

# Business Intelligence

## 01 (Meta-)Introduction

Prof. Dr. Bastian Amberg  
(summer term 2024)  
17.4.2024

### Professur für Wirtschaftsinformatik

#### Prof. Dr. Natalia Kliewer

Operations Research & Analytics  
Planungssysteme in Transport und Verkehr  
Robuste Effizienz  
Revenue Management

### Professur für BWL, Information und Organisation

#### Prof. Dr. Martin Gersch

E-Business  
Informationsmanagement  
Service Engineering  
Entrepreneurship Education

### Juniorprofessur für Advanced Decision Analytics

#### Prof. Dr. Bastian Amberg

Entscheidungs-  
unterstützungs-  
systeme  
Robuste Effizienz  
In Dienstleistungs-  
industrien

### Assoziierte ECDF-Professur



Demnächst wieder



#### Prof. Dr. Daniel Fürstenau

Digitale  
Infrastrukturen  
IT-Governance &  
IT-basierte Risiken  
Digital Transitions in  
Healthcare

### Assoziierte Professur



#### Prof. Dr. Christian Meske

Digital Workplace  
Collaboration  
Technologies  
Management of DT  
Processes

### Assoziierte Professur



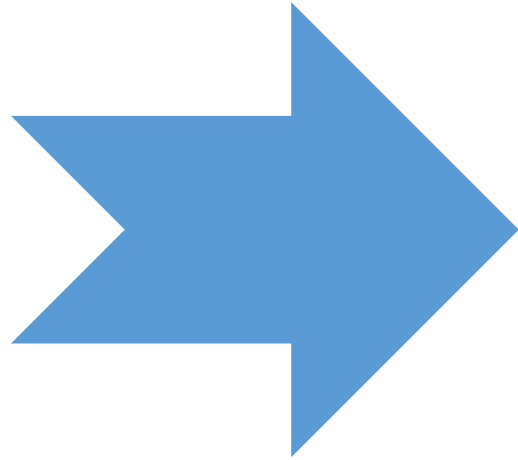
#### Prof. Dr. Hannes Rothe

Digital  
Entrepreneurship  
Digital  
Infrastructures &  
Ecosystems  
Organizing Data &  
Knowledge

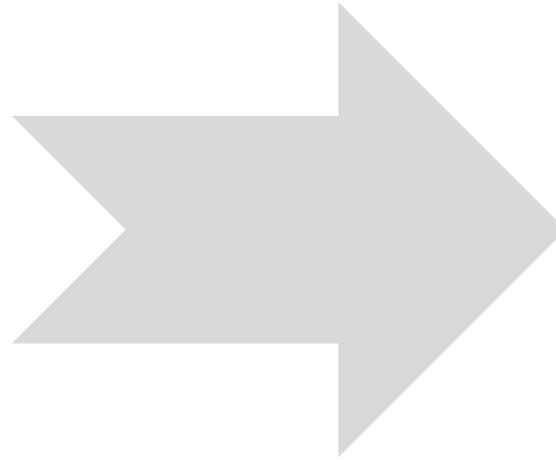
### Juniorprofessur für BWL, insb. Digital Entrepreneurship und Diversity

#### Prof. Dr. Janina Sundermeier

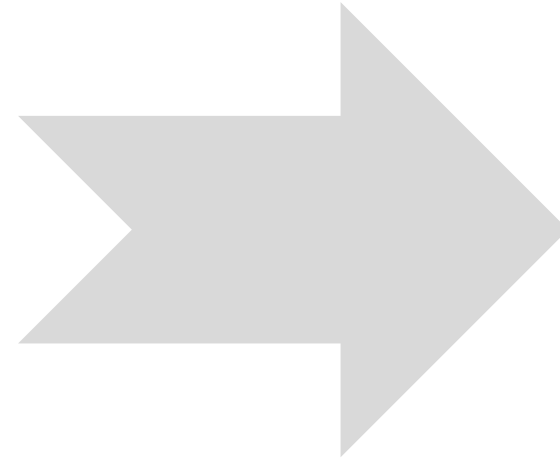
Digital Entrepreneur-  
ship  
Unternehmerische  
Diversität  
Gründungsbezogene  
Persönlichkeits-  
merkmale



