

# Integrated Processes for Modelling & Simulation Health Care Systems

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Coordinator Centre for Computational Complex Systems, TU Wien

CSO dwh GmbH – Simulation Services and Technical Solutions, Vienna

# de<sup>x</sup>helpp

Decision Support for Health Policy and Planning



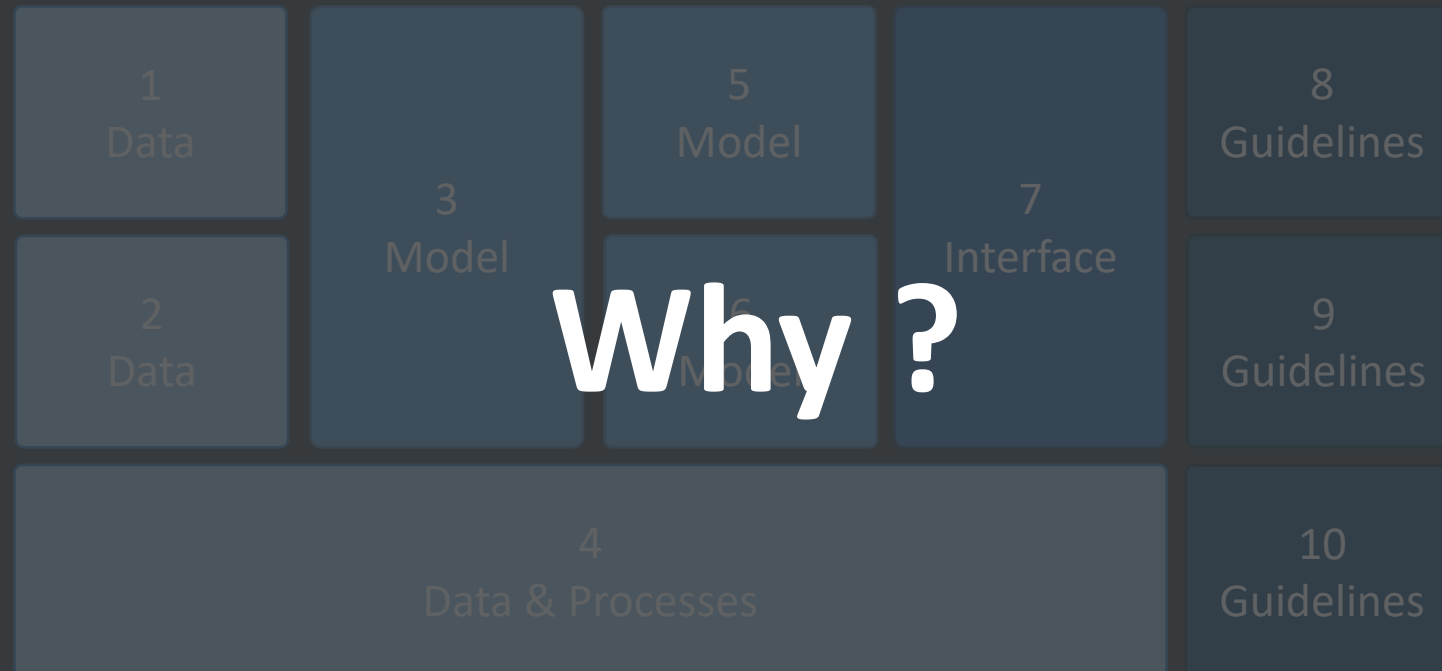
## 10 Concepts to Integrate

### Why?

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

### How?

- > 4<sup>th</sup> 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 CO<sub>2</sub> 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 CO<sub>2</sub>-equivalent (tCO<sub>2</sub>e) emission reductions per year), living car-free (2.4 tCO<sub>2</sub>e saved per year), avoiding airplane travel (1.6 tCO<sub>2</sub>e saved per roundtrip transatlantic flight) and eating a plant-based diet (0.8 tCO<sub>2</sub>e 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?

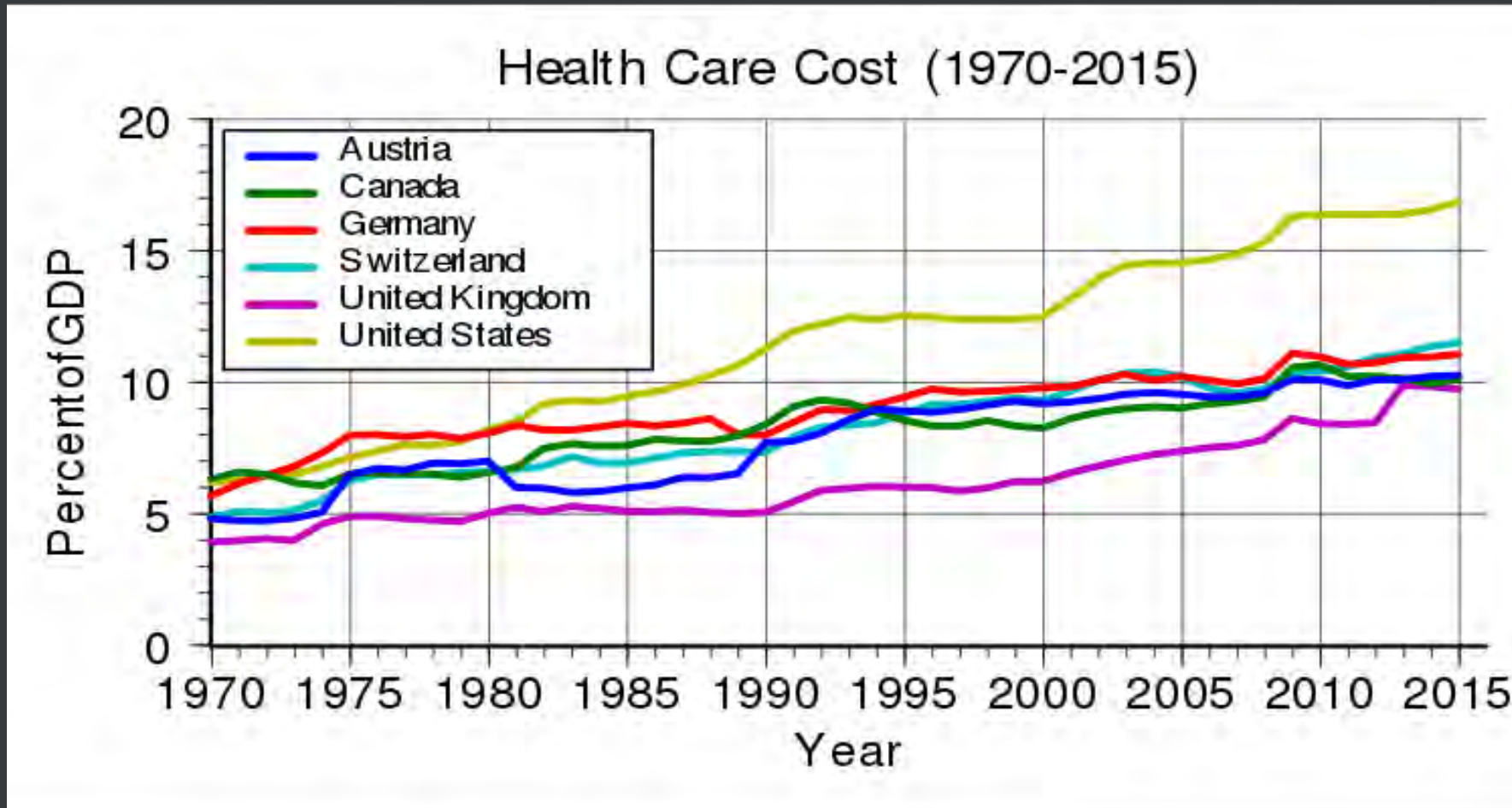
# Economics Example Health Systems

## 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	???

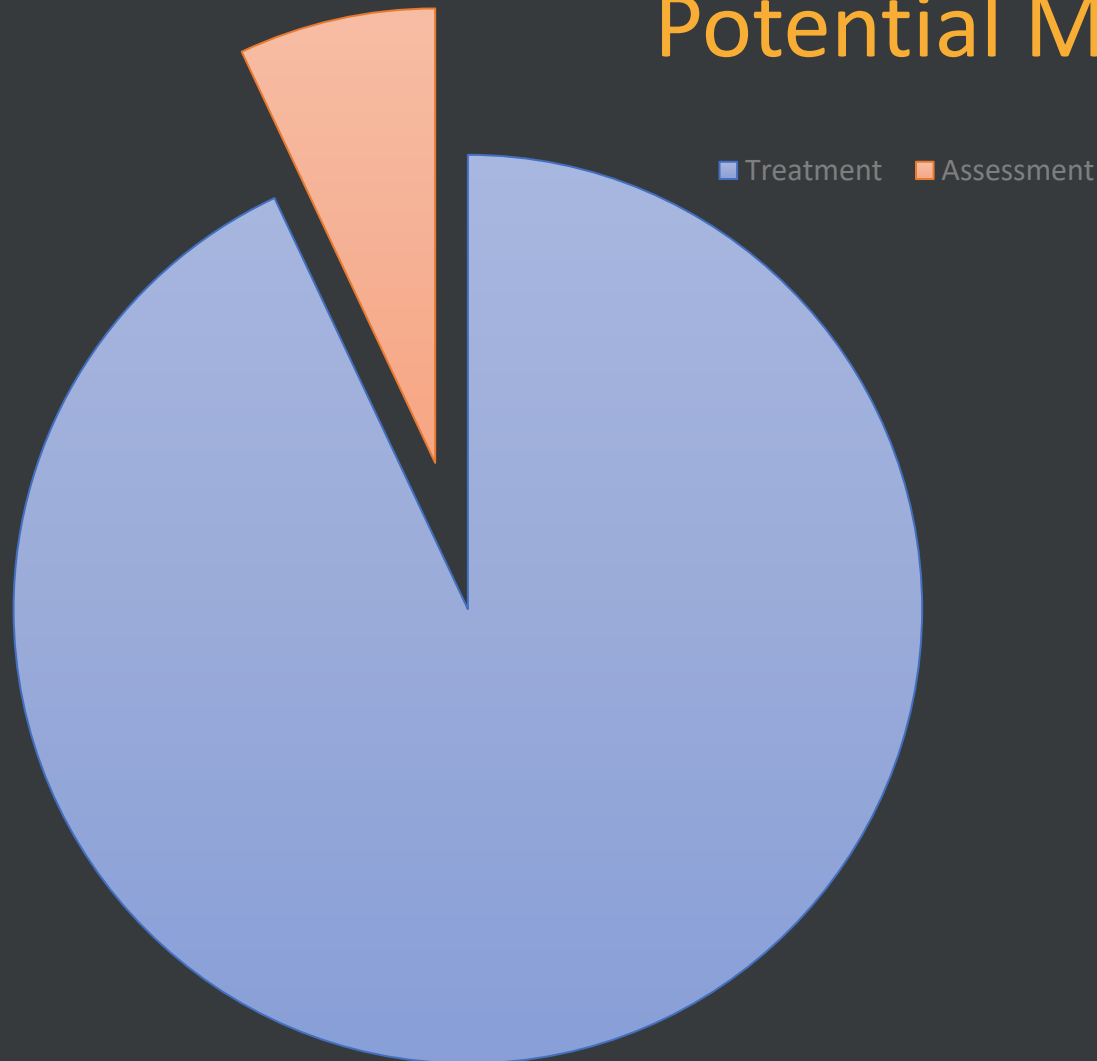
Source: C. Wild, Ludwig Boltzmann Institute for Health Technology Assessment, Vienna

# Economics Example Health Systems



[https://en.wikipedia.org/wiki/Healthcare\\_in\\_Austria](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

New Medication

Cutting Edge Technology

Innovative Diagnosis

Fair Distribution of Health Services

Disease Prevention

Dealing with Increasing Costs

Estimating Reachable Outcomes

Measuring of Adherence

Managing # of Patients

Developing Registers

Benchmarking Interventions

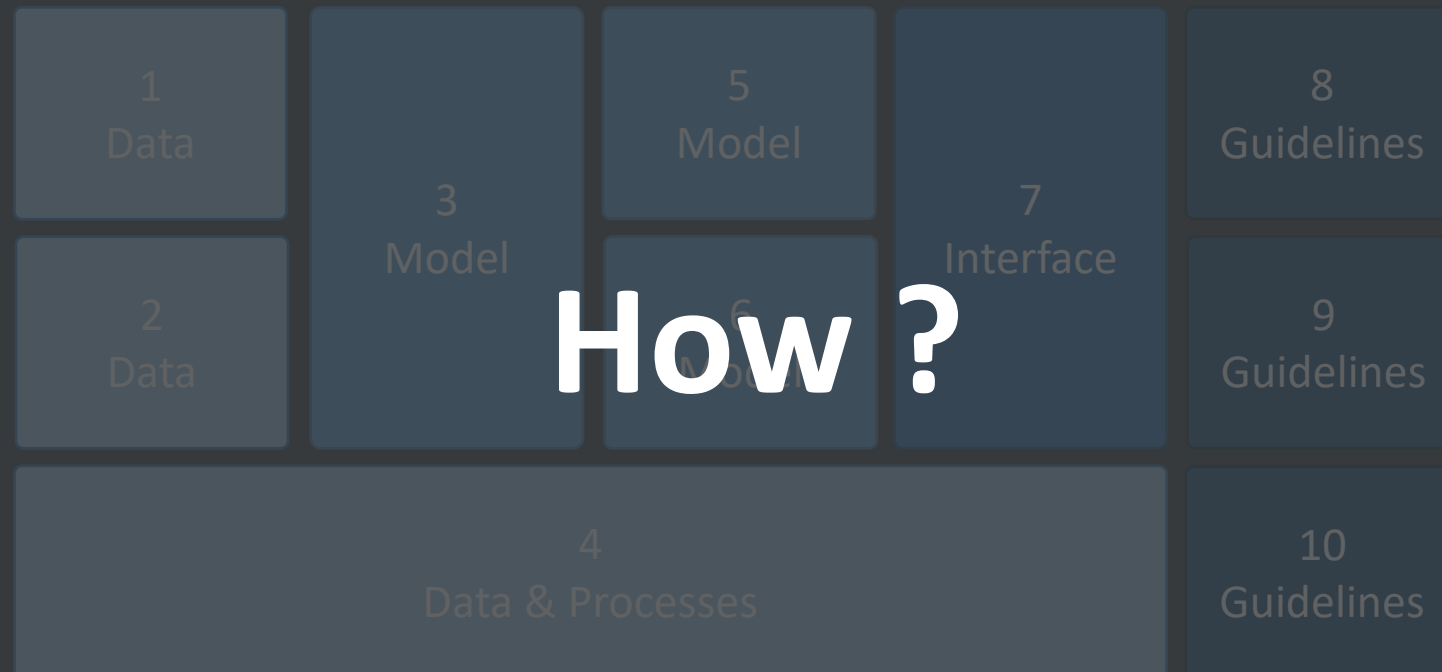
# Economical Need

## Combination of Methods

## Complex & Dynamic Processes

## Increasing & Complex Data

Integrated Processes for **Modelling & Simulation**



Integrated Processes for **Modelling & Simulation**

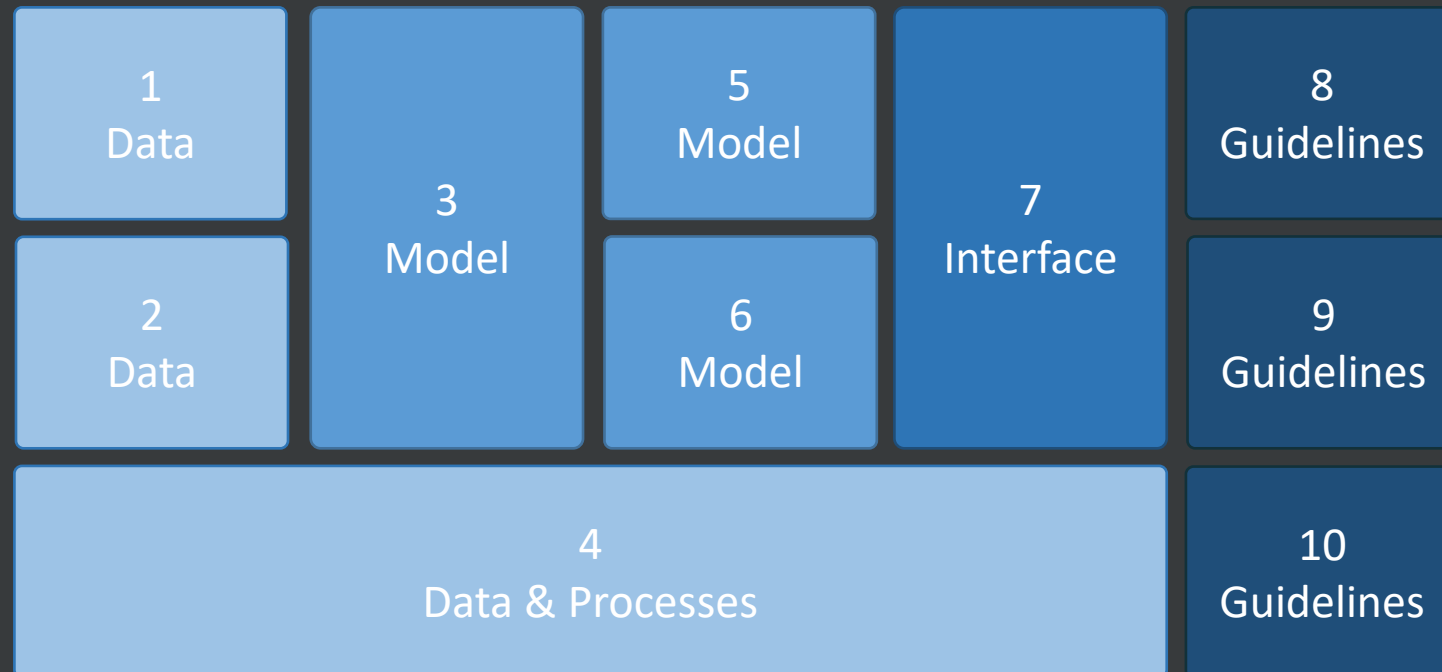
# Data Driven System Simulation

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)

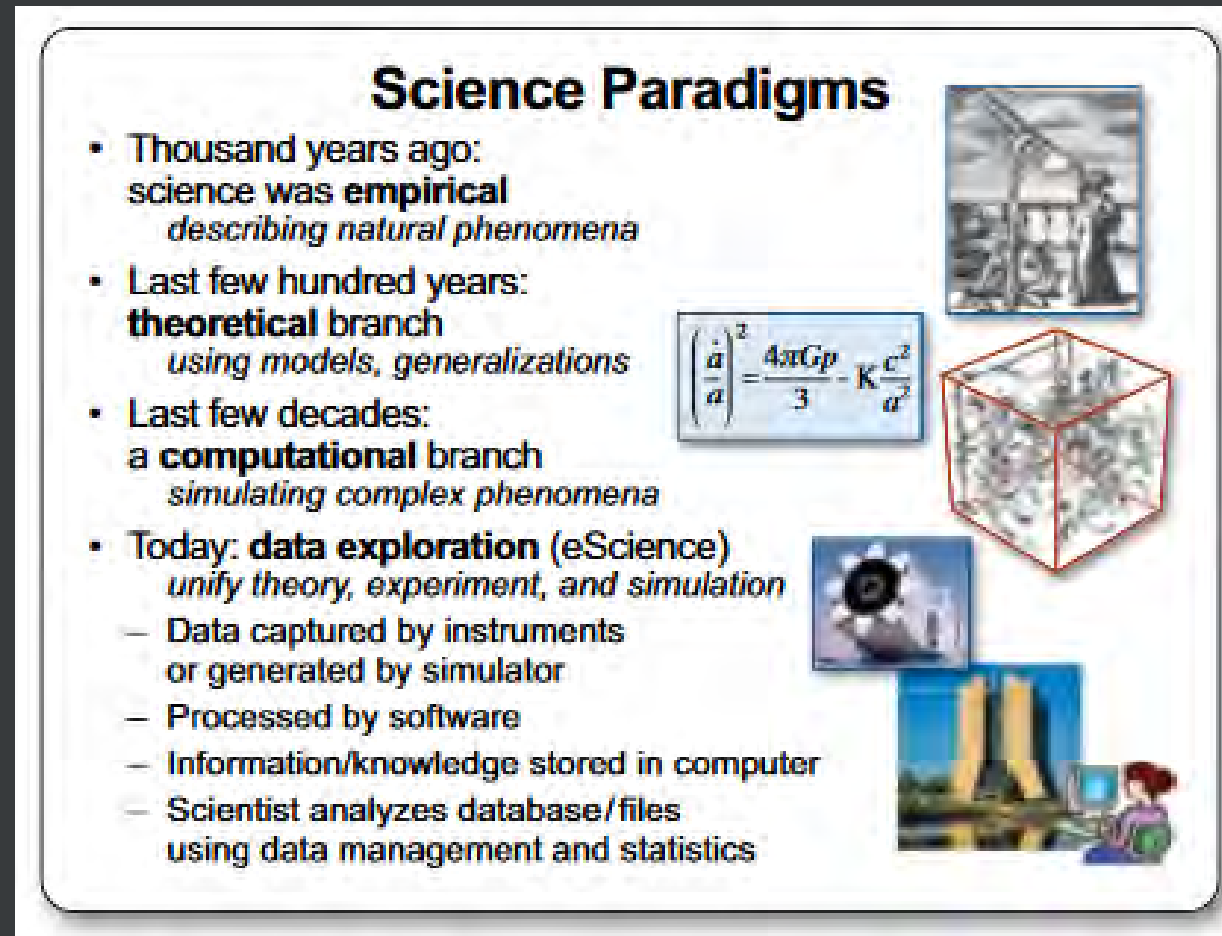
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### How?

