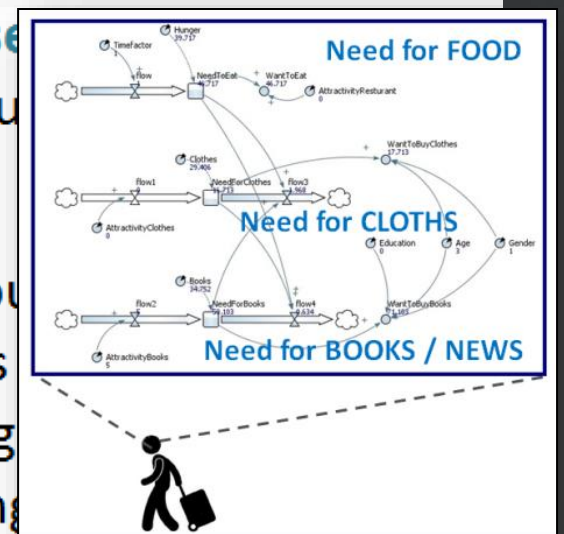
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Methods & Concepts to be Developed and Integrated

- | | | | | |
|--|--|--|---|---|
| <ul style="list-style-type: none"> • Explorative Visual Computing – Visual Analytics and Statistics | <ul style="list-style-type: none"> • Data Processes (Integration & Linkage) & Modelling Tools (Parametrization & Calibration) | <ul style="list-style-type: none"> • Modular Models, Coupling of Models | <ul style="list-style-type: none"> • Validation & Data Citation | <ul style="list-style-type: none"> • Methods for Choosing Models |
| <ul style="list-style-type: none"> • Comparative Modelling | <ul style="list-style-type: none"> • Data Representation & Human Computer Interfaces (HCI) | <ul style="list-style-type: none"> • Open Access & Public Domain | <ul style="list-style-type: none"> • Data Security & Governance | <ul style="list-style-type: none"> • Co-Simulation, Multi Method Modelling |
| <ul style="list-style-type: none"> • Co-Simulation, Multi Method Modelling | <ul style="list-style-type: none"> • Open Access & Public Domain | <ul style="list-style-type: none"> • Data Security & Governance | <ul style="list-style-type: none"> • Data Representation & Human Computer Interfaces (HCI) | <ul style="list-style-type: none"> • Validation & Data Citation |

10
Guidelines

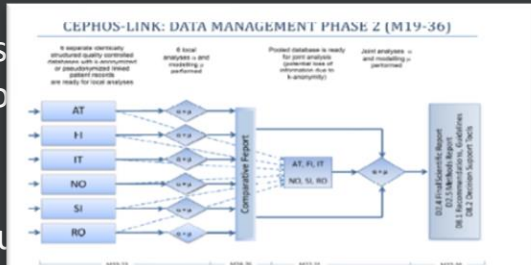
Implementation DEXHELPP

10 Concepts to Integrate

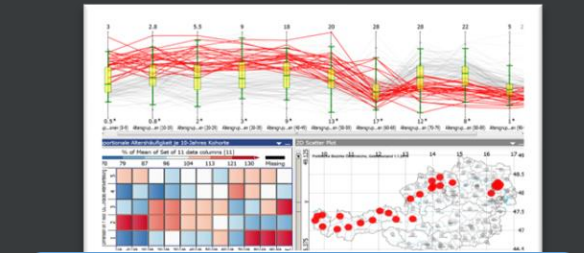
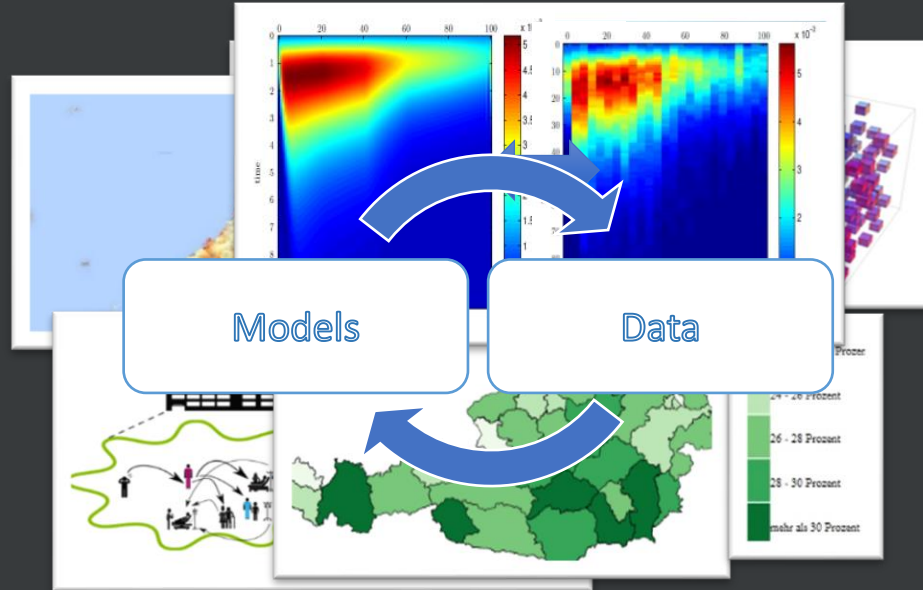
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Projects with and Support for Decision Makers