(1)  
Metaintroduction



(2)  
Decision Support and  
Business Intelligence



(3)  
Data-analytic thinking

# Herzlich Willkommen...

...zur Pflichtvorlesung im 2. Semester des Studiengangs **Master-Wirtschaftsinformatik**

Interaktive Vorlesung/Seminar mit Übungseinheiten/Miniprojekten am Rechner

Seminaristischer Unterricht mit **regelmäßiger** und **aktiver Teilnahme**: auch Sie produzieren Inhalte

Unterrichtssprache: Deutsch;  
Materialien in der Regel auf Englisch

**Modul:** Business Intelligence

**Hochschule/Fachbereich/Institut:** Freie Universität Berlin/FB Wirtschaftswissenschaft/Institut für Wirtschaftsinformatik

**Modulverantwortliche/r:** Dozentinnen und Dozenten des Moduls

**Zugangsvoraussetzungen:** Keine

**Qualifikationsziele:**

Die Studentinnen und Studenten können mithilfe von Methoden der intelligenten Datenanalyse Erkenntnisse aus der Analyse großer und komplexer Datenmengen gewinnen. Sie besitzen die Fähigkeit, Simulationssysteme für die Entscheidungsunterstützung insbesondere unter Unsicherheit im betriebswirtschaftlichen Umfeld zu entwerfen, zu implementieren und einzusetzen. Sie sind in der Lage, die vermittelten Modelle, Methoden und Algorithmen in der den Fragestellungen angemessenen Weise auszuwählen und anzuwenden sowie die Handlungsempfehlungen aus der Methodenanwendung abzuleiten. Das Seminar am PC adressiert insbesondere auch überfachliche Qualifikationsziele, insb. eigenständiges Arbeiten, analytisches Denken, Präsentationsfähigkeiten sowie Fähigkeiten zur technikassistierten Aufgabenlösung in Teams.

**Inhalte:**

Spezielle Modelle und Algorithmen des Datamining, Modelle, Methoden und Grundlagen der Simulation sowie Nutzung einschlägiger Softwarewerkzeuge zur Datenanalyse und Simulation, eine Auswahl aus speziellen Techniken, wie z. B. Clustering, Assoziationsanalyse, Klassifikation, diskrete und ereignisgesteuerte, stochastische, agentenbasierte Simulation etc.

Lehr- und Lernformen	Präsenzstudium (Semesterwochenstunden = SWS)	Formen aktiver Teilnahme	Arbeitsaufwand (Stunden)
Seminaristischer Unterricht	1	Unterrichtsgespräch, Beantwortung von Diskussionsfragen, Diskussion von Anwendungsproblemen	Präsenzzeit Seminaristischer Unterricht 15
			Vor- und Nachbereitung Seminaristischer Unterricht 30
Seminar am PC	2	Kurzvorträge mit Diskussion, Diskussion von Literatur und Anwendungsbeispielen	Präsenzzeit Seminar am PC 30
			Vor- und Nachbereitung Seminar am PC 30
			Prüfungsvorbereitung und Prüfung 75
<b>Veranstaltungssprache:</b>		Deutsch	
<b>Pflicht zur regelmäßigen Teilnahme:</b>		Ja	
<b>Arbeitszeitaufwand insgesamt:</b>		180 Stunden	6 LP

Umfang der Veranstaltung: 6 LP/ **3 SWS** Dies beinhaltet:

Mittwoch, **10-12 Uhr**

→ interaktive Vorlesungen

Freitag, ~~12-14 Uhr~~ **14-16 Uhr**

→ digitale Übungen, Zeit für Vor- und Nachbereitung der Vorlesungen

Terminplan und Ankündigungen im **Blackboard** beachten!