- > 4<sup>th</sup> Paradigm
- > Modelling Dynamics
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- > Processes



Hey, T., Tansley, S. & Tolle, K. (eds.) (2009). *The Fourth Paradigm: Data-Intensive Scientific Discovery*. Redmond, Washington: Microsoft Research.

# Modelling Dynamics



dwh  
technical solutions  
simulation services



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WIEN

## 10 Concepts to Integrate

### Why?

- > Dynamics & Complexity
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# #

Which **Questions**  
are to be answered?

# #

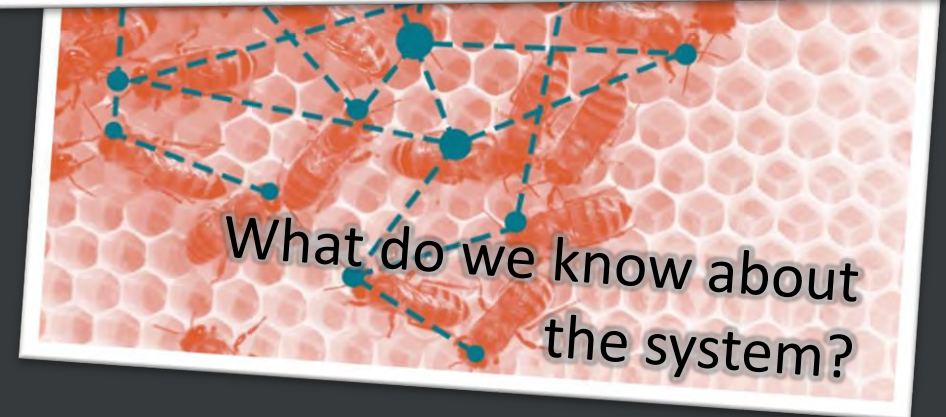
Which **Data**  
**Ressources** are  
available?

# #

Which **Systems &**  
**Processes** are  
described ?



What data is available?



What do we know about  
the system?

N. Popper: "Comparative Modelling and Simulation, A Concept for Modular Modelling and Hybrid Simulation of Complex Systems", PhD Thesis

# Modelling Dynamics

## 10 Concepts to Integrate

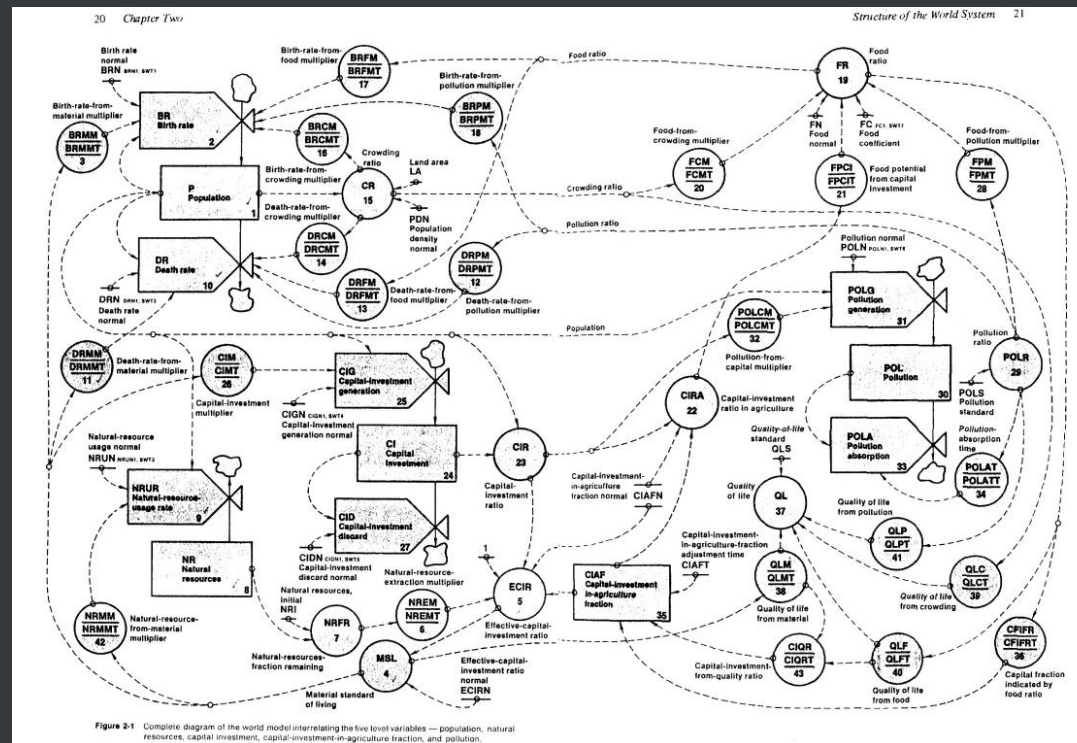
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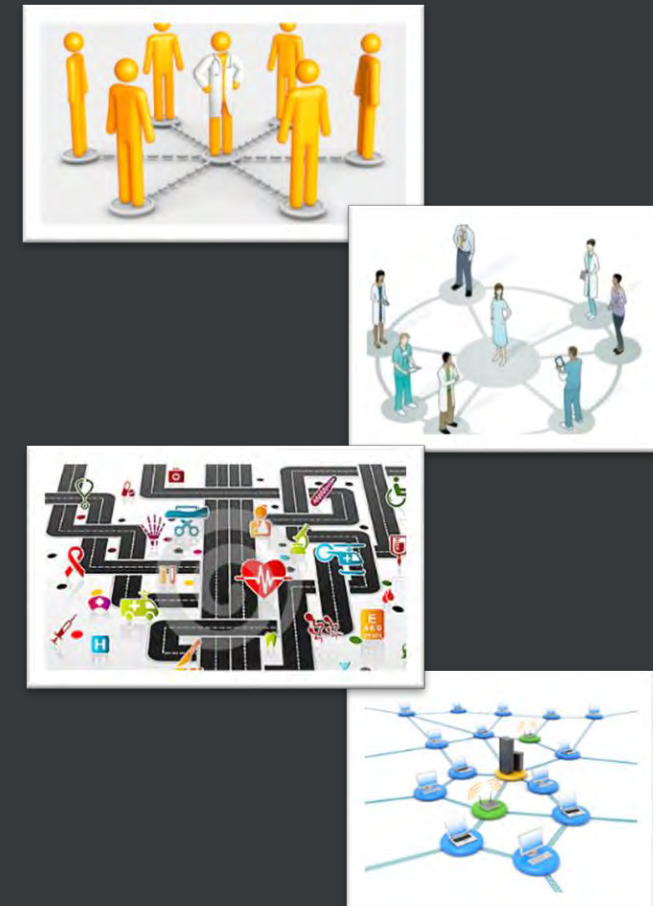
### How?

- > 4<sup>th</sup> 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

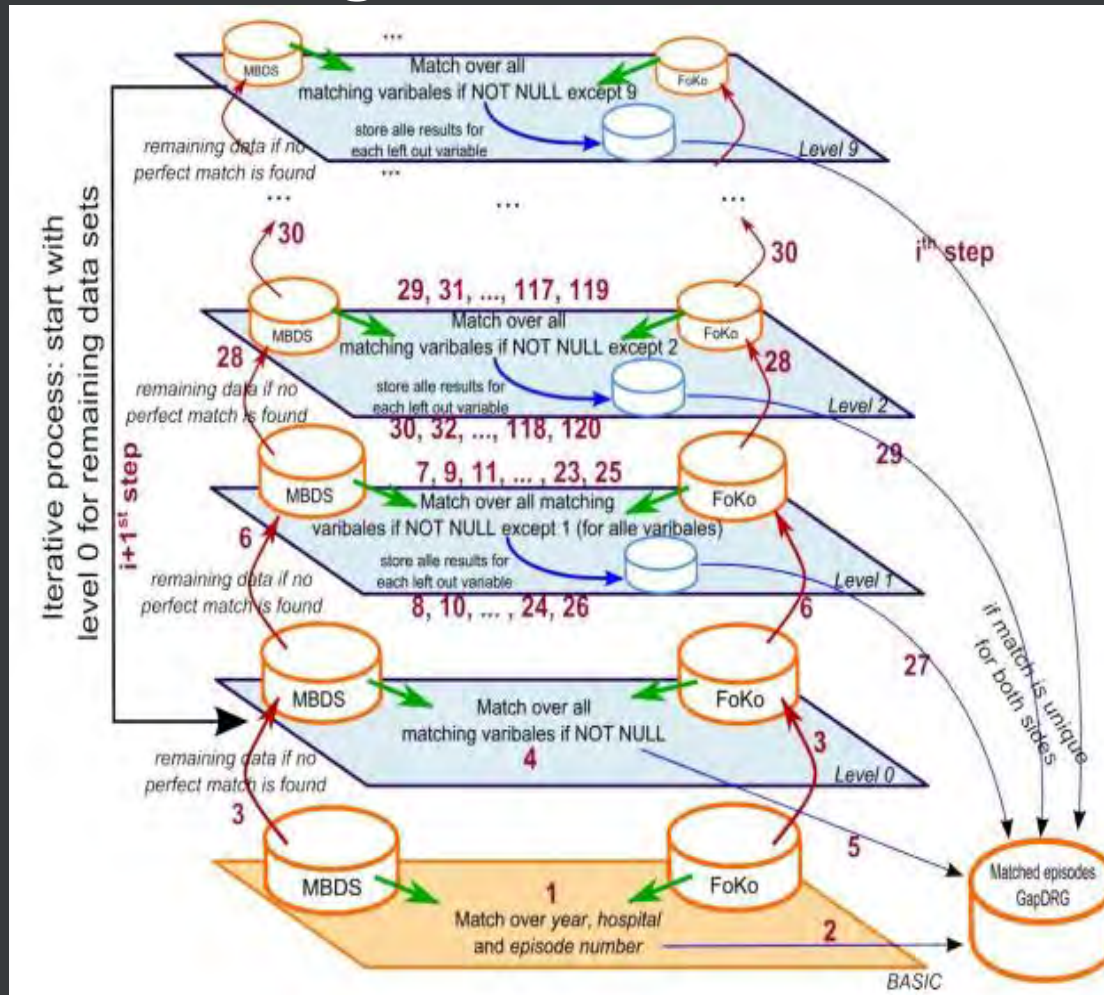
### Why?

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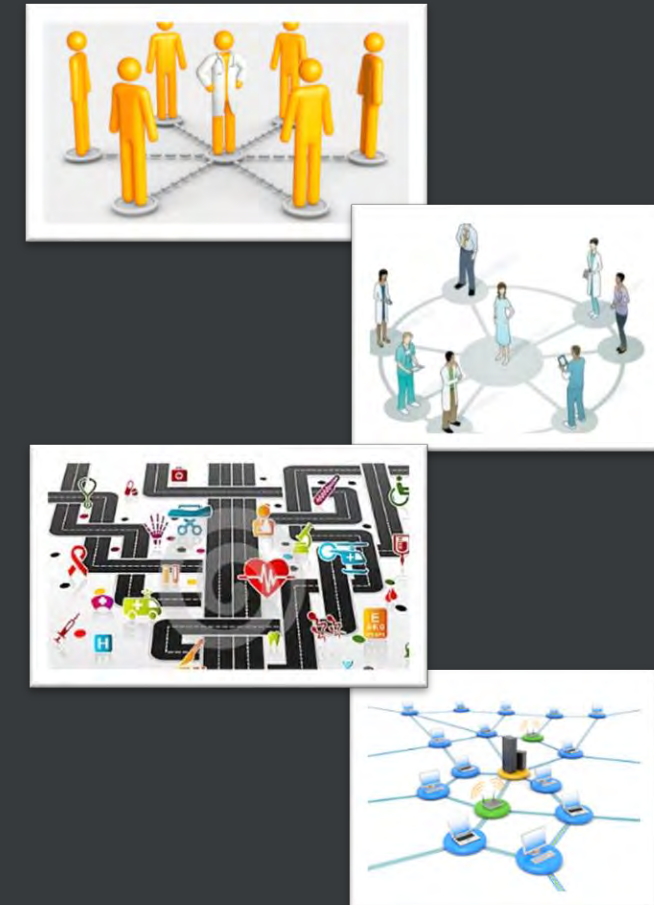
### How?

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

## Big Data Sets



## Structures & Knowledge



# Modelling and Simulation Processes

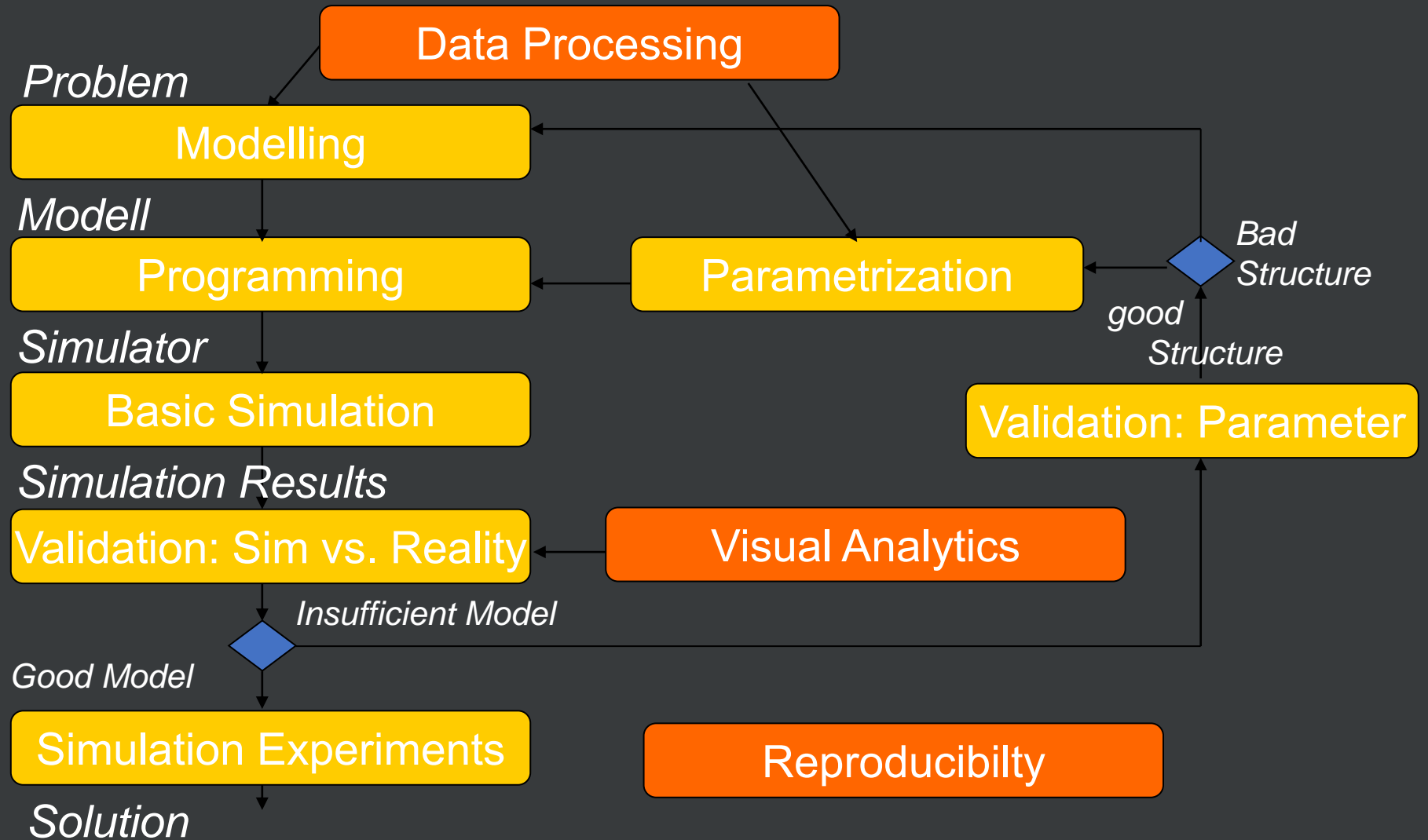
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- > Big Data

### How?

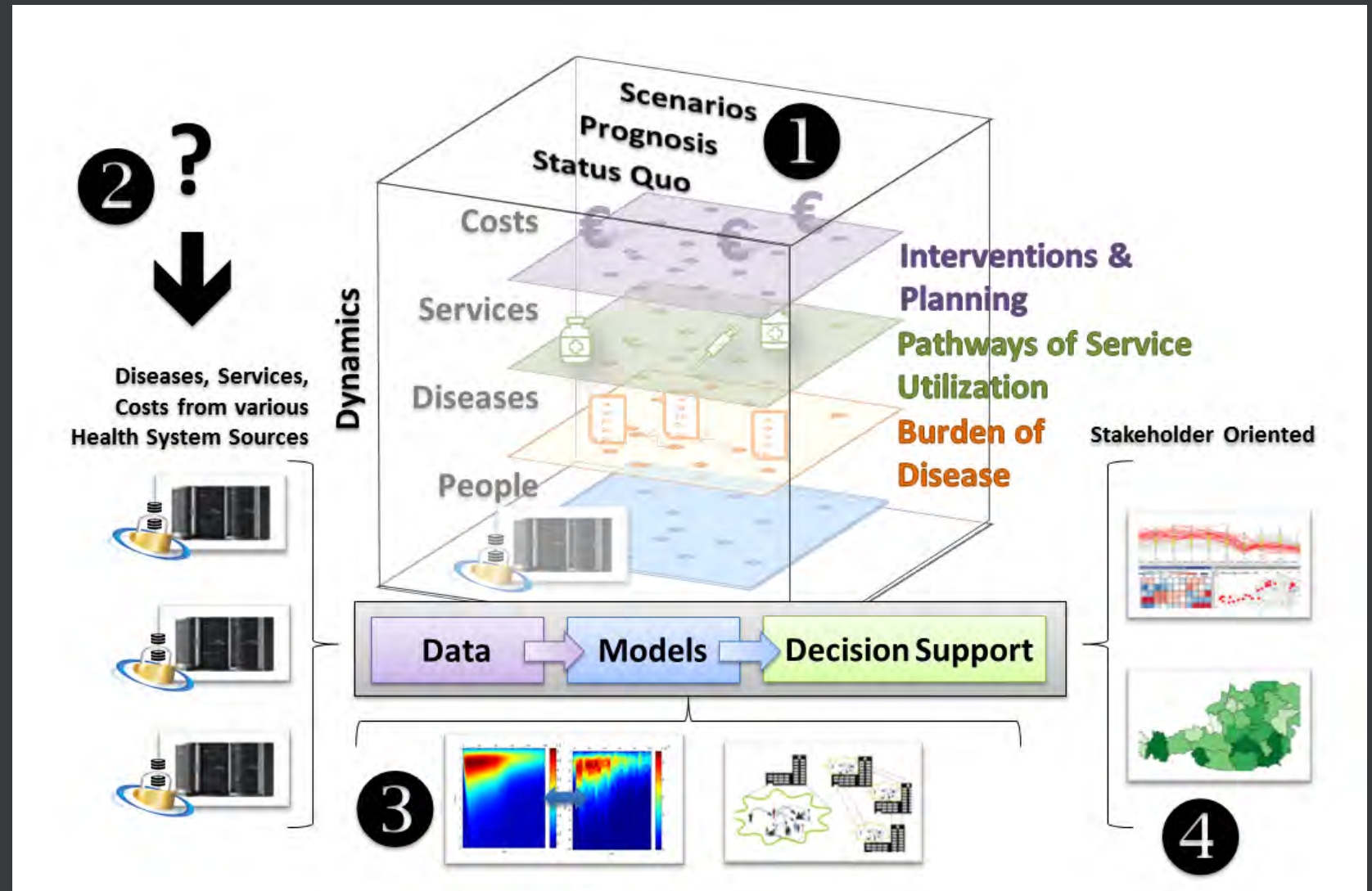
- > 4<sup>th</sup> Paradigm
- > Modelling Dynamics
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# Example Health System: DEXHELPP