Formal & Technological Processes



Decision Support & Process Integration (e.g.: HTA)



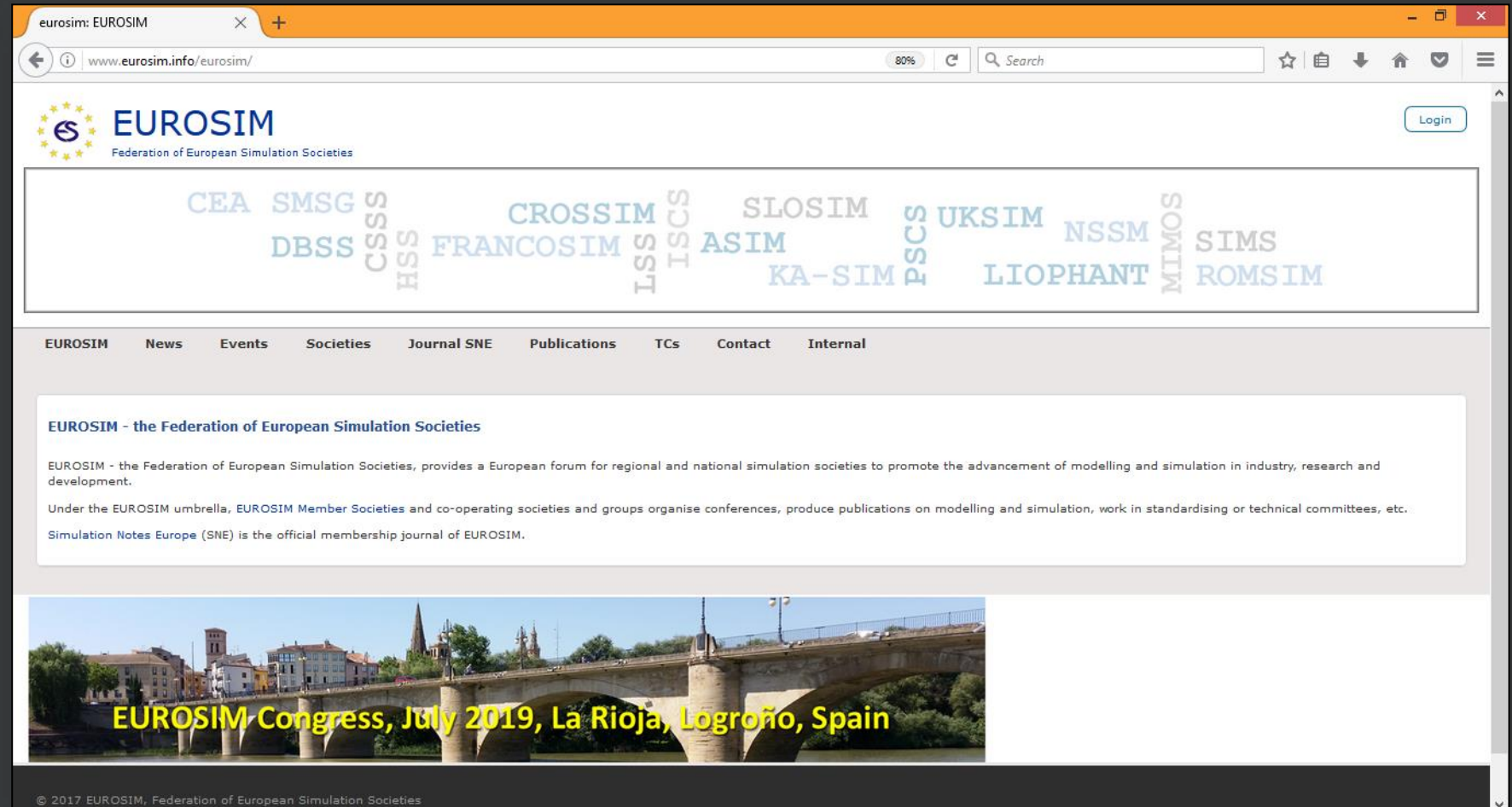
Infrastruktur
DEXHELPP Server & Services