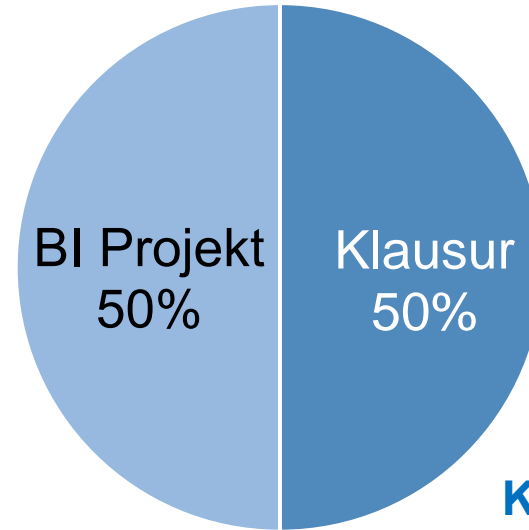
Kommunikation und Materialienbereitstellung über Blackboard

*Kurs Business Intelligence (SoSe 2024) (WIWISS\_S\_10180206\_24S)*

Unklarheiten direkt im Kontext, am Ende der Veranstaltung oder zu Beginn der nächsten Veranstaltung klären,  
ansonsten: [bastian.amberg@fu-berlin.de](mailto:bastian.amberg@fu-berlin.de)

Oder Blackboard-Forum „Organisatorische Fragen“ / „Inhaltliche Fragen“

Prüfungsleistung besteht aus **Klausur** und **Mini-Projekt**



### BI-Projekt

macht 50% der Note aus.

Bearbeitungsdauer ca. 6 Wochen, semesterbegleitend

### Klausur

macht 50% der Note aus.

Dauer 60 Minuten, im Klausurenzeitraum

(1. Termin im Zeitraum 22.7. bis 3.8.)

(2. Termin im Zeitraum 23.9. bis

In der Veranstaltung lernen Sie **Methoden und Werkzeuge** der Business Intelligence kennen.

Sie finden sich in **4er-Gruppen** zusammen und bearbeiten einen Datensatz auf relevante Fragestellungen.

*Sie wählen sich die dafür geeigneten Methoden und Werkzeuge selbstständig aus.*

Ca. **20-minütige Präsentation** der Fragestellung, des Vorgehens und Diskussion der Ergebnisse und Erstellen einen kurzen **Projektberichts**. (genaue Vorgaben hierzu später)

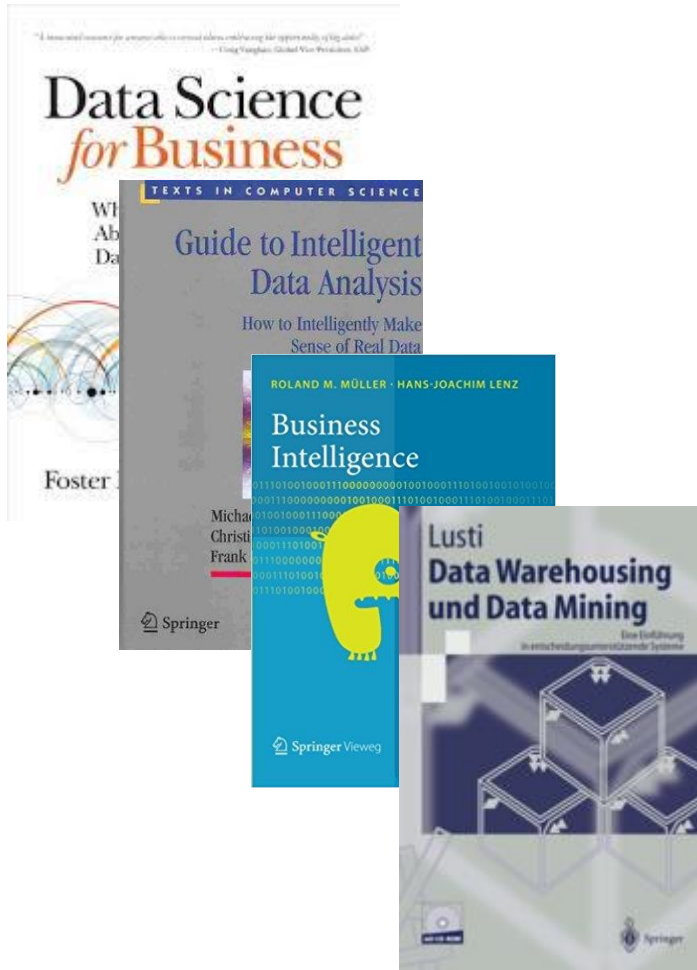
Ref.