## DEXHELPP

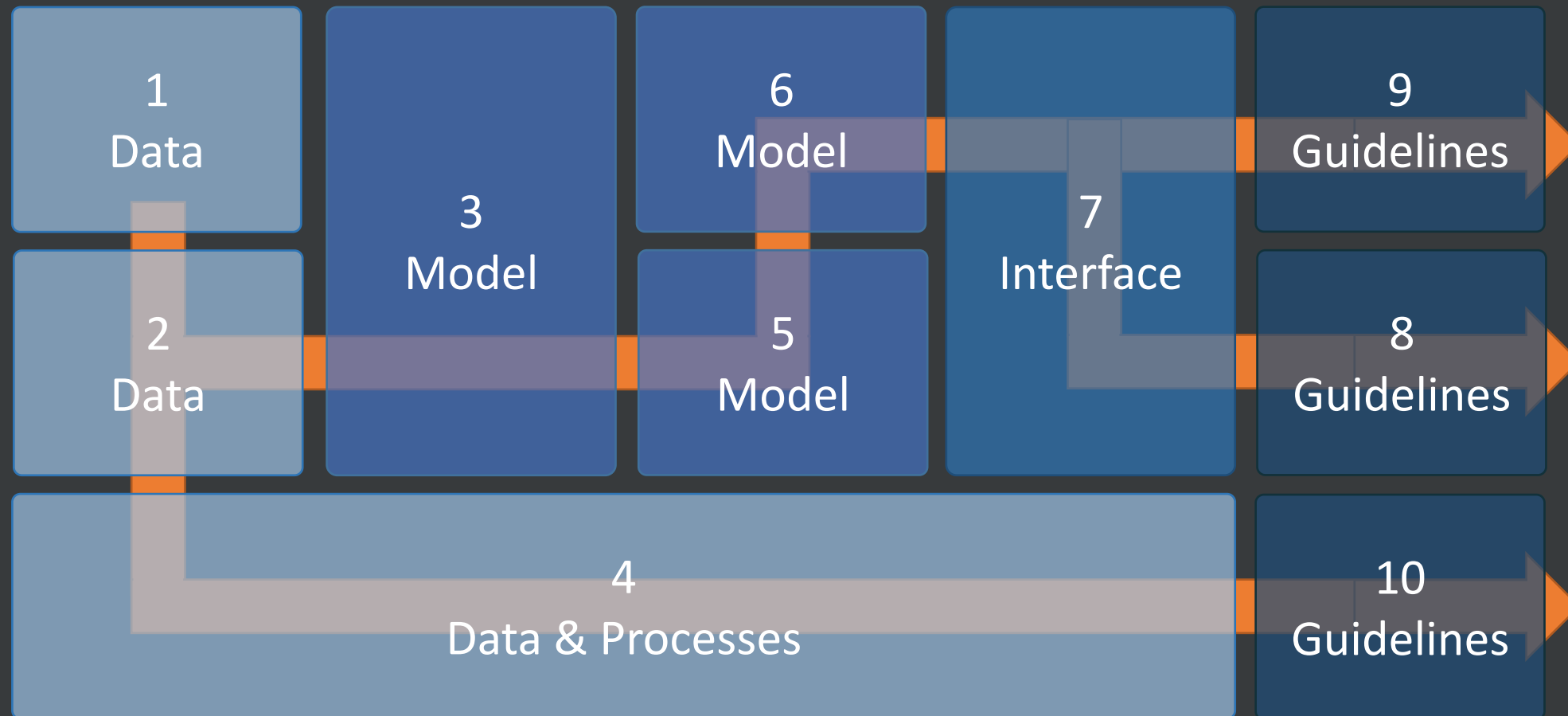
Decision Support for Health  
Policy and Planning: Methods,  
Models and Technologies based  
on existing health care data



# 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

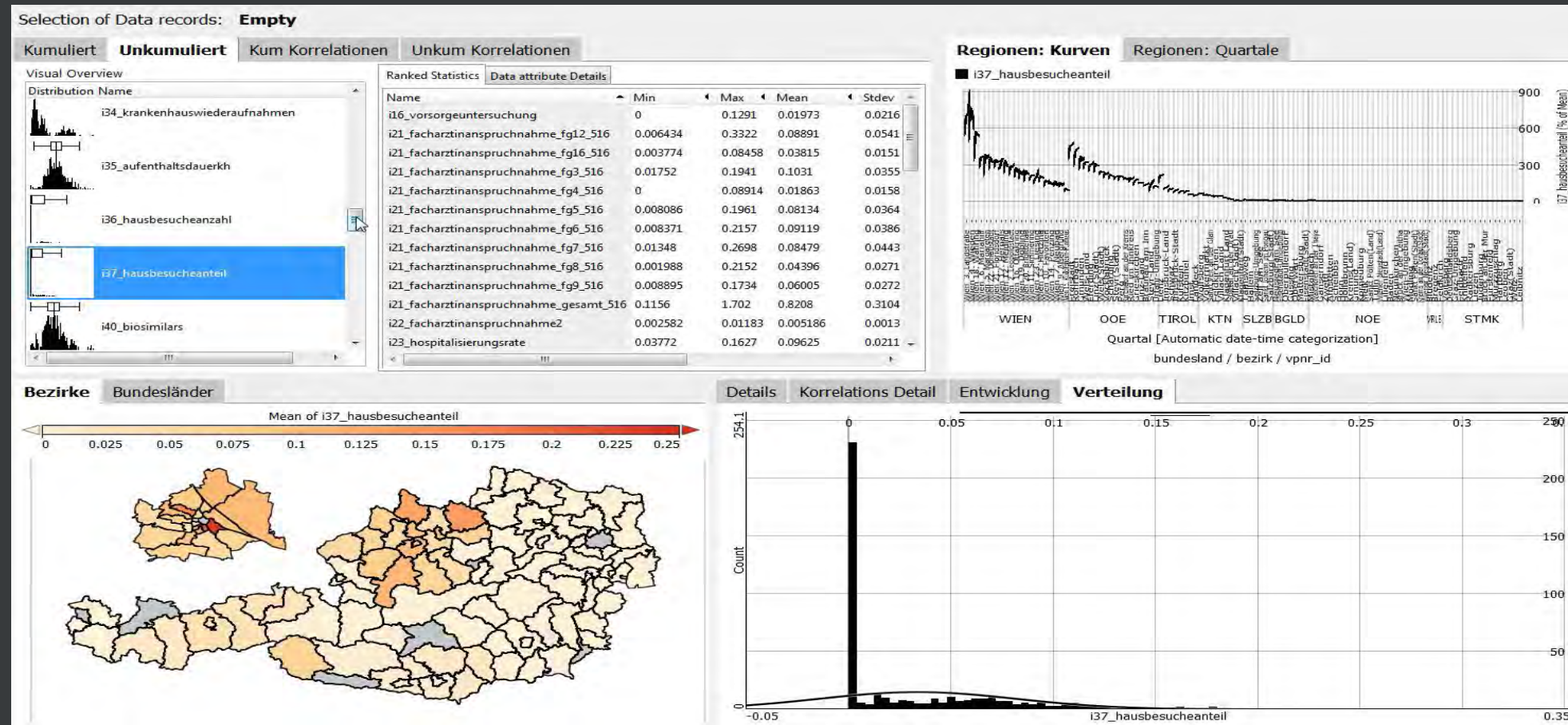
- Biased Collection of Data e.g. Sensor Data or Reimbursement Data is available
- Pre-processing at Various Stakeholders
- Privacy Demands

**METHOD: Explorative Visual Computing – Visual Analytics and Statistics**

# Interactive Dashboards

## 10 Concepts to Integrate

### 1. Methods to Assess and Improve Quality of Data



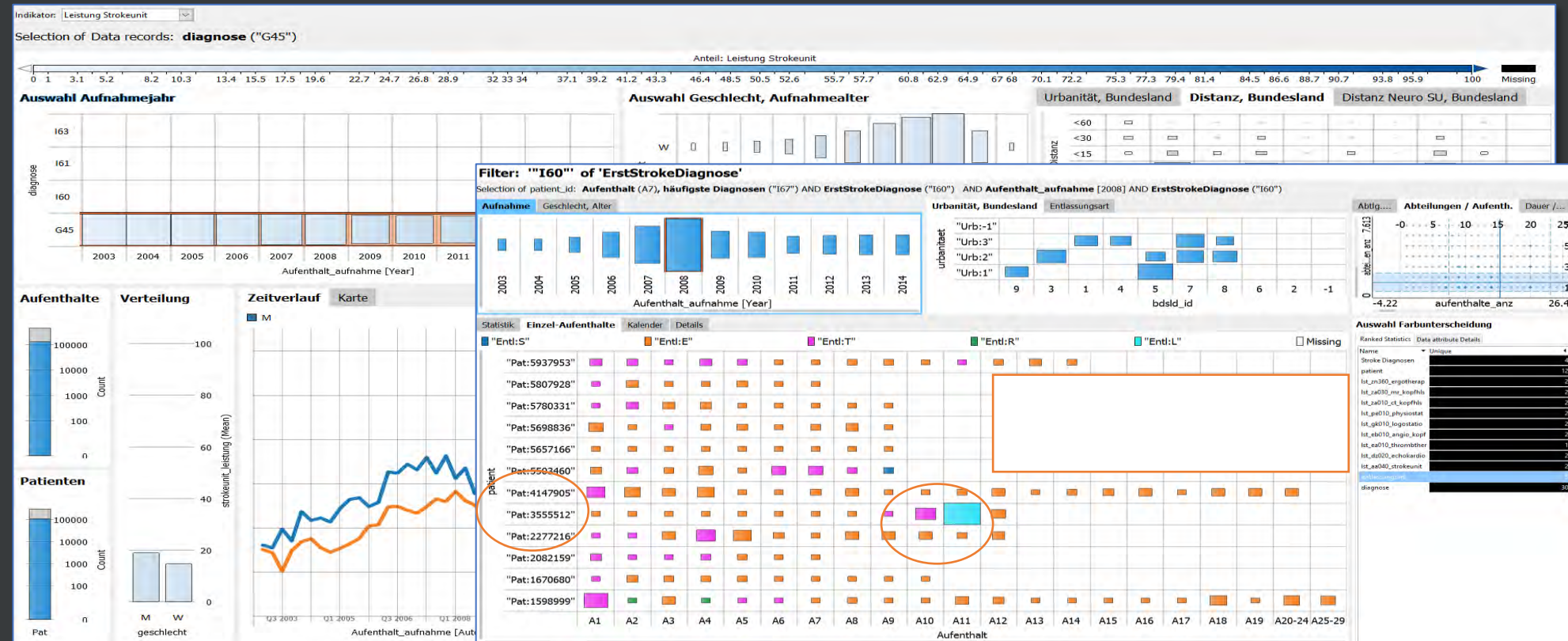
# Data on Stroke Treatment

## 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

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

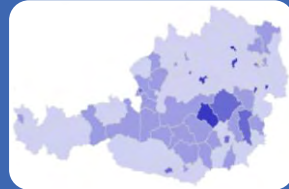
**METHOD: Data Processes (Integration & Linkage) & Modelling Tools (Parametrization & Calibration)**



# 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



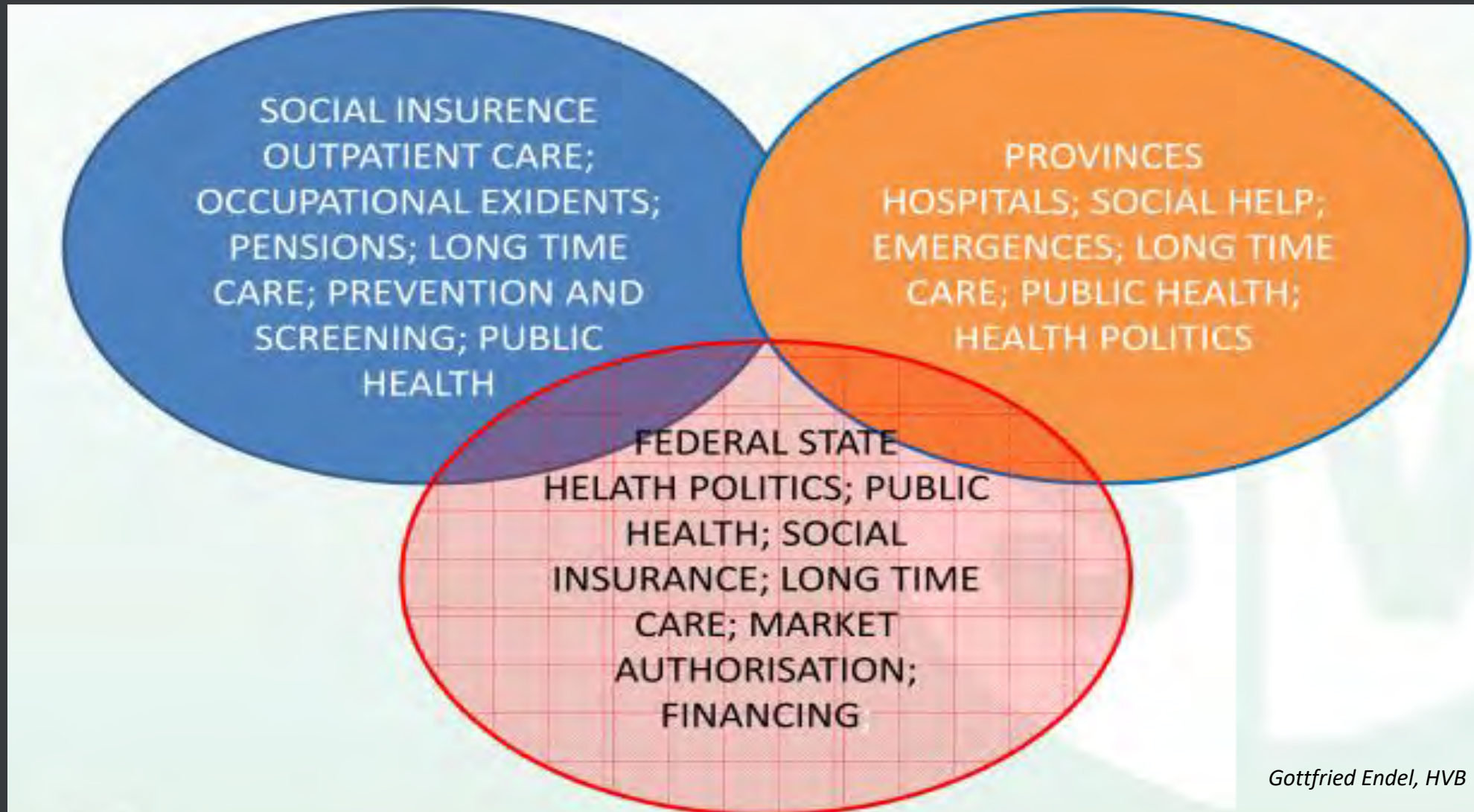
dwh  
technical solutions  
simulation services



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WIEN

### 10 Concepts to Integrate

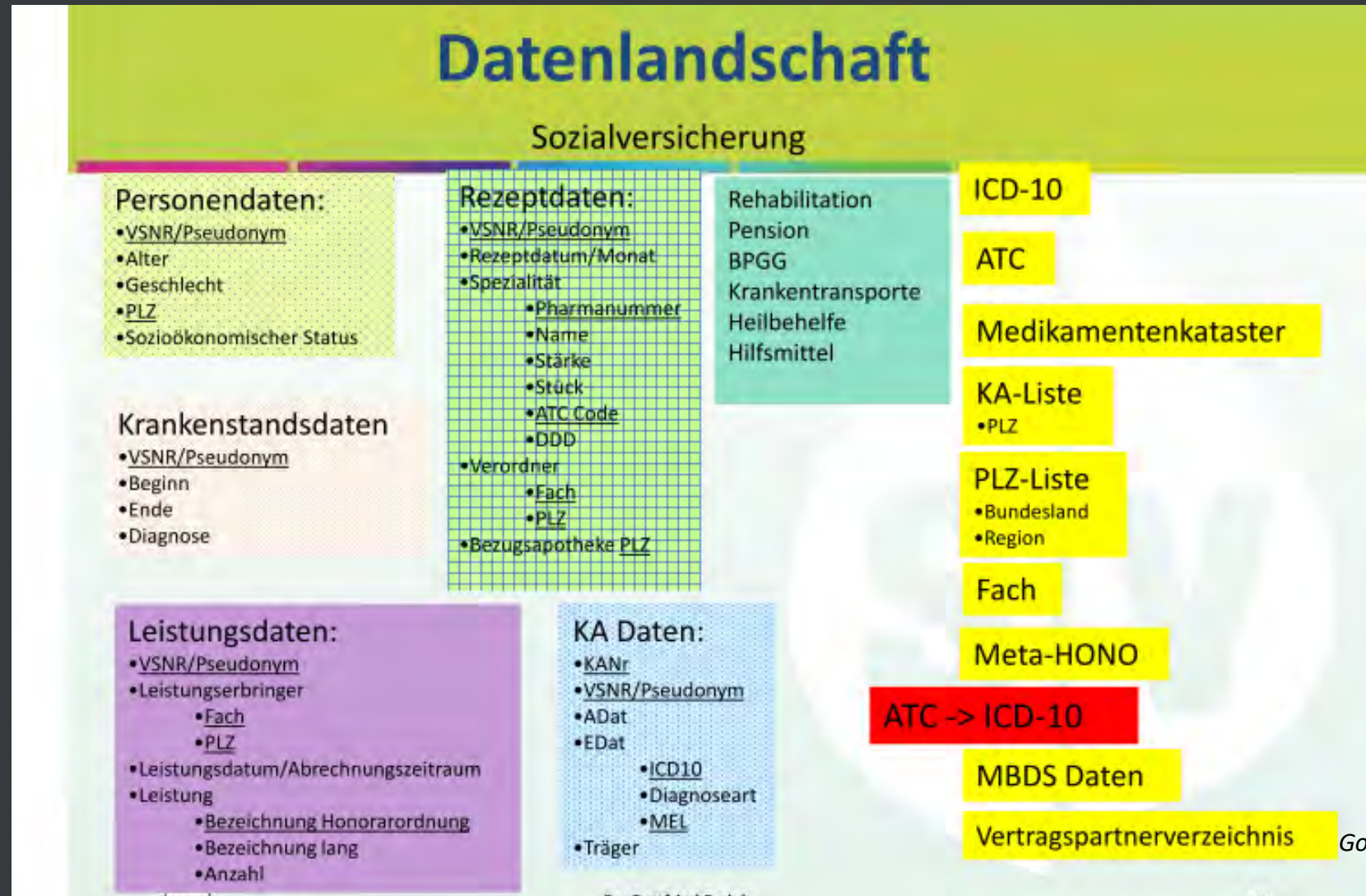
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data