EUROSIM Initiative

DDSS Technical Committee

10 Concepts to Integrate

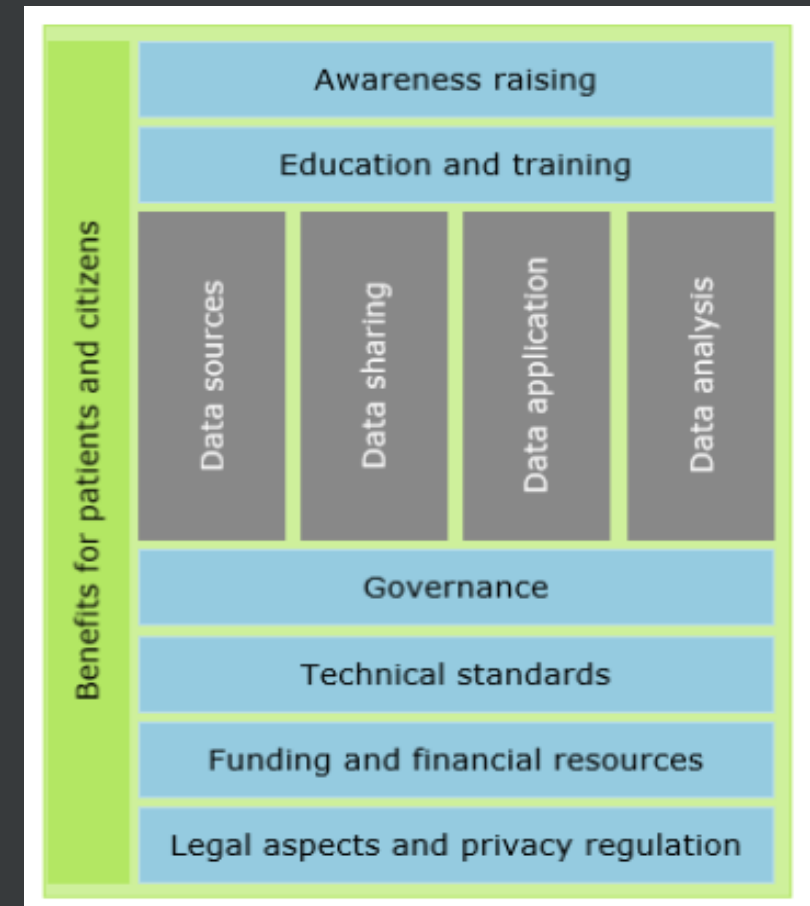
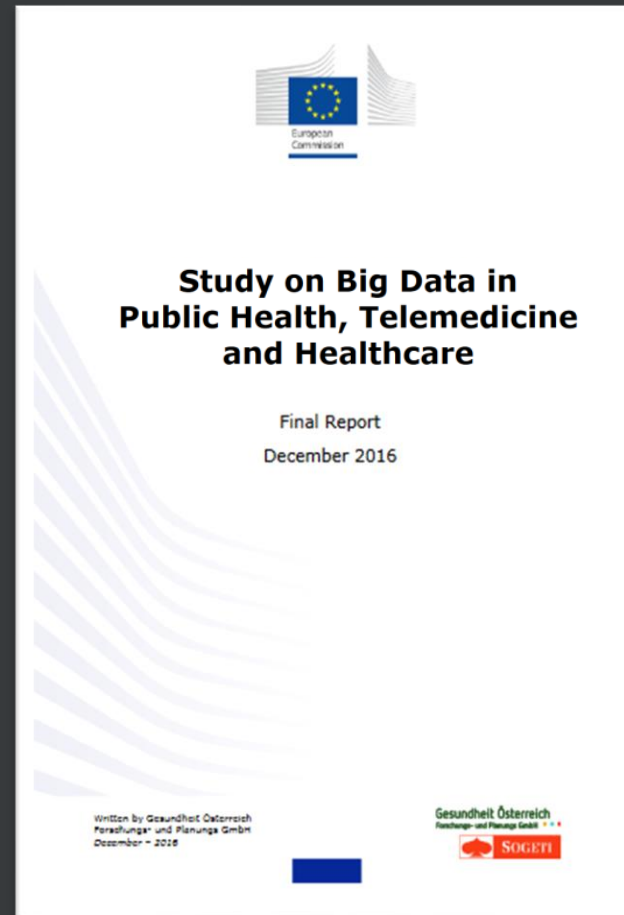
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Studies, Policies & Recommendations