Einzelleistung

Verständnisfragen und Anwendung

(Beispielaufgaben im späteren Veranstaltungsverlauf)





Provost, F.; Fawcett, T.: Data Science for Business. Fundamental Principles of Data Mining and Data-Analytic Thinking. O'Reilly, 2013

Berthold, M. R.; Borgelt, C.; Höppner, F.; Klawonn, F.: Guide to Intelligent Data Analysis. Springer, 2011

Müller, R. M.; Lenz, H.-J.: Business Intelligence. Springer, 2014

Lusti, M.: Data Warehousing und Data Mining, Springer, 2002

Additional readings: **see Bibliography** at the end of each presentation

# Software

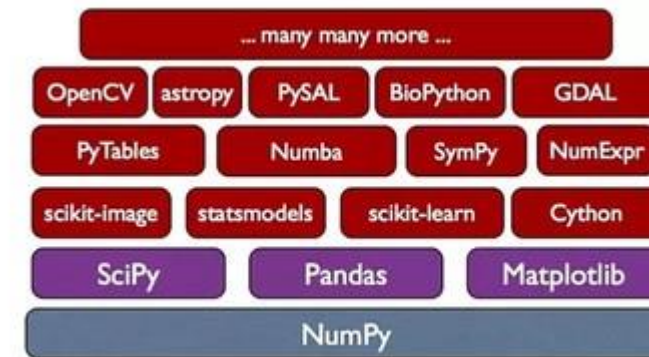
im Verlauf des Kurses

## Python Packages

- Let's agree on commonly using Python version  $\geq 3.7$   
<https://www.python.org/downloads/> (latest version is 3.12.3)
- Install Anaconda (<https://www.anaconda.com/distribution/>) (or the according packages sciPy, Pandas, NumPy, scikit-learn, Matplotlib in your own environment)

Deep Learning:

Tensorflow (<https://www.tensorflow.org/>) with conda ([install](#))



## Konstanz Information Miner (KNIME)\*

- <http://www.knime.org/>



## Weka Data Mining Software\*

- <http://www.cs.waikato.ac.nz/ml/weka/>



\*For individual needs

Ref.



# Overall goals of this class

We have three goals. After this course:

- You know how to solve business problems by **data-analytic thinking**
- You have an overview about principles of **how to model and solve** upcoming **business problems**.
- You know several **tools** and ways of how to practically **implement** solution methods

## Data Warehousing / Data Engineering

- How to **store and access** huge amounts of data?

## Data Mining / Data Science

- How to **derive knowledge and profitable business action** out of large databases?

		Wed., 10:00-12:00		Fr., 14:00-16:00 (Start at 14:30)		Self-study
Basics	W1	17.4.	(Meta-)Introduction	19.4.		Python-Basics Chap. 1
	W2	24.4.	Data Warehouse – Overview & OLAP	26.4.	[Blockveranstaltung SE Prof. Gersch]	Chap. 2
	W3	1.5.		3.5.	Data Warehouse Modeling I	Chap. 3
	W4	8.5.	Data Warehouse Modeling II	10.5.	Data Mining Introduction	
Main Part	W5	15.5.	CRISP-DM, Project understanding	17.5.	Python-Basics-Online Exercise	Python-Analytics Chap. 1
	W6	22.5.	Data Understanding, Data Visualization	24.5.	No lectures, but bonus tasks 1.) Co-Create your exam 2.) Earn bonus points for the exam	Chap. 2
	W7	29.5.	Data Preparation	31.5.		
	W8	5.6.	Predictive Modeling I	7.6.	Predictive Modeling II (10:00 -12:00)	BI-Project Start
	W9	12.6.	Fitting a Model I	14.6.	Python-Analytics-Online Exercise	
	W10	19.6.	Guest Lecture	21.6.	Fitting a Model II	
	W11	26.6.	How to avoid overfitting	28.6.	What is a good Model?	
Deepening	W12	3.7.	Project status update Evidence and Probabilities	5.7.	Similarity (and Clusters) From Machine to Deep Learning I	
	W13	10.7.		12.7.	From Machine to Deep Learning II	
	W14	17.7.	Project presentation	19.7.	Project presentation	End
Ref.					Klausur 1.Termin ~ 22.7. bis 3.8. Klausur 2.Termin ~ 23.9. bis 5.10.	Projektbericht