# Data Social Insurancess

## 10 Concepts to Integrate

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

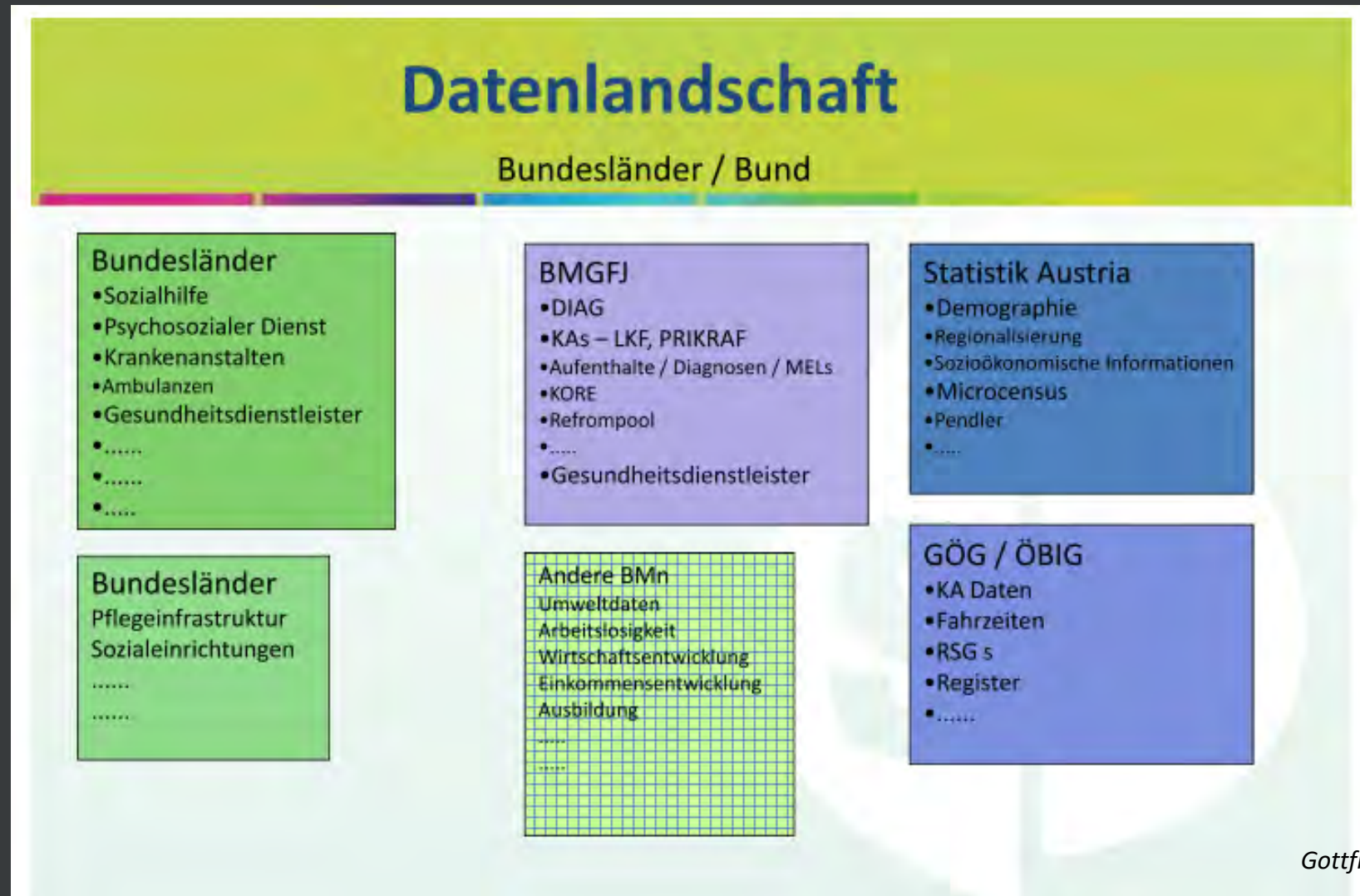




# Data Provinces & Austria

## 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

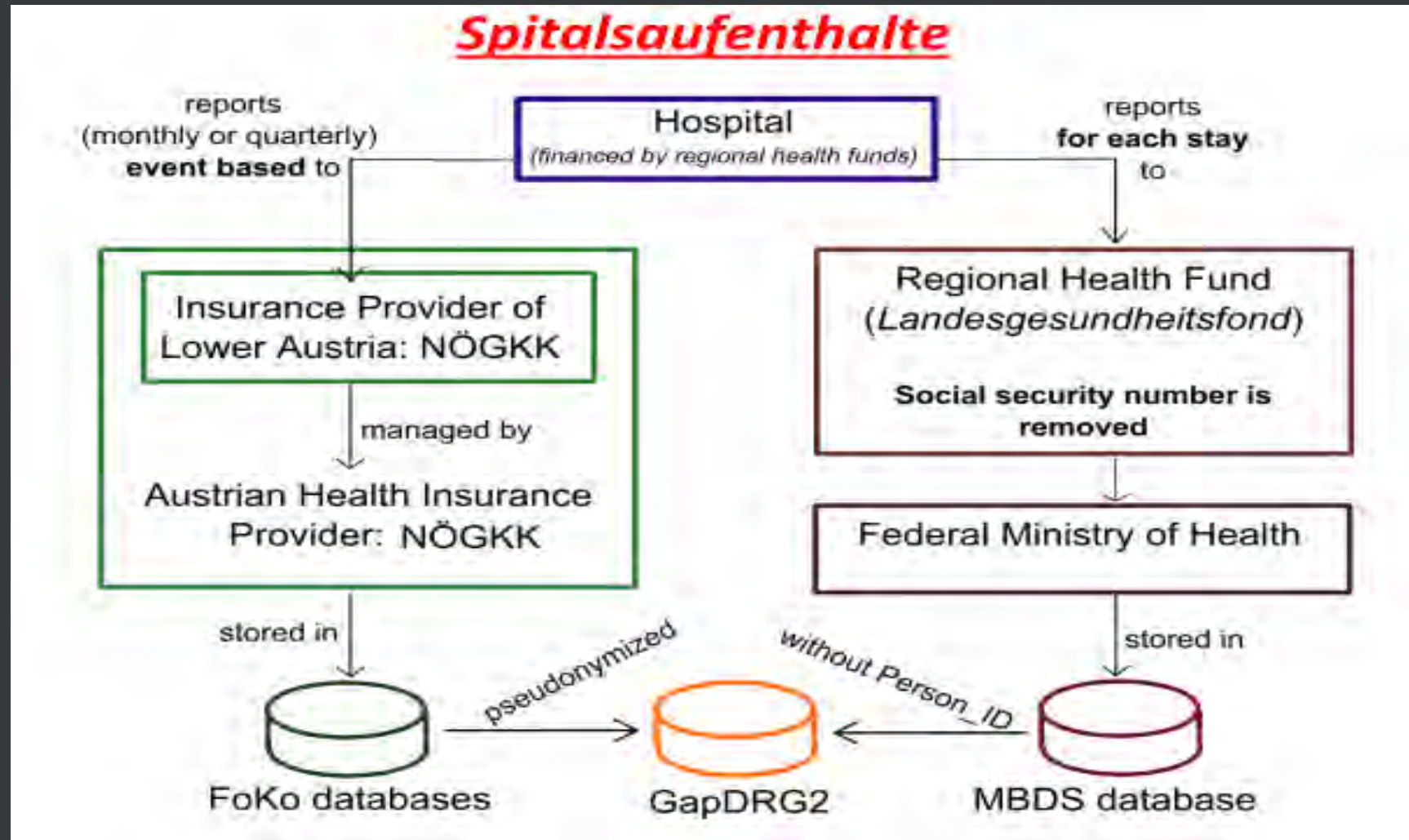




# Example Record Linkage

## 10 Concepts to Integrate

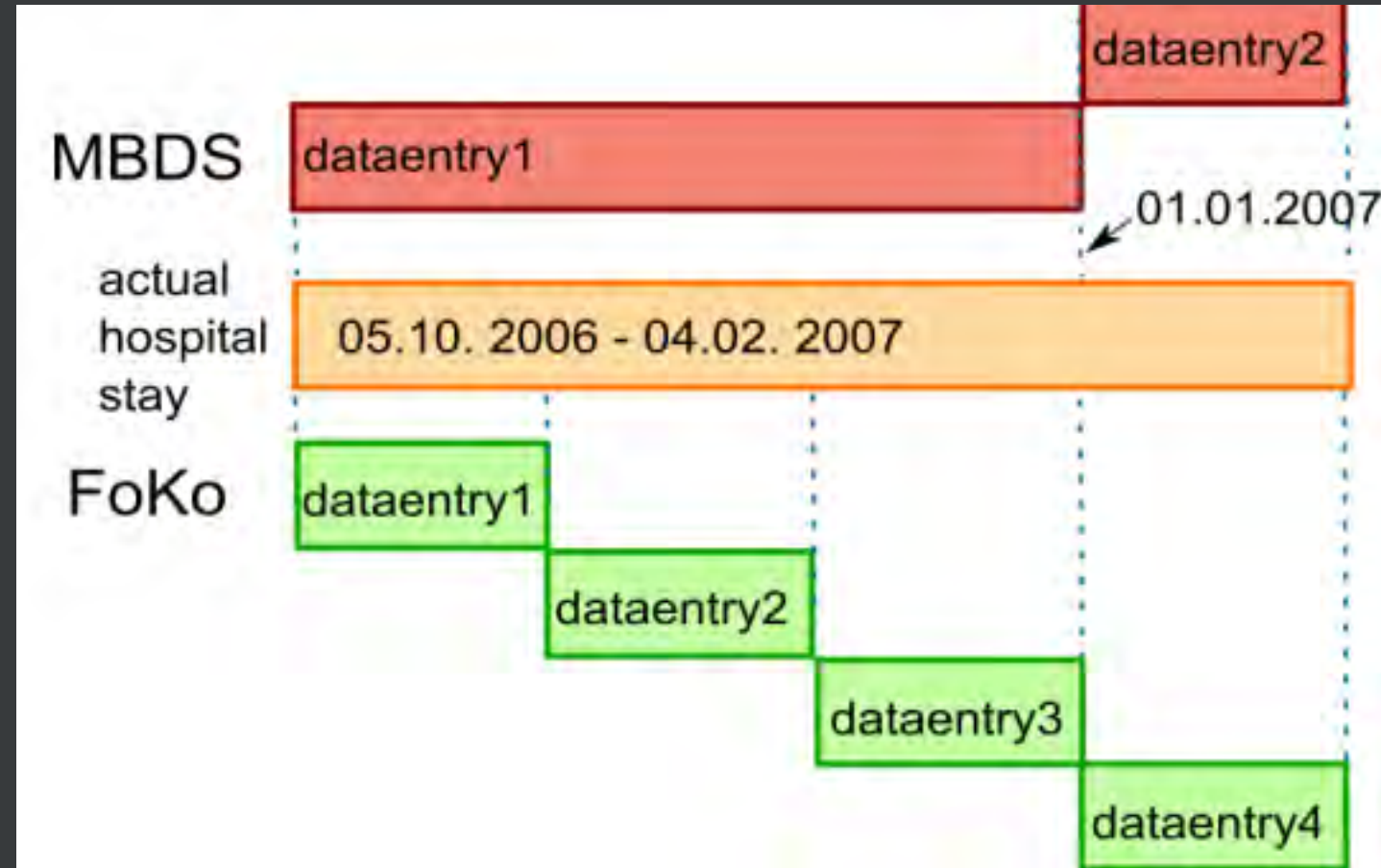
1. Methods to Assess and Improve Quality of Data
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# Example Record Linkage

## 10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
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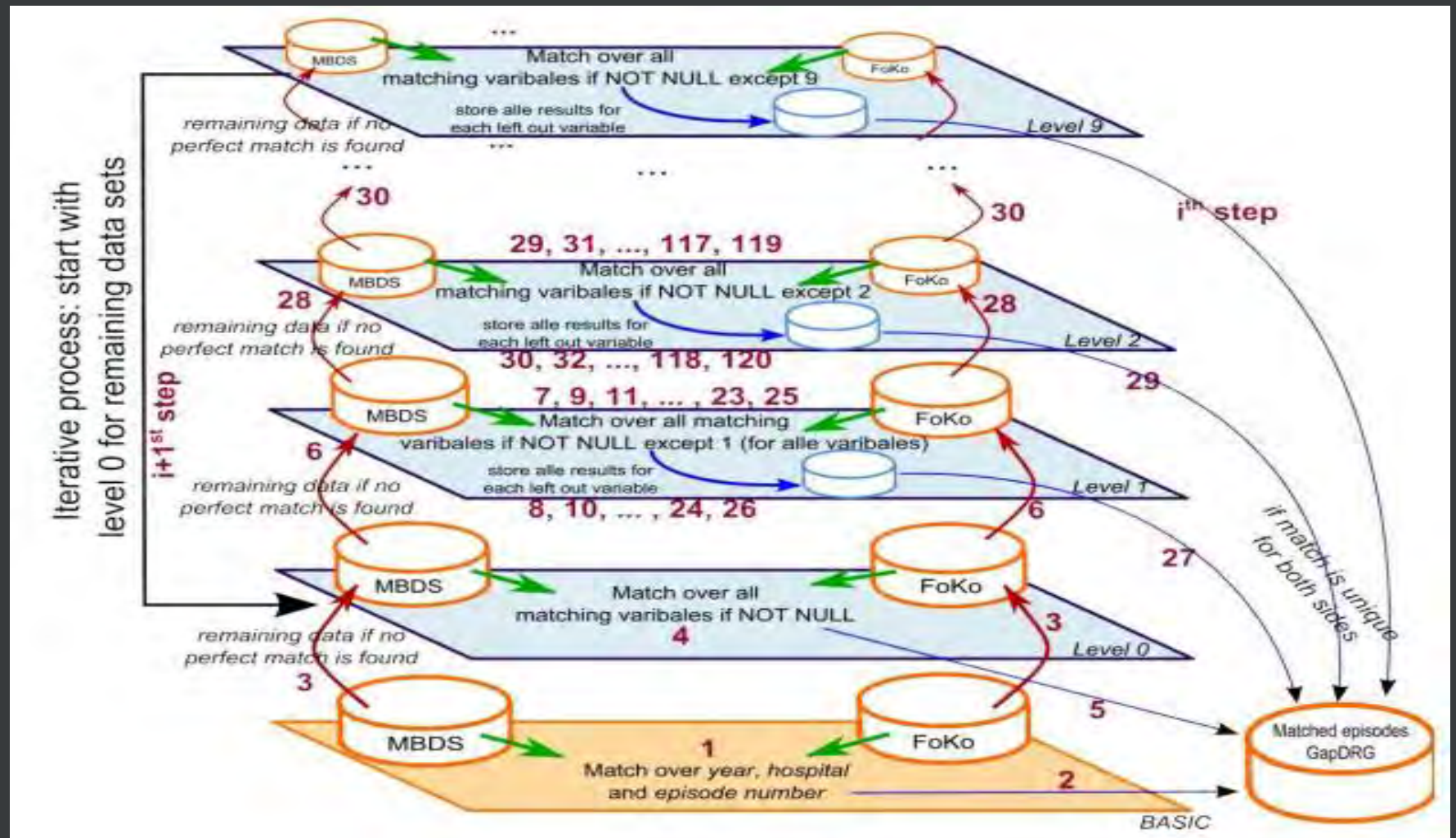


*B. Glock, F. Endel et al: Challenges and Results with the Record Linkage of Austrian Health Insurance Data of Different Sources, Informatics for Health Conference 2017 (24 – 26. April, Manchester, UK)*

# 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

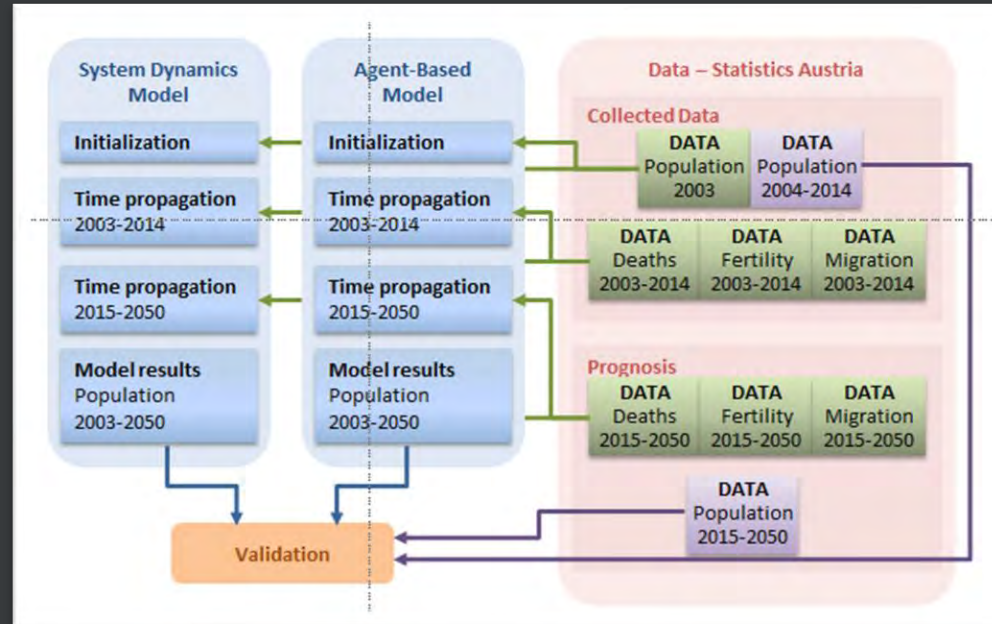
**METHOD: Modular Models,  
Coupling of Models**



# 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

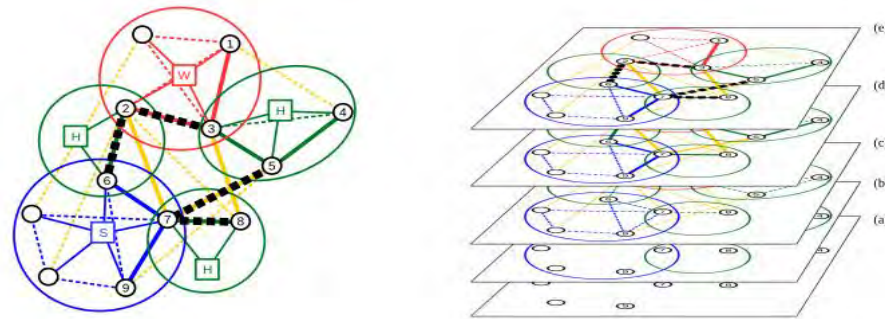


Figure 1: Visualization of the basic network layout. Households (H), workplaces (W) and school classes (S) and their memberships are depicted as green, red and blue blocks (b) of nodes (a). Social ties are indicated by dashed lines (c), derived contact patterns are shown as thick lines (d). Inter-block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