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Overview of fields of policy recommendations

https://ec.europa.eu/health/sites/health/files/ehealth/docs/bigdata_report_en.pdf

Studies, Policies & Recommendations

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The following twenty examples of use of Big Data in Health were identified and selected for further analysis:

- Comet K-Project DEXHELPP – AT
- The Shared Care Platform – DK
- E-Estonia – National Identity Scheme – EE
- AEGLE (An analytics framework for integrated and personalized healthcare services in Europe) – UK, IT, GR, SE, BE, NL, PT, FR
- The Business Intelligence database system – GR
- PASSI (Progressi delle Aziende Sanitarie) – IT
- Arno Observatory – IT
- The Swedish Big Data Analytic Network – SE
- Clinical Practice Research Datalink (CPRD) – UK
- Sentinel Stroke National Audit Programme (SSNAP) – UK
- Hospital Episode Statistics (HES) – UK (England)
- The YODA Project (Yale University open data access) – US
- FDA Adverse Event Network Analyser - US
- CEPHOS-LINK – FI, AT, RO, NO, SI, IT
- Twitter (Adverse drug reactions and public health) – International
- Flatiron – US
- UK Biobank – UK
- Semantic Data Platform for Healthcare (SEMCARE) – DE, NL, AT, UK, ES
- Integrated BioBank of Luxembourg (IBBL) – LU
- Spanish Rare Diseases Registries Research Network (SpainRDR) – ES

Studies, Policies & Recommendations

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Recommendations

- Recommendation 1 on Awareness Raising:
Develop and implement a communication strategy to increase the awareness of the added value of Big Data in Health and encourage a positive public mind set towards Big Data in Health
- Recommendation 2 on Education and Training
Strengthen human capital with respect to the increasing need for a workforce that can utilize the potential of Big Data in Health
- Recommendation 3 on Data Sources:
Expand existing and explore new sources of Big Data in Health and secure their quality and safety
- Recommendation 4 on Open Data and Data Sharing:
Promote open use and sharing of Big Data in Health without compromising patients' rights to privacy and confidentiality
- Recommendation 5 on Applications and Purposes:
Increase target-oriented application of Big Data analysis in health based on the needs and interests of stakeholders including patients

Studies, Policies & Recommendations

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Recommendations

- Recommendation 6 on Data Analysis:
Identify the potentials of Big Data analysis, improve analytical methods and facilitate the use of new and innovative analytical methods
- Recommendation 7 on Governance of Data Access and Use:
Implement governance mechanisms to ensure secure and fair access and use of Big Data for research in health
- Recommendation 8 on Standards:
Develop standards for Big Data in Health to enhance and simplify its application and improve interoperability
- Recommendation 9 on Funding and Financial Resources:
Ensure purposeful investment steered by the European Commission to warrant cost-effectiveness and sustainability
- Recommendation 10 on Legal Aspects and Privacy Regulations:
Clarify and align existing legal and privacy regulation of Big Data in Health

10 Concepts to Integrate

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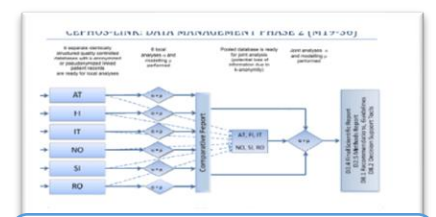
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Outlook