Case Study

# Informations on our digital Python exercises

## A) Exercises for self-study - based on Jupyter notebooks:

Notebooks can be found in Blackboard:

“Kursmaterialien > Readings & Übungen > Python-Übungen > Jupyter Notebooks“

Chapters are unlocked at

### Python-Basics:

17.04.: Chapter 1 – Erste Schritte

*solutions: 24.04.*

24.04.: Chapter 2 – Strings & String-Funktionen

*solutions: 02.05.*

02.05.: Chapter 3 – Bedingungen

*solutions: 08.05.*

### Python-Analytics:

15.05.: Chapter 1 – Spezielle Datentypen

*solutions: 22.05.*

22.05.: Chapter 2 – Datenanalyse

*solutions: 29.05.*

## B) Exercises via Webex:

**Python-Basics**

~ 17.5.

**Python-Analytics**

~ 14.6.


The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text. Uses include: data cleaning and transformation, numerical simulation, statistical modeling, data visualization, machine learning, and much more.



<https://jupyter.org/>

Rel. **Questions during/after exercises:** bastian.amberg@fu-berlin.de

# Getting Started with Jupyter-notebooks

 **ANACONDA** Products ▾ Pricing Solutions ▾ Resources ▾ Partners ▾ Blog Company ▾ [Contact Sales](#)

Individual Edition is now  
**ANACONDA DISTRIBUTION**

The world's most popular open-source Python distribution platform <https://www.anaconda.com/distribution/>

Anaconda Distribution

[Download](#)

## Anaconda Installers

### Windows

Python 3.9  
64-Bit Graphical Installer (510 MB)  
32-Bit Graphical Installer (404 MB)

### MacOS

Python 3.9  
64-Bit Graphical Installer (515 MB)  
64-Bit Command Line Installer (508 MB)

### Linux

Python 3.9  
64-Bit (x86) Installer (581 MB)  
64-Bit (Power8 and Power9) Installer (255 MB)

**2.**

Anaconda Navigator

File Help

**ANACONDA NAVIGATOR**

Home

Applications on root Channels

**jupyter notebook** 4.2.1  
Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis.  
[Launch](#)

**qtconsole** 4.2.1  
PyQt GUI that supports inline figures, proper multiline editing with syntax highlighting, graphical calltips, and more.

**spyder** 4.1.2  
Scientific Python Development Environment. Powerful Python IDE with advanced editing, interactive testing, debugging and introspection features

**3.**

Jupyter

localhost:8888/tree/jupyter/notebooks

**4.**

3.0 Dein erstes Python-Programm

In dieser Übung programmierst du gleich ein komplettes Programm mit. Anhand dessen bekommst du einen ersten Eindruck davon, was man alles mit Python und bereits ein paar Zeilen Code vereinfachen kann.

Du kennst dich damit über eingesperrte Videos an Pausen vom Computer erinnern lassen und die Dauer deiner Übungzeit festlegen. Die Dauer gibt dir in dem Programm an, z.B. 45 Minuten, danach spielt dir dein Programm ein zufällig ausgewähltes Video aus einer von dir angelegten Videolink-Sammlung ab.

Das Programm kannst du natürlich auch bei allen anderen Aktivitäten an deinem Computer laufen und dich somit an regelmäßige Pausen vom Bildschirm erinnern lassen -)

Die Einzelschritte zur Erstellung des Programms werden dir in diesem Jupyter Notebook genau aufgeschlüsselt.

**Tipp:** Probiere die Code-Zellen von Anfang bis Ende in ihrer gegebenen Reihenfolge aus. Da sie ein zusammenhängendes Programm darstellen, wirst du Fehlermeldungen erhalten, wenn du z.B. die 3. Code-Zelle vor der 1. Code-Zelle ausführen lässt. Der Grund hierfür ist der generelle Aufbau eines Programms, der von oben nach unten beginnt und dann mit dem letzten Ausdrucks von unten nach oben endet.