Interaction & Networks

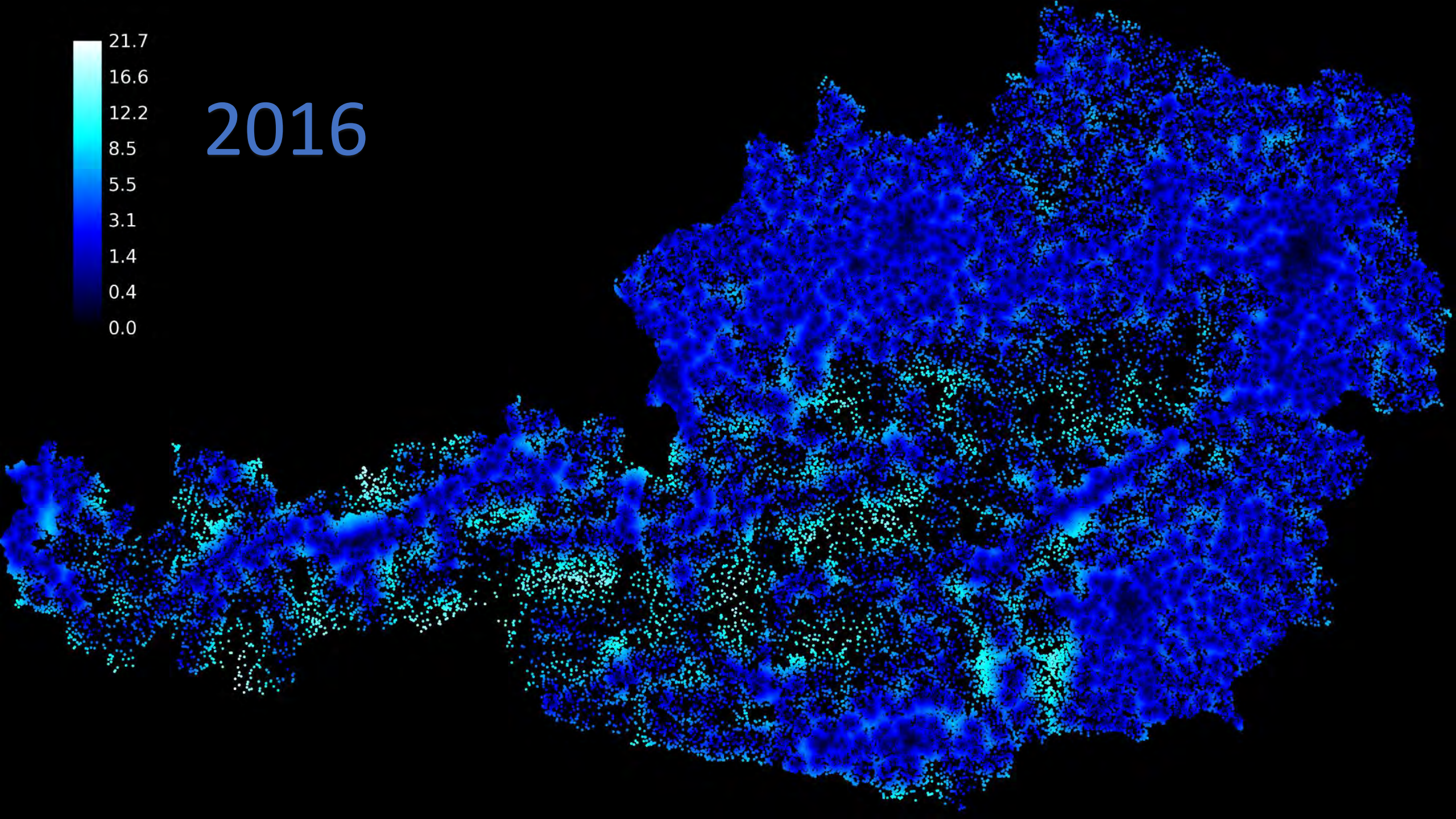
Runtime, Parallelisation

Parameter-Calibration

Time Enhancement

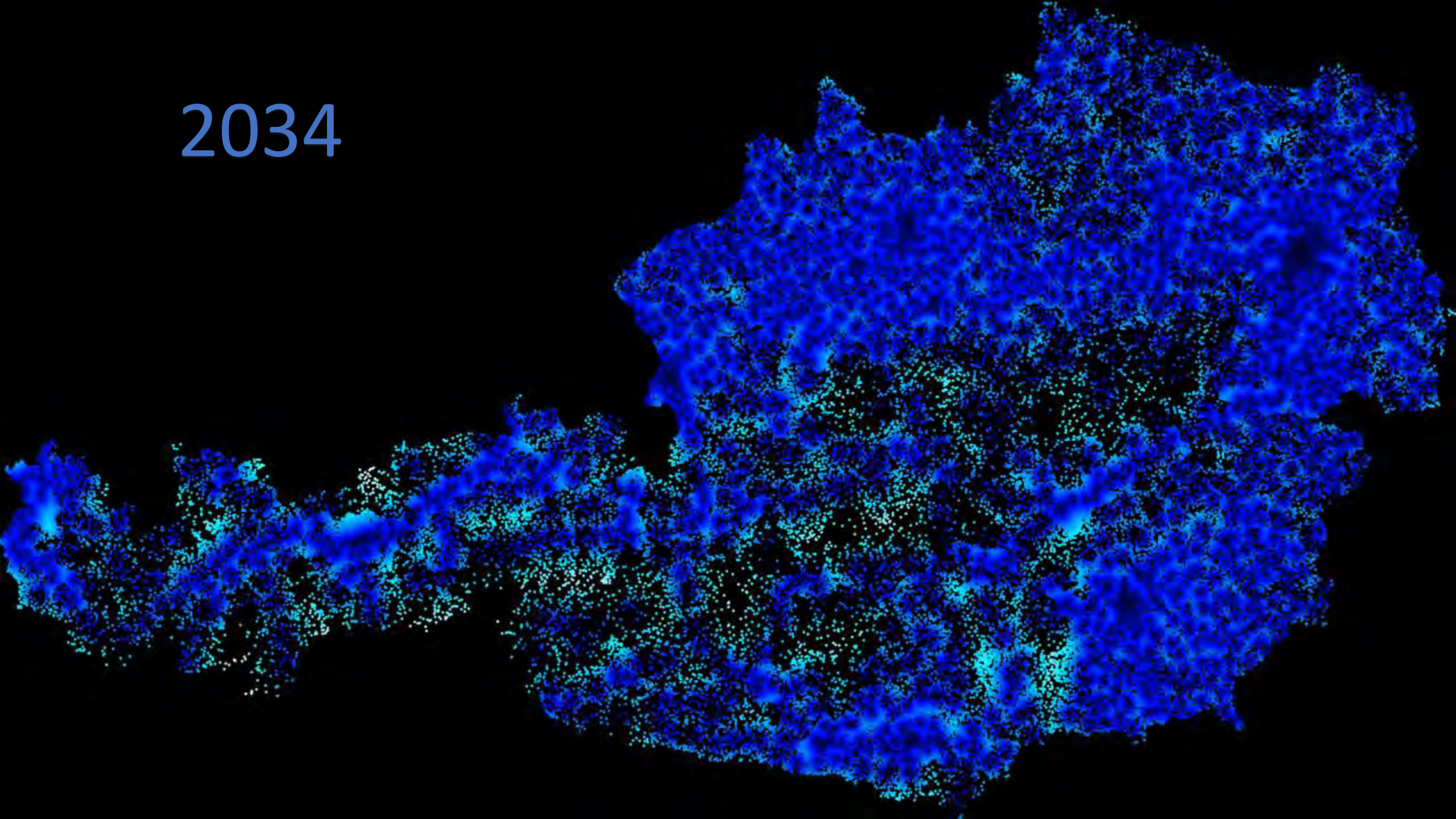
Internal Migration and Spatial Relationships







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

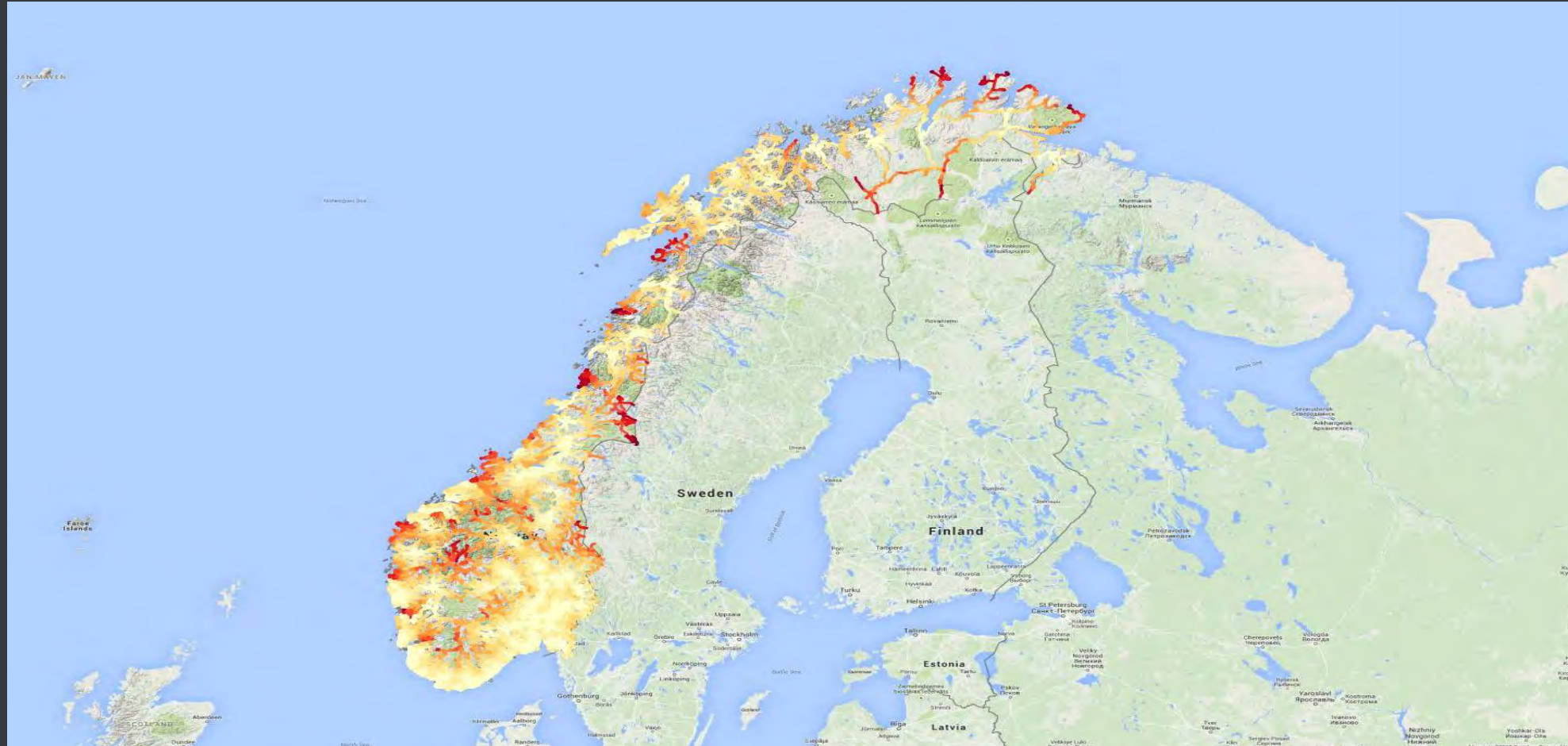
CEPHOS-LINK is a comparative European register-based study, performed by record linkage technique. It will develop a methods toolkit for conducting record linkage studies in the mental health care field and produce recommendations, guidelines and a set of decision support tools for decision makers in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



# Best Practice CEPHOS

## 10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
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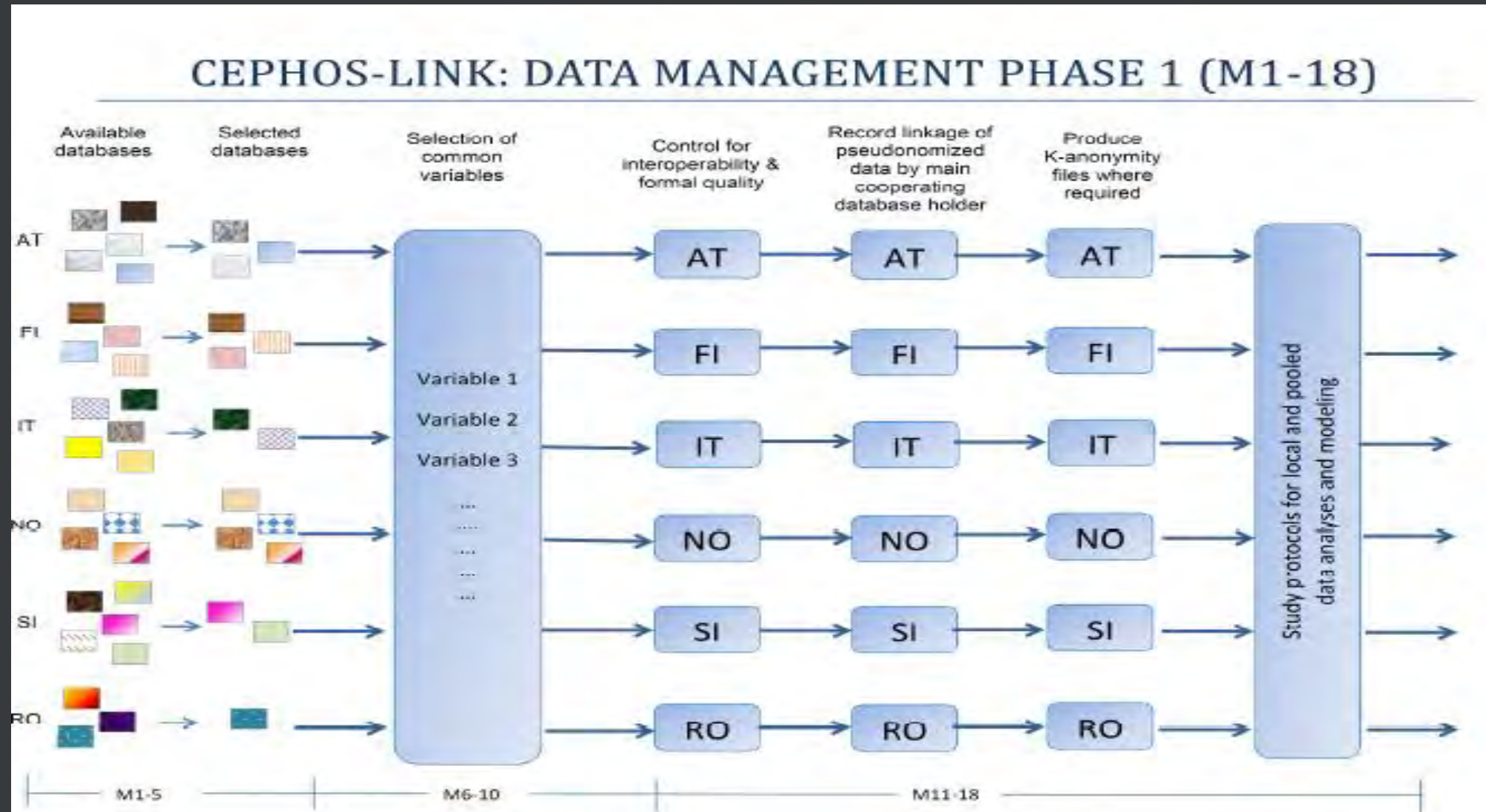


*C. Urach, G. Zauner et al "Statistical methods and modelling techniques for analysing hospital readmission of discharged psychiatric patients: a systematic literature review"; BMC Psychiatry (2016)*

# Best Practice CEPHOS

## 10 Concepts to Integrate

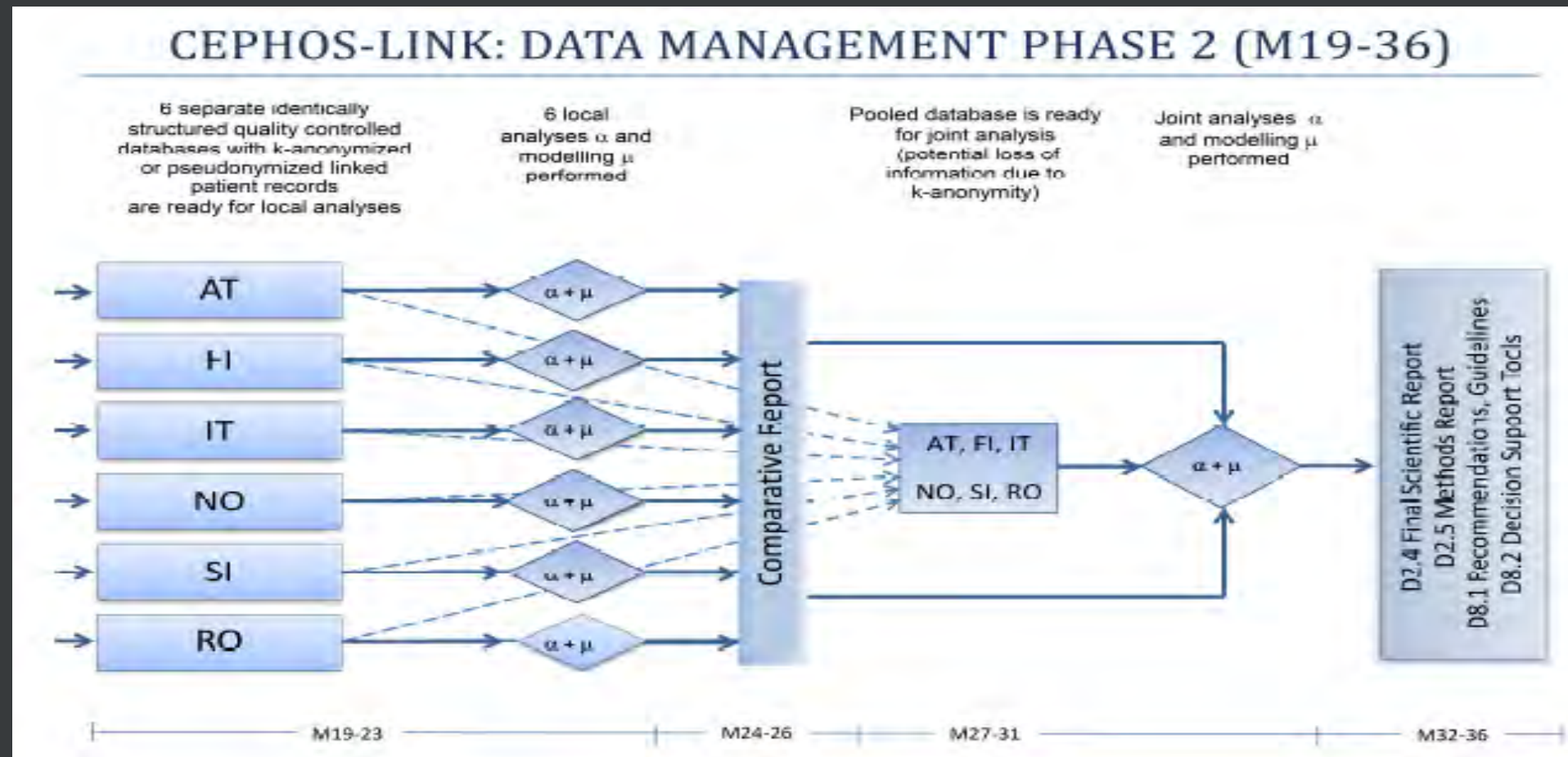
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# Best Practice CEPHOS

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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

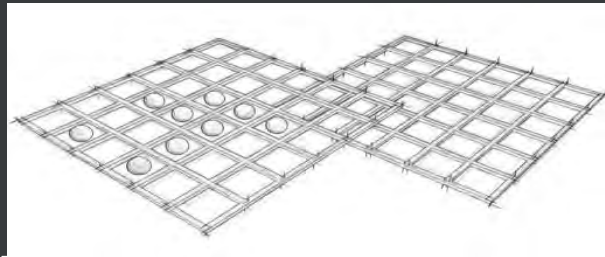
## 10 Concepts to Integrate

1. Methods to Assess and Improve Quality of Data
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3. Modular & Efficient Solutions
4. Reproducible Processes
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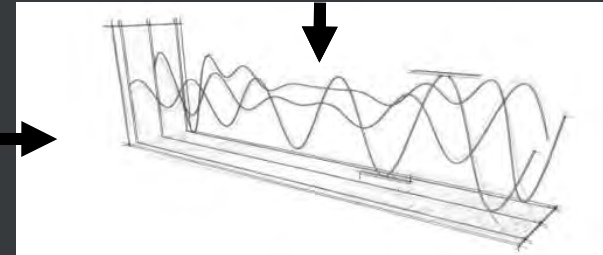
## Modelling of infectious diseases: Pneumococcal modelling systems