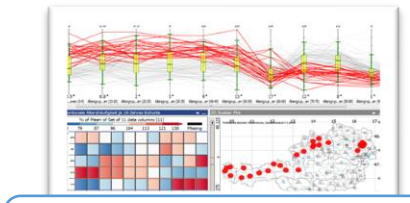
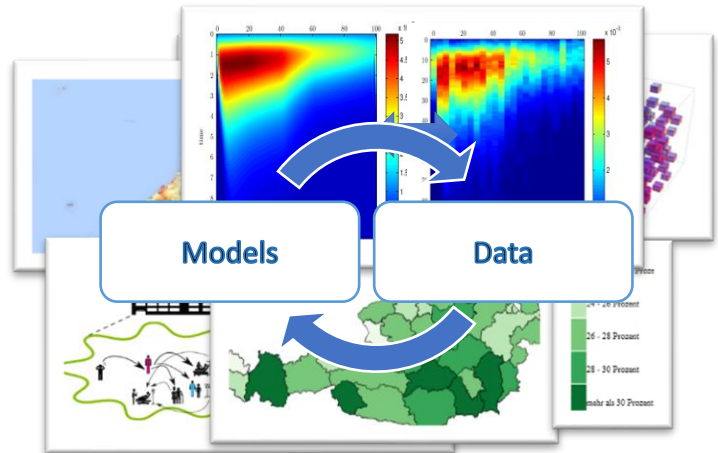
- Putting it all together – a huge challenge, but necessary because of external reasons
- A new “Process for Processes” instead of “Modelling & Simulation” integrating all areas of research (4th Paradigm of Jim Gray)
- Not only for models, but also for the modelling process: not static anymore -> dynamic approaches - also for smaller models
- Huge methodological Differences between “Data Driven” and “Model Driven” Approaches – e.g. Data Driven Journalism
- No “Generic World Model” possible (e.g. Co-Morbidities), but there are Processes to generate controllable complex and dynamic models



Projects with and Support for Decision Makers



Formal & Technological Processes



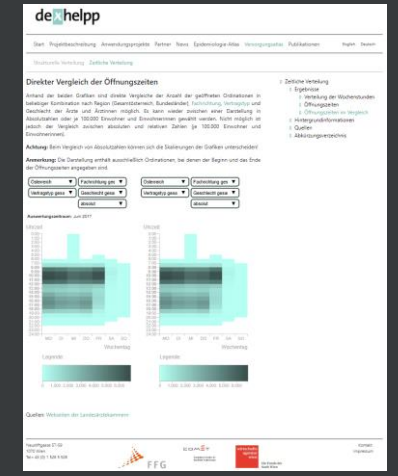
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Competence Centers for Excellent Technologies



Infrastruktur DEXHELPP Server & Services





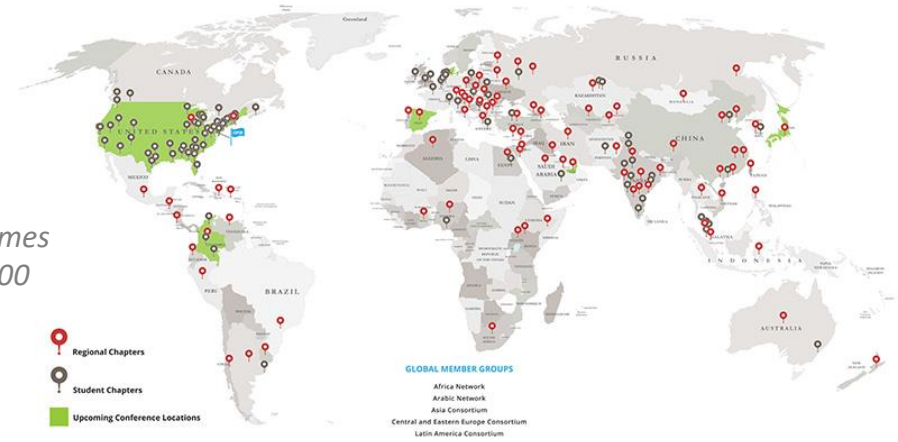
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COLOMBIA



ISPOR Europe 2019

2-6 NOVEMBER 2019 | COPENHAGEN
| DENMARK



Dr. Niki Popper
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