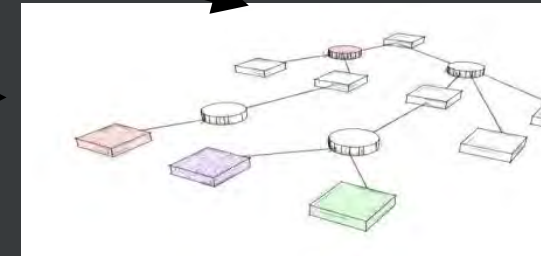
Real world



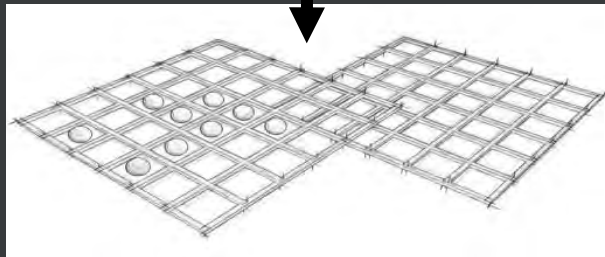
AB with const. population



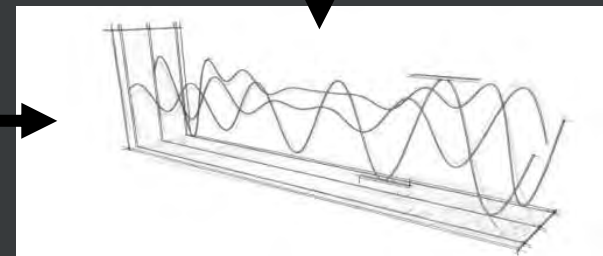
ODE with const. population



Markov Model  
- Decision tree  
- Statistics



AB with populationdynamics

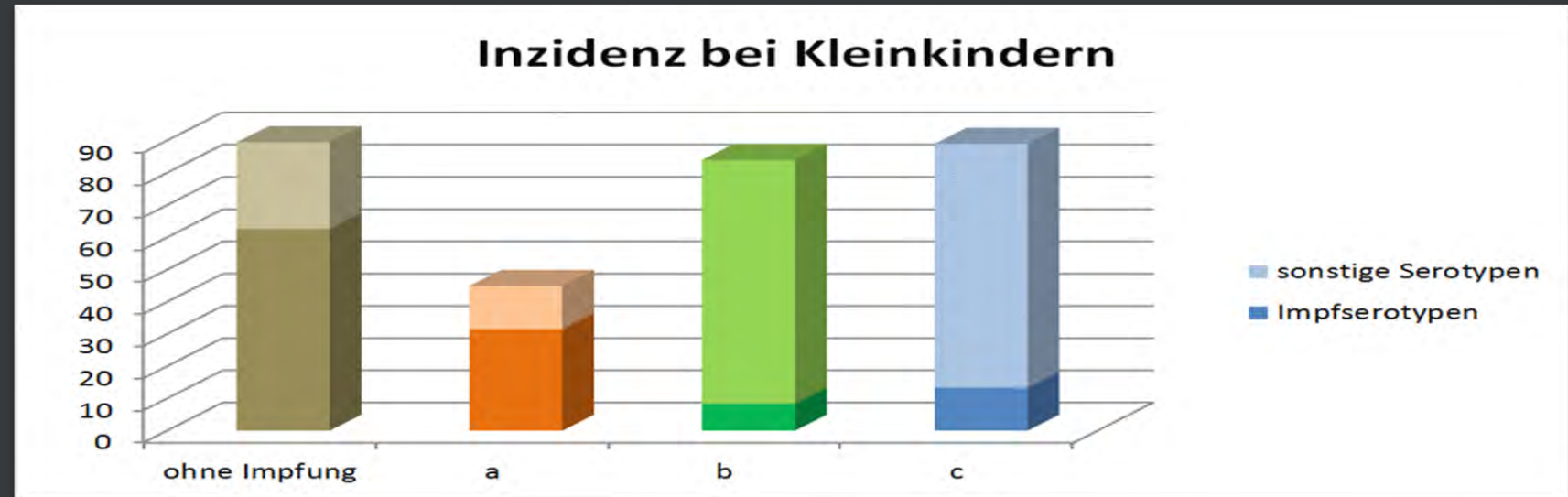


ODE with populationdynamics

# Example Comparison Vaccination

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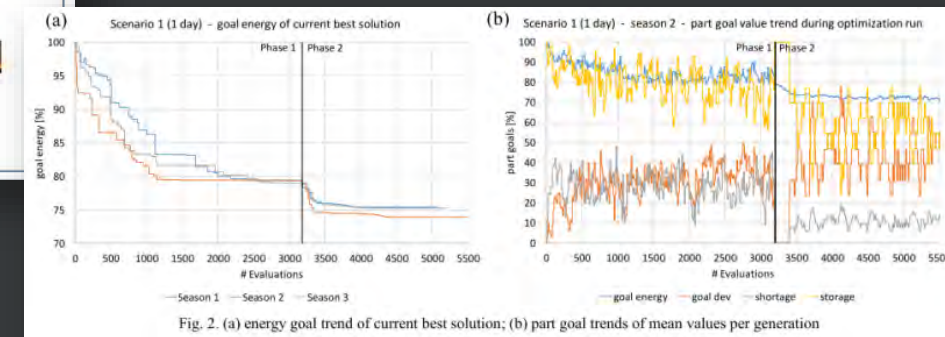
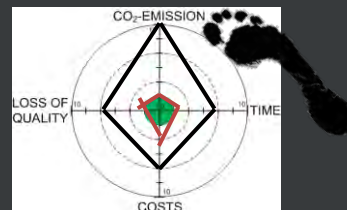
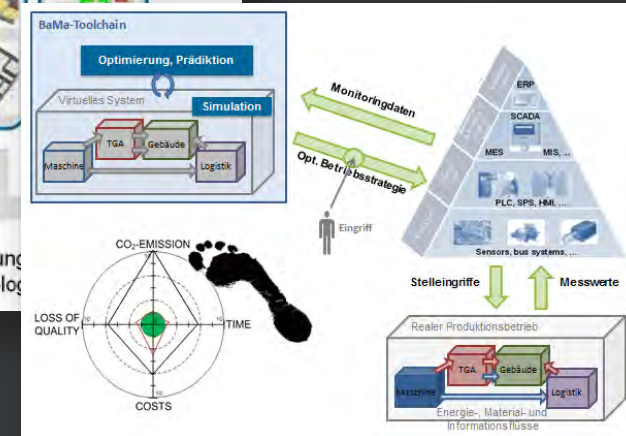
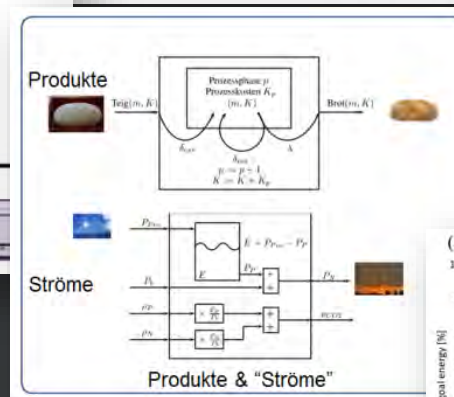
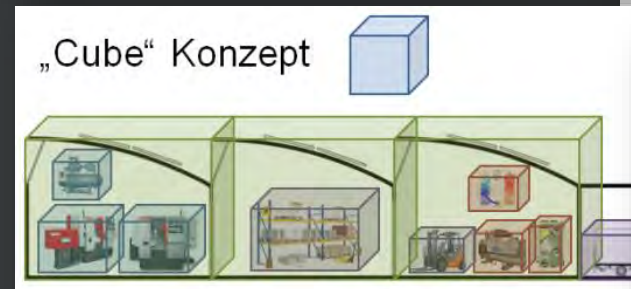
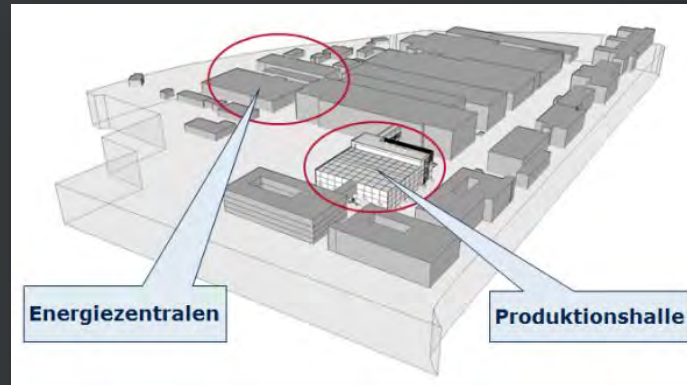


- a) **Gebräuchliche Prognosemodelle (Markov-Modelle, auf Österreich umgerechnet):**  
E.D.G. McIntosh, P. Conway, J. Willingham, R. Hollingsworth, and A. Lloyd. The cost-burden of paediatric pneumococcal disease in the UK and the potential cost-effectiveness of prevention using 7-valent pneumococcal 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

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2. Potential to Integrate Missing Data
3. Modular & Efficient Solutions
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)

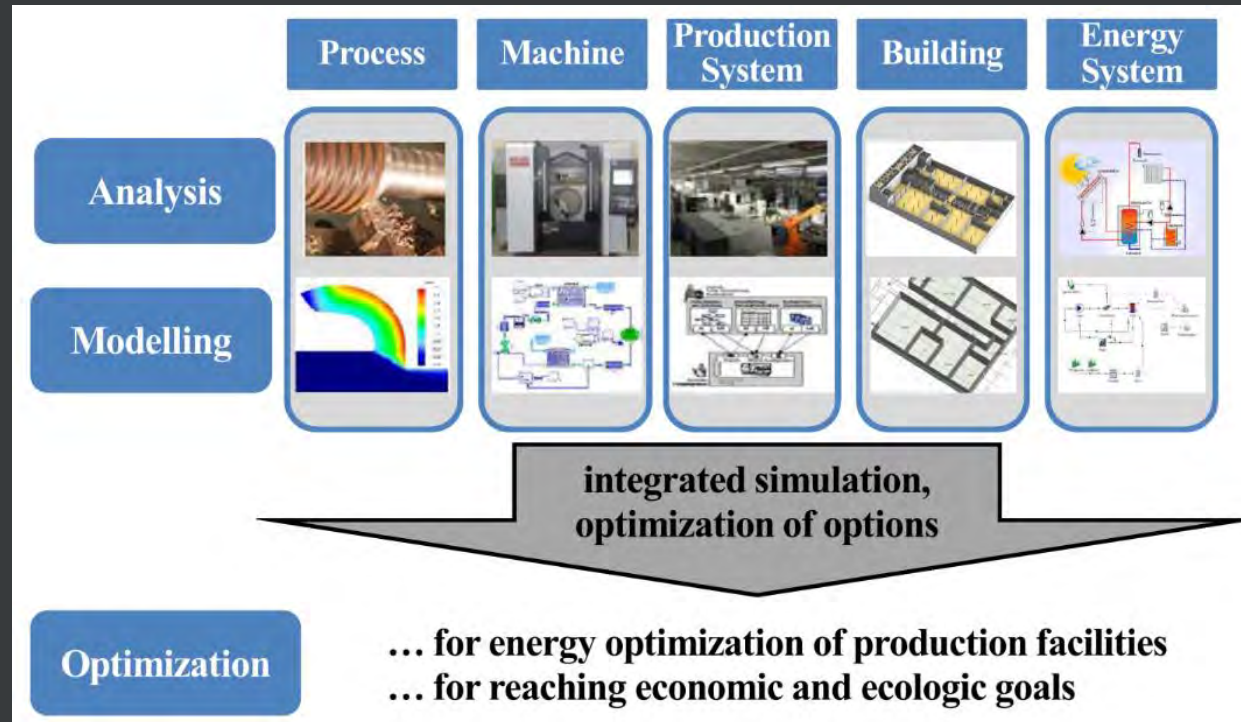




# Example Combination Balanced Manufacturing

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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)



- **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



dwh  
technical solutions  
simulation services



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UNIVERSITÄT  
WIEN

## 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

## Comparability of Models and Results

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

**METHOD: Comparative Modelling**

# Example: Calibration

## 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

Calibration algorithms require thousands of simulation runs

**Problematic**



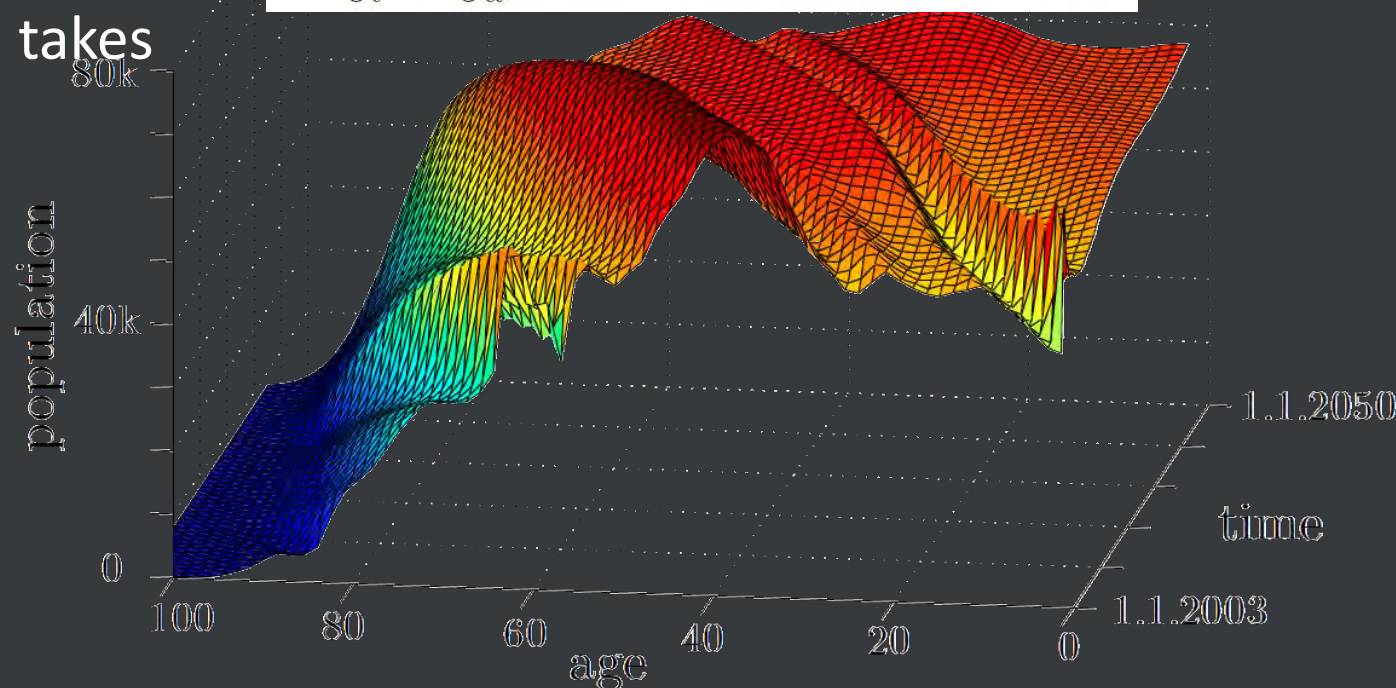
Each simulation run contains random elements and takes some time

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



Formal model analysis

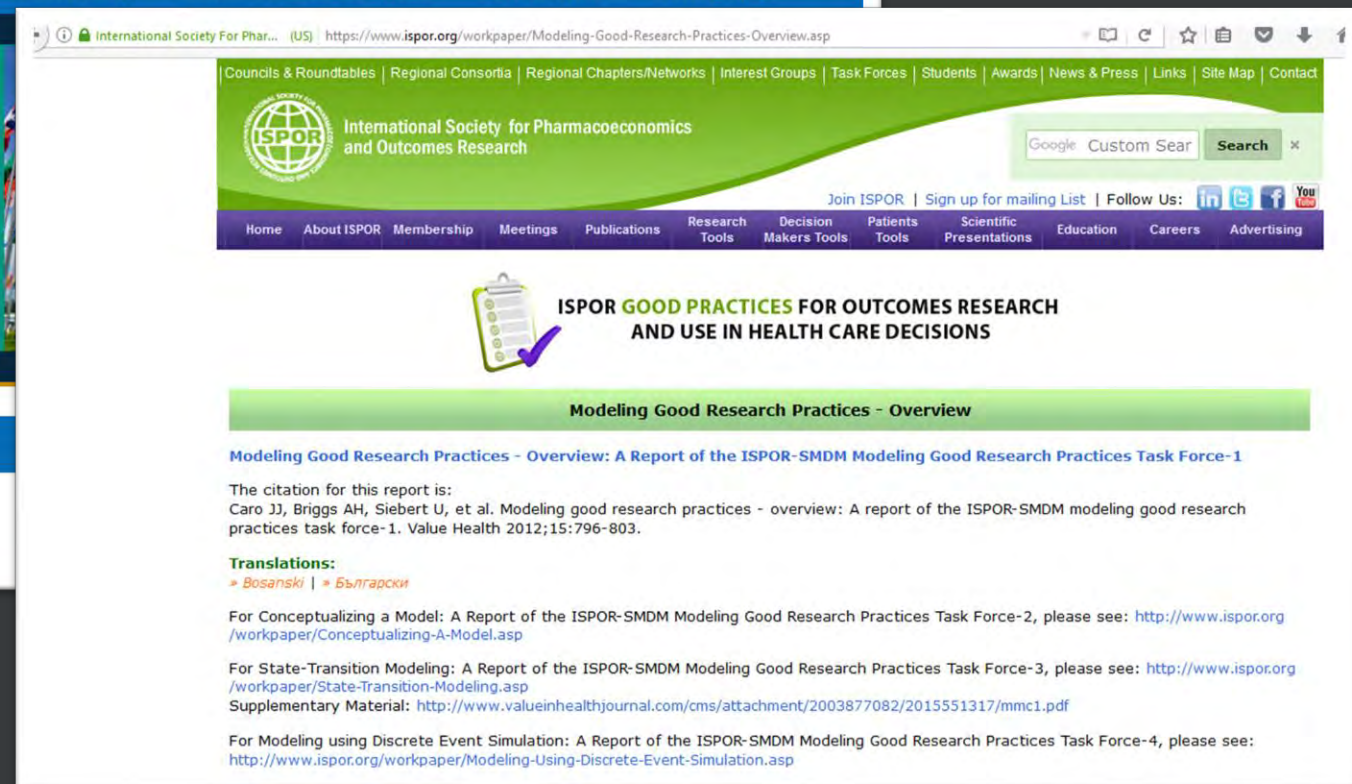
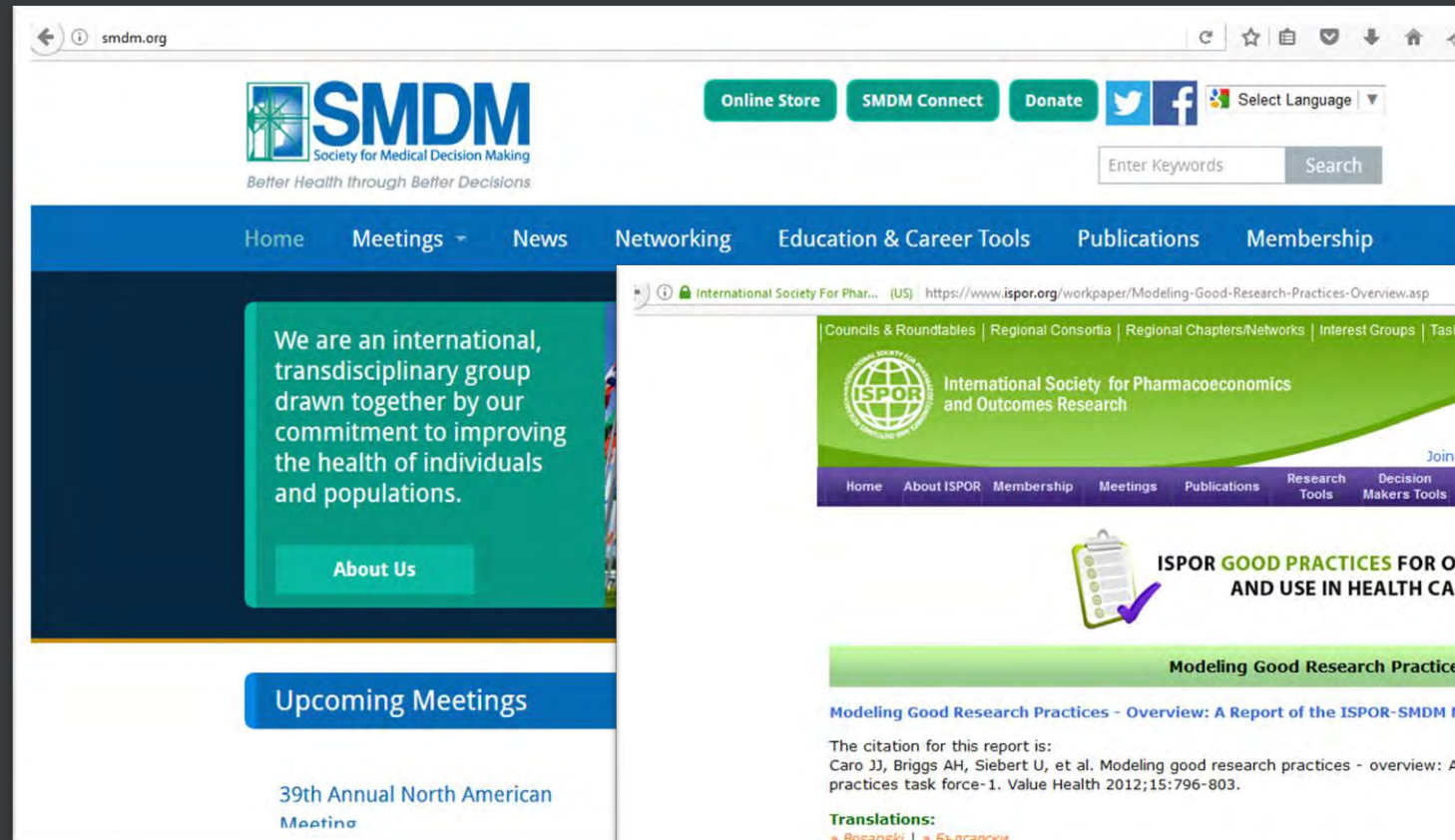
$$\begin{aligned}\frac{\partial M}{\partial t} - \frac{\partial M}{\partial a} &= \alpha_1 + \gamma_1 \mathbb{1}_{[0,dt)}(a) \Psi(a, t) - M \delta_1 \\ \frac{\partial F}{\partial t} - \frac{\partial F}{\partial a} &= \alpha_1 + \gamma_2 \mathbb{1}_{[0,dt)}(a) \Psi(a, t) - F \delta_2\end{aligned}$$



# Cross Model Validation

## 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





# 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  $J \pm i := (j_1 \pm i_1, \dots, j_k \pm i_k)$  is well-defined.

**Definition**  
Given a relative index tuple  $J$  we define the **index translation**  $T_J$  of an index  $i$  by  $T_J(i) : \mathbb{Z}^d \rightarrow \mathbb{Z}^d : i \mapsto i + J$  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 a 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  $N_{m_i, J} := \{m_{(i_1, \dots, i_k)} \in (M \cup \{\emptyset\})^k \text{ where}$

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

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  $T_{\text{metric}}$ . 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  $\mathcal{N} := T^{-1} \circ T \circ \mathcal{I} : M \rightarrow I \rightarrow \mathbb{Z}^d \rightarrow (M \cup \{\emptyset\})^k : m_i \mapsto i \mapsto (i_1, \dots, i_k) \mapsto (m_{i_1}, \dots, m_{i_k})$ .

## 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**.

Optional 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

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

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

**Special Case**  
What kind of preconditions to or characteristics of  $\tau$  generate which type geometry? (TODO)

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

<sup>a</sup>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  $S^* := (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k : (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k)$  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:  
 $\mathcal{F} \circ S \circ \mathcal{N} :$   
 $M \rightarrow (M \cup \{\emptyset\})^k \rightarrow (S \cup \{\emptyset\})^k \rightarrow S :$   
 $m \mapsto (m_1, \dots, m_k) \mapsto (s_1, \dots, s_k) \mapsto s$

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

**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-configurations<sup>a</sup> arise from the update rule.

<sup>a</sup>except 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  $S$  which can be identified with an element of  $\mathfrak{S} := S^M$ . We then also call  $S$  the **(temporary) global state**.

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

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  $\mathcal{E} : \mathfrak{S} \rightarrow \mathfrak{S} : S \mapsto \hat{S} := \mathcal{F} \circ S \circ \mathcal{N}$ .

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

**Definition**  
For an iteration, an **initial state** or **initial condition**  $S_0$  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 ...

- (i) an indexed set of cells
- (ii) an initial state mapping  $S_0$
- (iii) a relative index tuple  $J$
- (iv) a generalised index translation (only if the geometry shall be "manipulated")
- (v) an update rule  $\mathcal{F}$
- (vi) 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. Schneckenreither, Thesis: "Developing Mathematical Formalisms for Cellular Automata in Modelling and Simulation"; 2014

G. Schneckenreither, 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
4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results

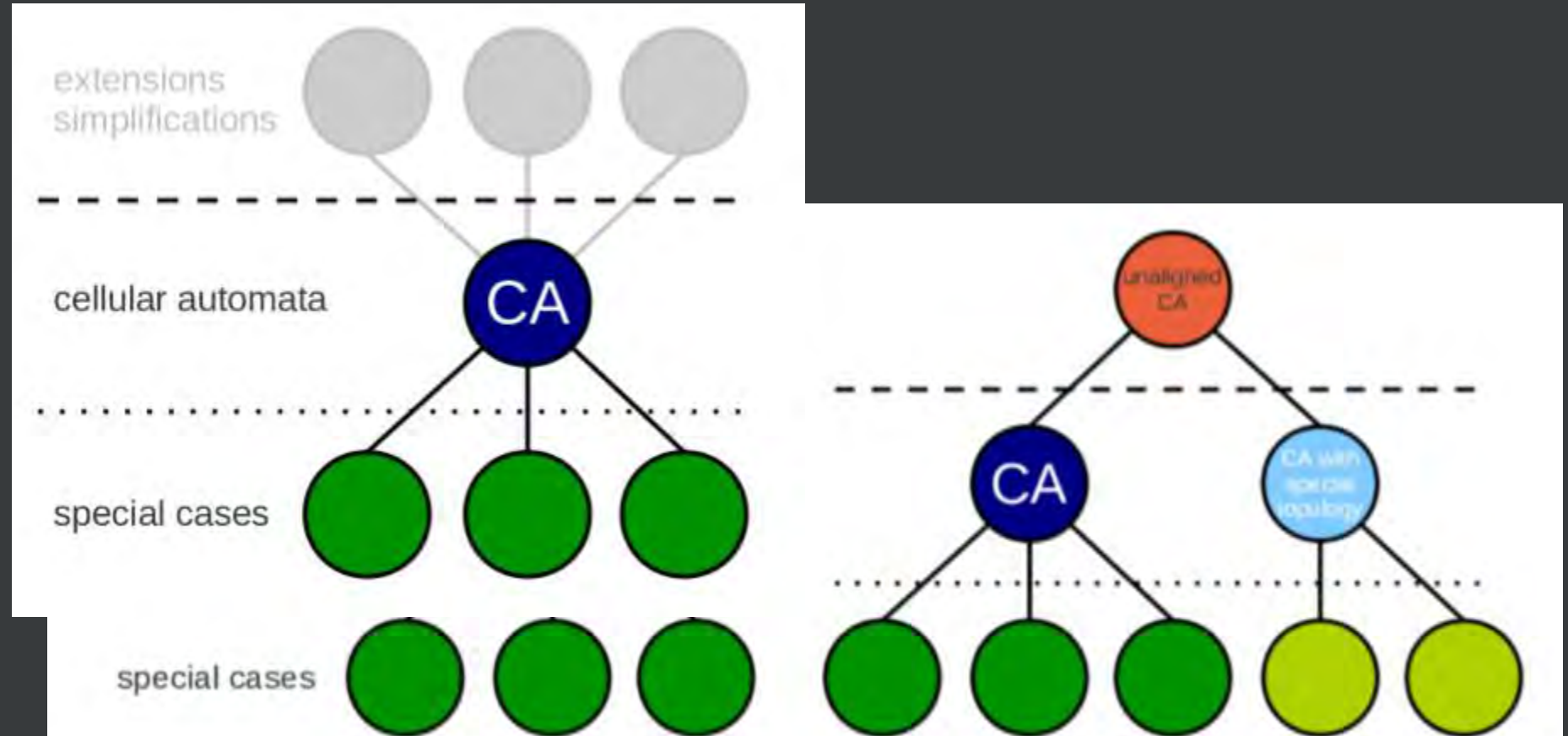


Figure 1: Different hierarchies for the classification of cellular automata.

# Concept 7

## 10 Concepts to Integrate

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4. Reproducible Processes
5. Different Methods for Different Questions (Complexity)
6. Comparability of Results
7. Make it Understandable

## Complicated Models and Various Results

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

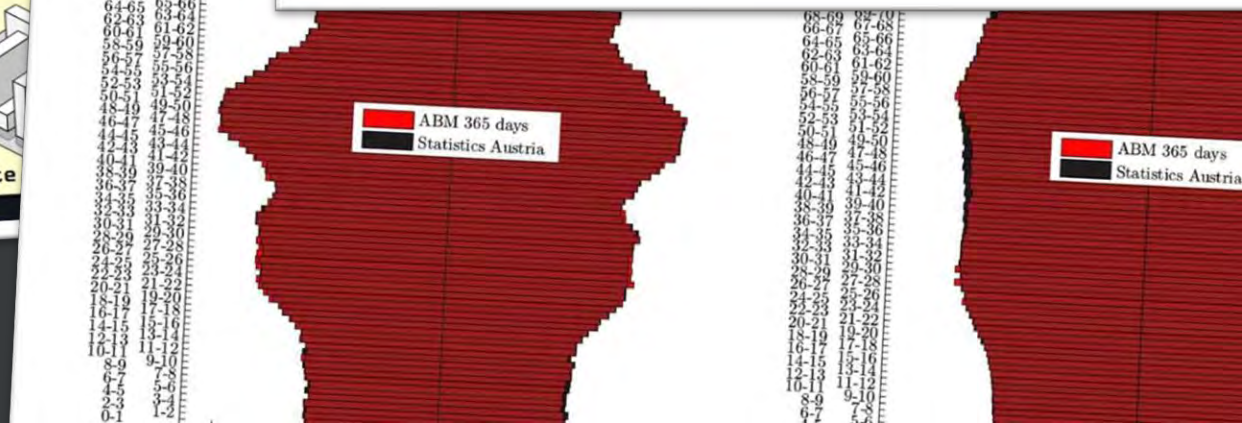
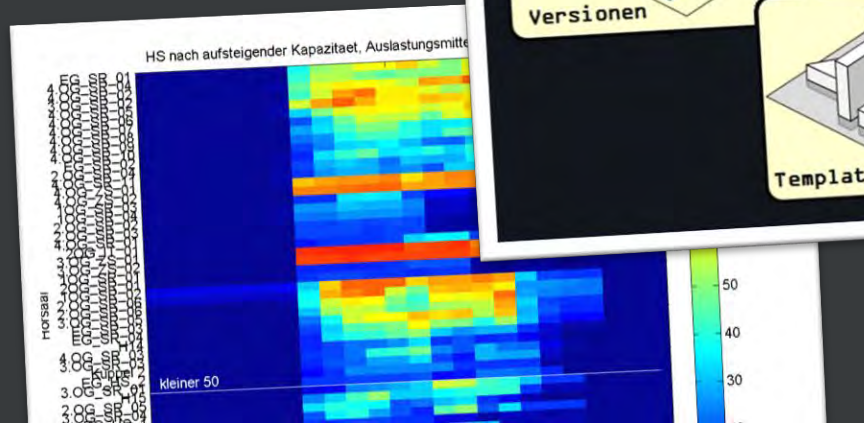
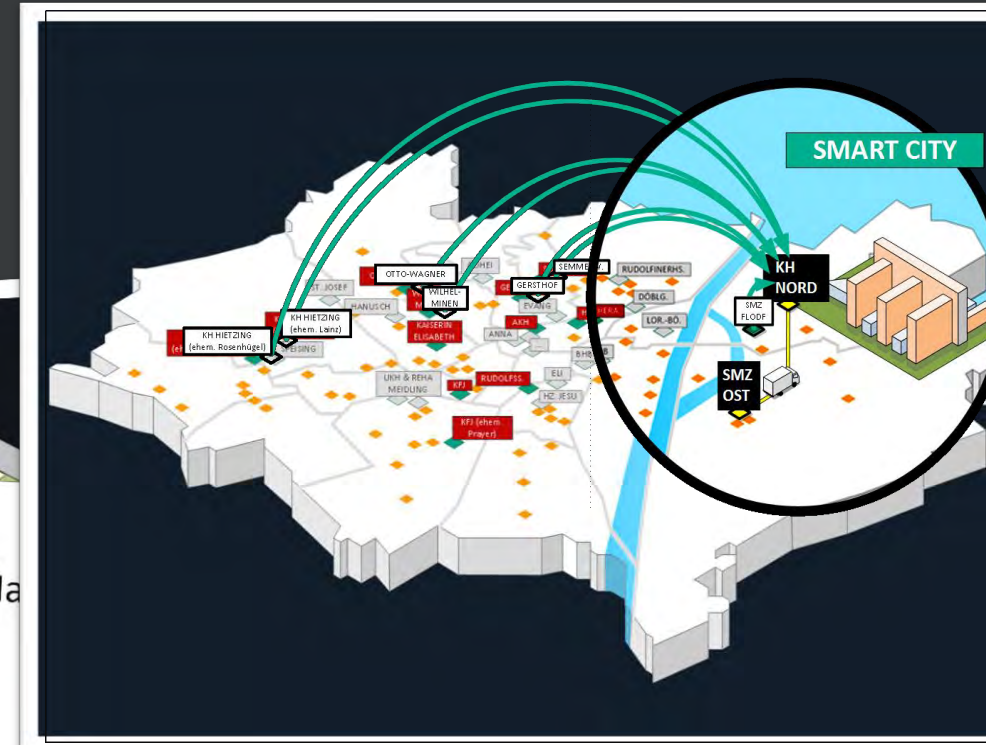
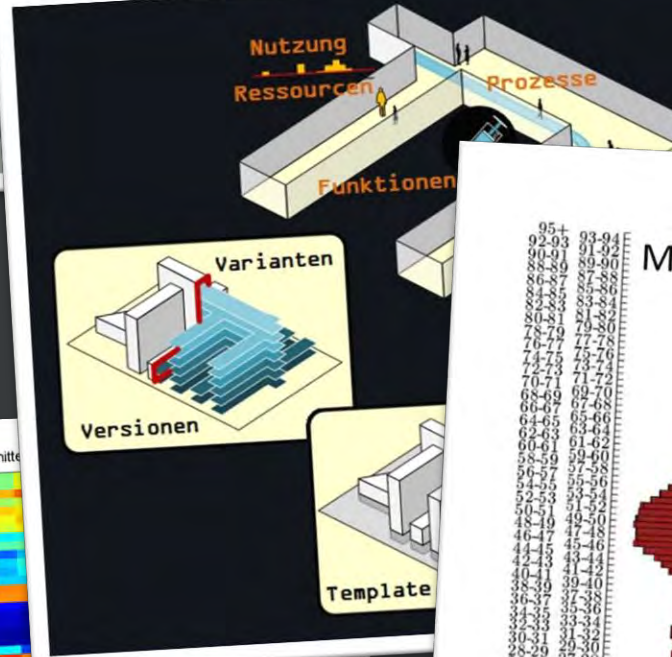
-> Change Management & Interdisciplinarity

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

# Concept 7

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1. Methods to Assess and Improve Quality of Data
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# Having Fun with Epidemics and Herd Immunity...



dwh  
technical solutions  
simulation services



TECHNISCHE  
UNIVERSITÄT  
WIEN

## 10 Concepts to Integrate

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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

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

## 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**

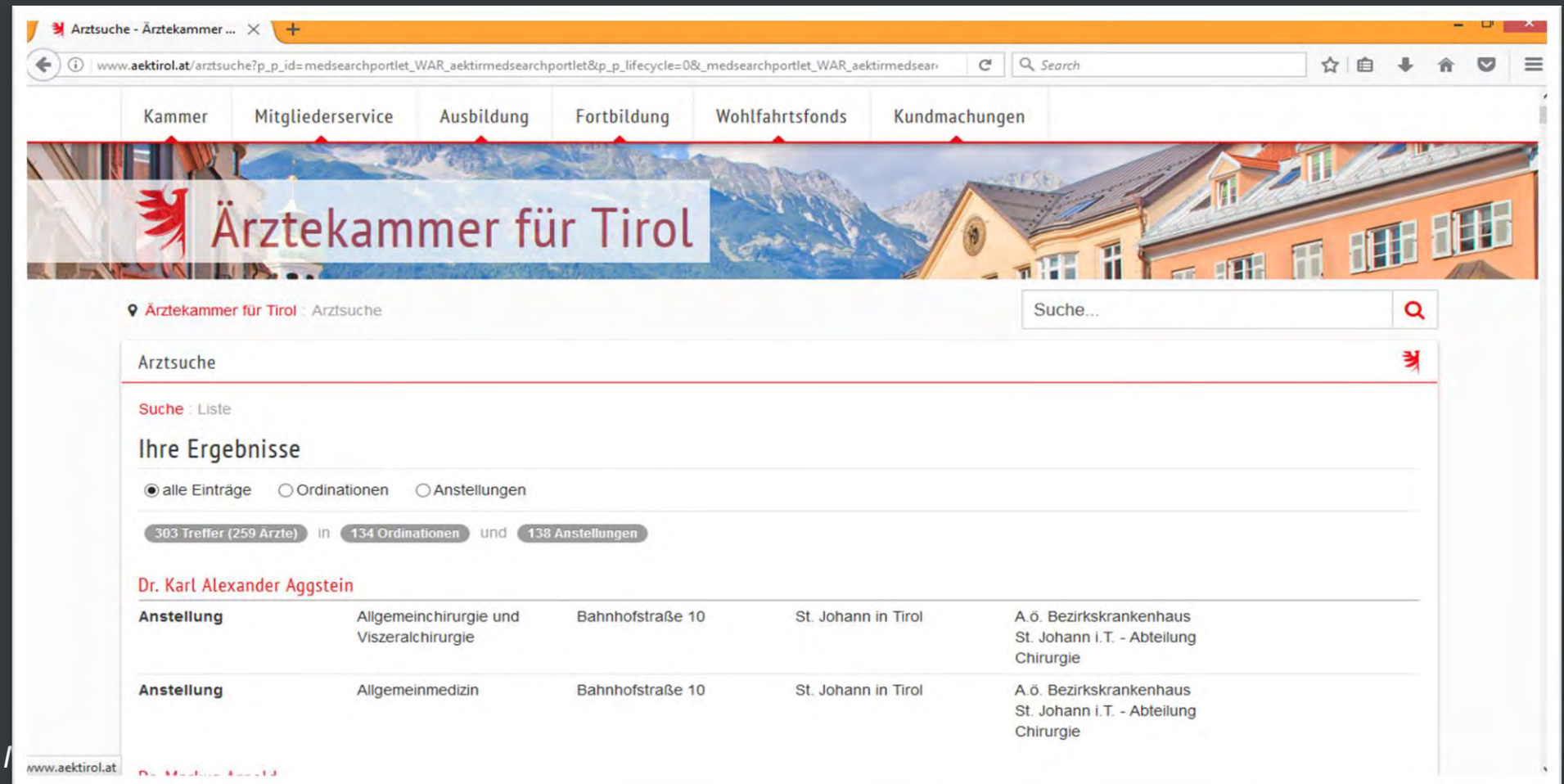


# Additional Sources: Example Web Scraping

## 10 Concepts to Integrate

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## Integration of Data from Web Pages of Physicians – Actualisation monthly



The screenshot shows the website of the Ärztekammer für Tirol (Medical Association of Tyrol). The page is titled "Arztsuche" (Physician Search) and displays search results for "Dr. Karl Alexander Aggstein". The results are organized into a table with columns for "Anstellung" (Position), "Fachbereich" (Specialty), "Adresse" (Address), "Ort" (Location), and "Krankenhaus" (Hospital).

Anstellung	Fachbereich	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

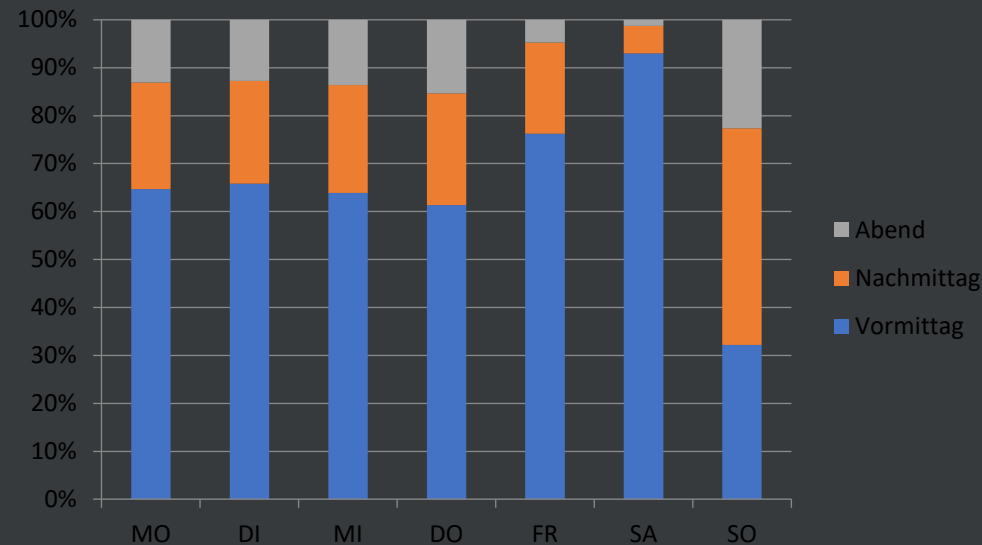
# Health Care: Physicians

## 10 Concepts to Integrate

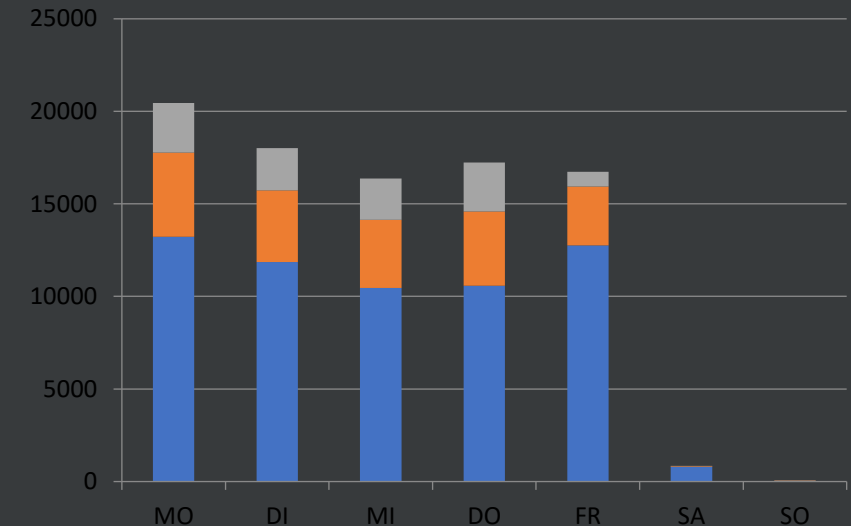
1. Methods to Assess and Improve Quality of Data
2. Potential to Integrate Missing Data
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## Example Medical Practices Hours :

relative



absolut



# Concept 9

## 10 Concepts to Integrate

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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

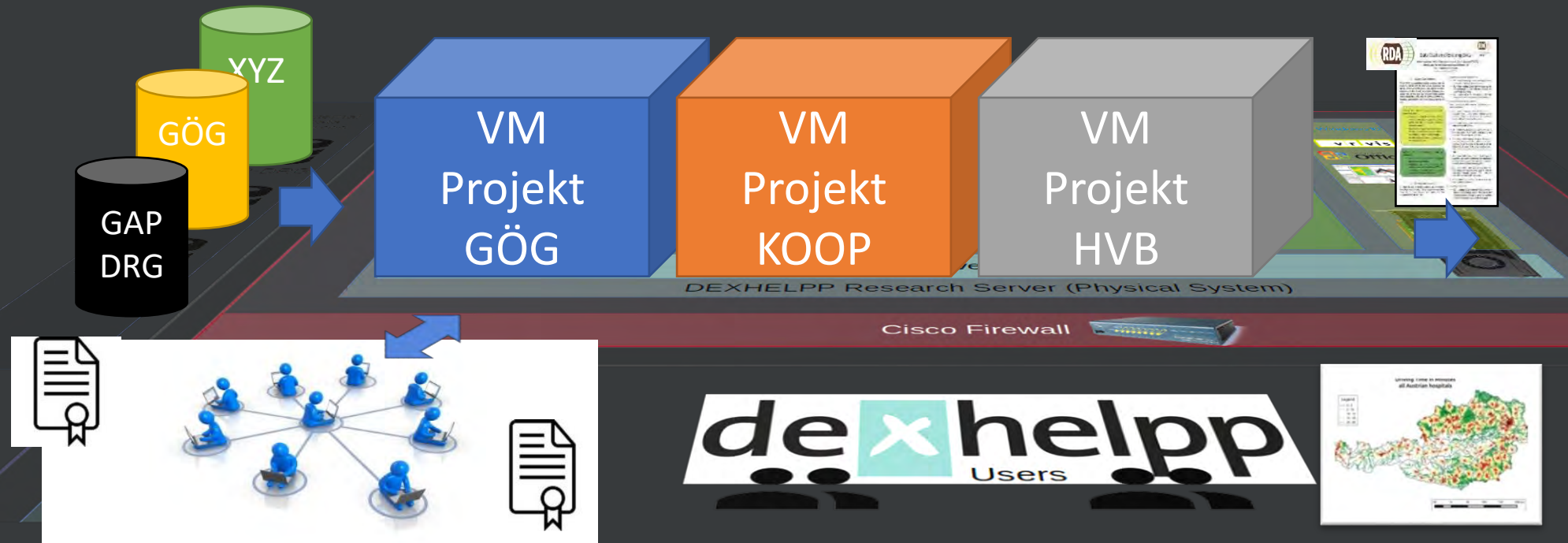
**METHOD: Data Security & Governance**



# 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

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8. Open and Independent Solutions
9. Priority for Data Security and Stake Holder Interests
10. Broad Applications (Health System, Energy, Industry, Energy, Mobility, Infra-structure)

## Broad Applications for Models

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

**METHOD: Co-Simulation, Multi Method Modelling**

# Multi-Method Airport Modelling



dwh  
technical solutions  
simulation services



TECHNISCHE  
UNIVERSITÄT  
WIEN

## 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

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## The airport city – A large complex socio-technical system



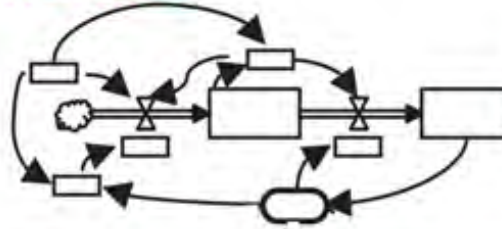
*International Civil Aviation Organization:*  
**Growth Rate Passenger Transport increased (2012) from 4.9% to 6.3% (2015)!!**

# Multi-Method Airport Modelling

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### System Dynamics



Aggregated states

Time-continuous

**Causal Links & Feedback**

average computing times

### Discrete Events Simulation



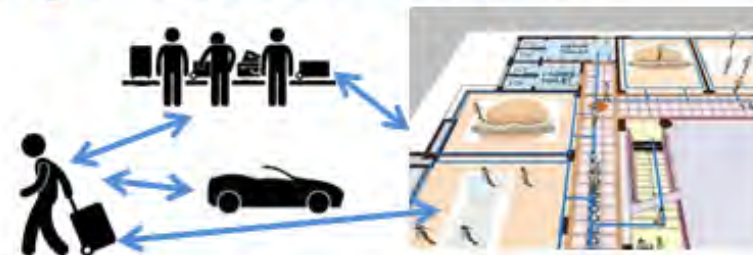
Individuals (Entities) & Resources

Changes in discrete time (Events)

**Given processes**

average computing times

### Agent-Based Models



Individuals

Time-continuous

Spatial aspects

**Interaction** (agents, environment)

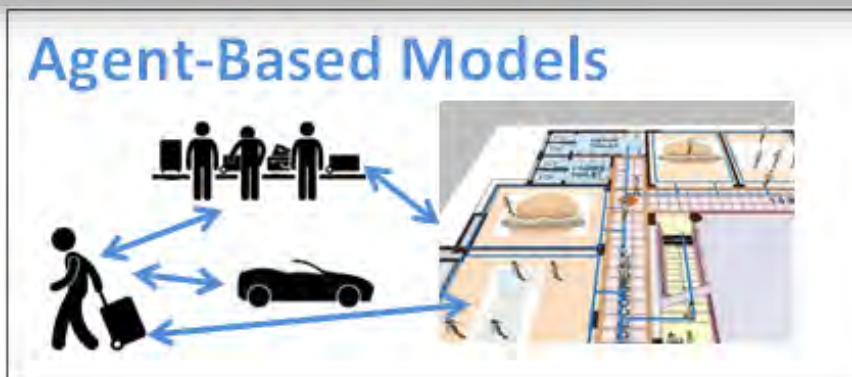
High computing times



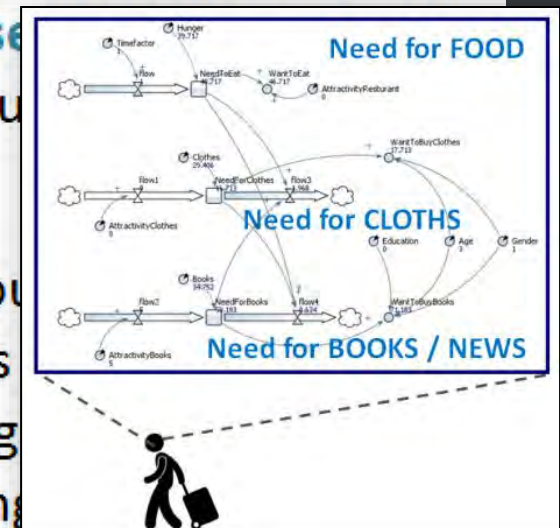
# Multi-Method Airport Modelling

## 10 Concepts to Integrate

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Individuals  
Time-continuous  
Spatial aspects  
**Interaction** (agents)  
High computing





# 10 Concepts to Integrate

## 10 Concepts to Integrate

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## Methods & Concepts to be Developed and Integrated

- |                       |  |                |                  |
|-----------------------|--|----------------|------------------|
| 1<br>Data             | • Explorative Visual Computing –Visual Analytics and Statistics                            | 5<br>Model     | 8<br>Guidelines  |
| 2<br>Data             | • Data Processes (Integration & Linkage) & Modelling Tools (Parametrization & Calibration) | 6<br>Model     | 9<br>Guidelines  |
| 3<br>Model            | • Modular Models, Coupling of Models   | 7<br>Interface | 10<br>Guidelines |
| 4<br>Data & Processes | • Validation & Data Citation   |                |                  |
|                       | • Methods for Choosing Models  |                |                  |
|                       | • Comparative Modelling  |                |                  |
|                       | • Data Representation & Human Computer Interfaces (HCI)                                    |                |                  |
|                       | • Open Access & Public Domain  |                |                  |
|                       | • Data Security & Governance   |                |                  |
|                       | • Co-Simulation, Multi Method Modelling  |                |                  |

# Implementation DEXHELPP

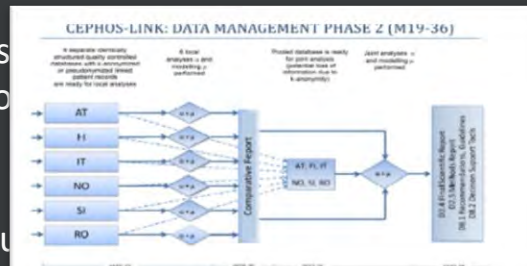
## 10 Concepts to Integrate

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)

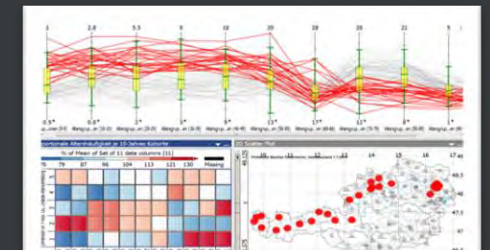
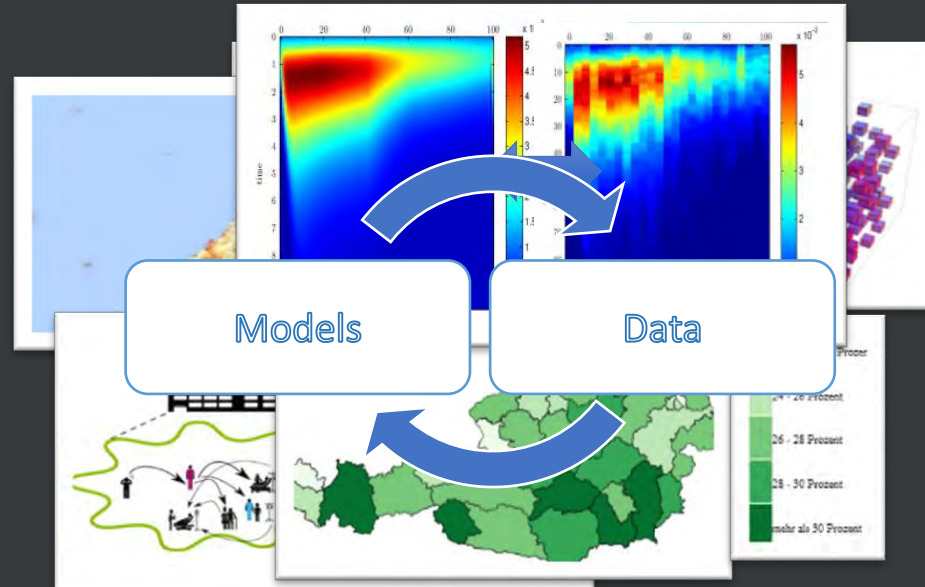


Projects with and Support for  
Decision Makers

**de**xhelpp



Formal & Technological  
Processes



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



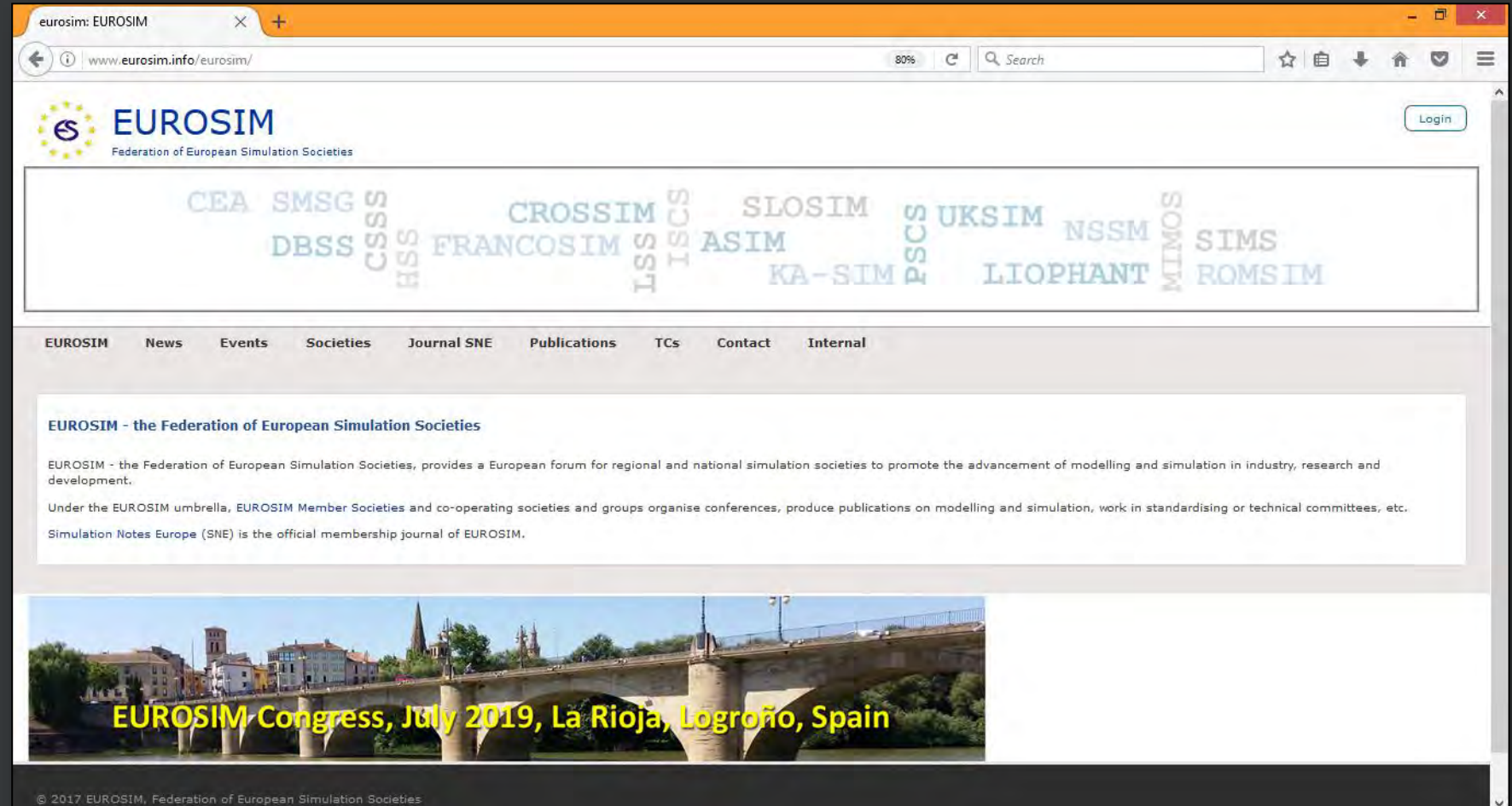
Infrastruktur  
DEXHELPP Server & Services

# EUROSIM Initiative

## DDSS Technical Committee

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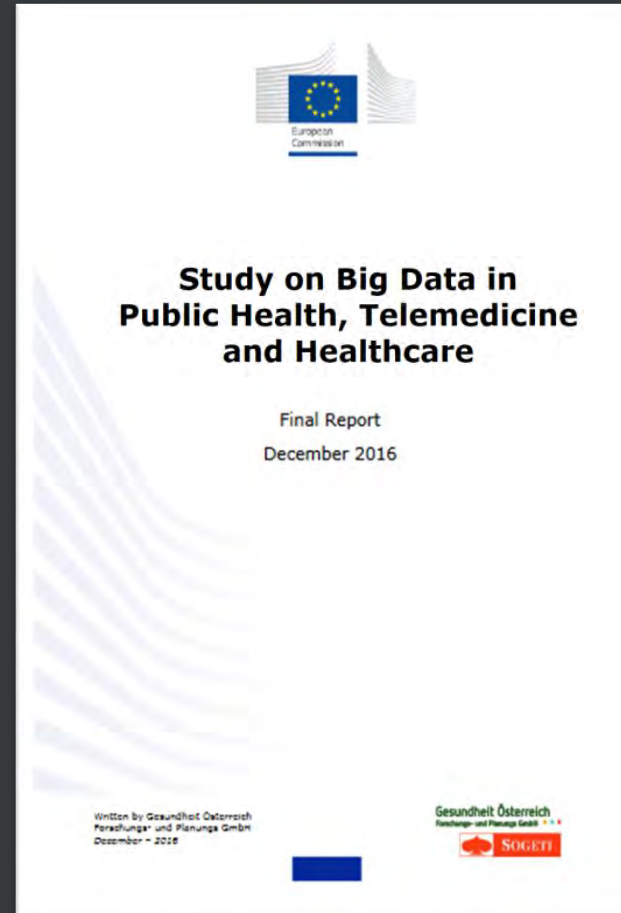




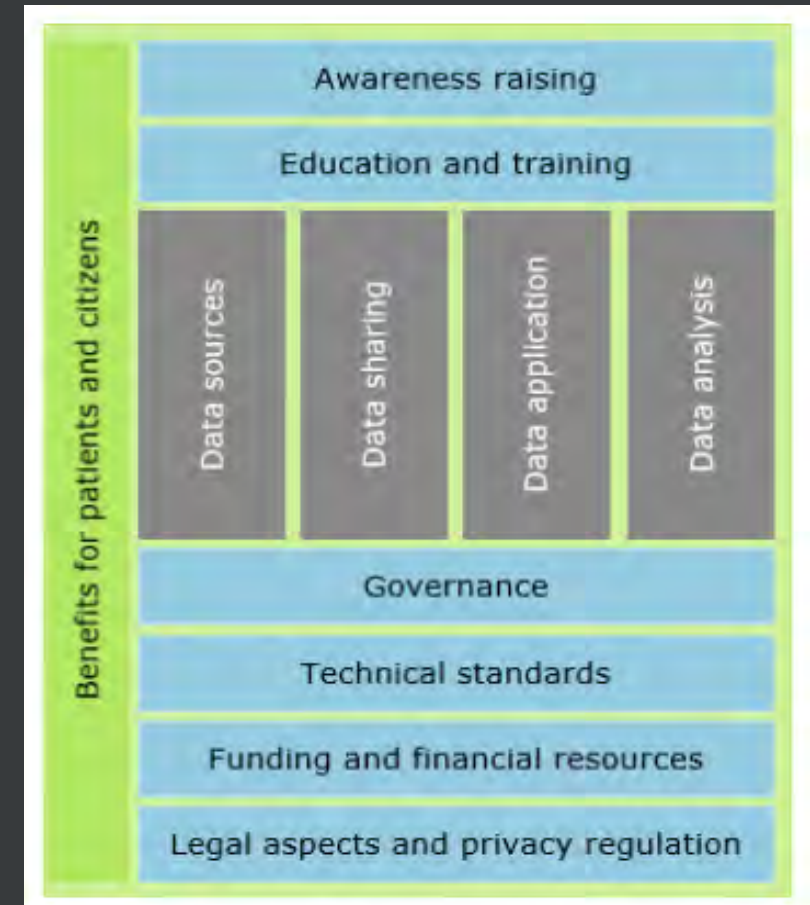
# Studies, Policies & Recommendations

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[https://ec.europa.eu/health/sites/health/files/ehealth/docs/bigdata\\_report\\_en.pdf](https://ec.europa.eu/health/sites/health/files/ehealth/docs/bigdata_report_en.pdf)



Overview of fields of policy recommendations

# 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

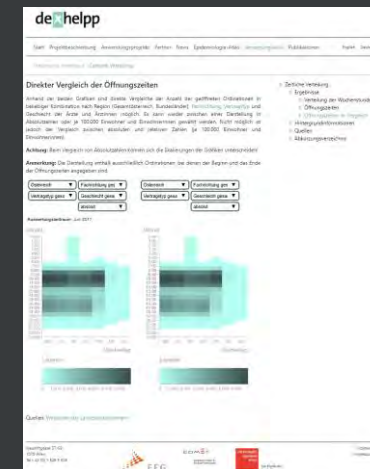
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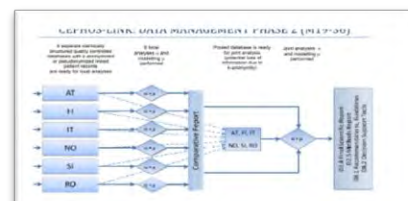
## Outlook

- 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 (4<sup>th</sup> 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

de<sup>x</sup>helpp



Formal & Technological  
Processes

  
 Competence Centers for  
Excellent Technologies

Models

Data

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

Infrastructur  
DEXHELPP Server & Services





*ISPOR - The Professional Society for Health Economics and Outcomes Research is a multistakeholder organization with more than 20,000 individual and chapter members from 110+ countries worldwide.*

## Upcoming Conferences and Education



### ISPOR Warsaw 2019

27-28 MARCH 2019 | WARSAW |  
POLAND



### ISPOR 2019

MAY 18-22, 2019 | NEW ORLEANS, LA  
| USA



### ISPOR Latin America 2019

12-14 SEPTEMBER 2019 | BOGOTÁ |  
COLOMBIA



### ISPOR Europe 2019

2-6 NOVEMBER 2019 | COPENHAGEN  
| DENMARK



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technical solutions  
simulation services



TECHNISCHE  
UNIVERSITÄT  
WIEN



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