In der folgenden Code-Zelle siehst du das komplette Programm. Wichtige Zeilen Code rechen schon aus, um die Funktionalitäten umzusetzen, die du für deine Pausen-Erinnerungen brauchst. Betrachte die Video-Links nur als Beispiele, die du durch eigene ausgewählte Links ersetzen darfst.

**Übung:** Starte die untere Code-Zelle, indem du in sie hinein und dann auf Run (oben im Jupyter Notebook) klickst. In der Ausgabe unter der Programm-Code-Zelle wird ausgegeben, zu welcher Zeit das Pausen-Programm gestartet ist. Nach einer Stunde beginnt die erste Pause.

```
In [ ]: import time
import webbrowser
import random

print("Ich bin zu Hause und beginne mit dem Programm. Nach einer Stunde beginnt die erste Pause.")
```

**Alternative zum schnellen Testen ohne Installation:** Sind die Notebooks einmal lokal gespeichert, können sie z.B. über *CoCalc* (*Collaborative Calculation and Data Science Service*) geöffnet, kopiert, umbenannt, ausgeführt und bearbeitet werden. Es ist – Stand April 2024 – allerdings die Erstellung eines (kostenlosen) Accounts notwendig. Link zu CoCalc: <https://cocalc.com/> bzw. <https://cocalc.com/features/jupyter-notebook>. Nach Login kann ein Notebook oder mehrere hochgeladen werden (unter „File > Upload... > Upload“ oder unter „File > Open... > Upload“) und es kann zum Editieren kopiert, umbenannt und bearbeitet werden. Das bearbeitete Notebook kann gespeichert und anschließend heruntergeladen werden, um es lokal zu sichern.

# What are your expectations?

?

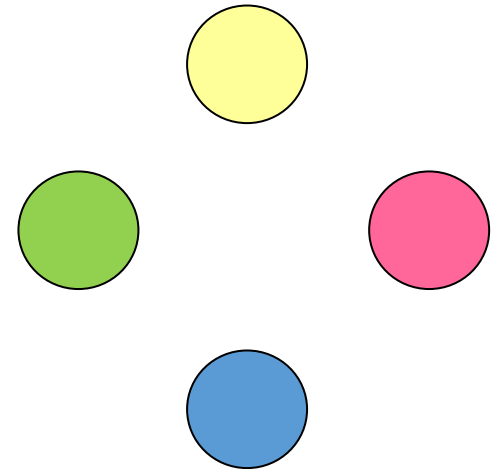
?

Kahoot  
[www.kahoot.it](https://www.kahoot.it)  
(über Smartphone  
oder Laptop)  
PIN folgt

# What is your background?

!

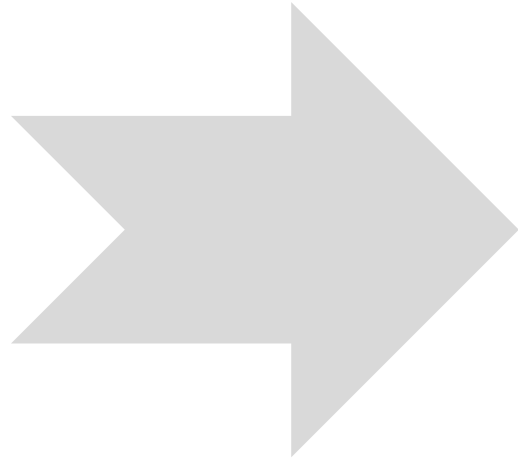
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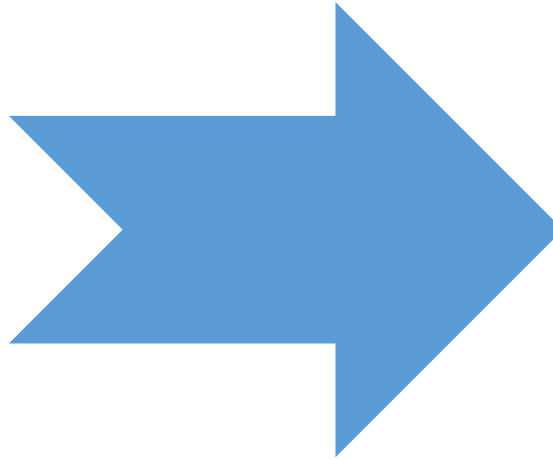


# Agenda

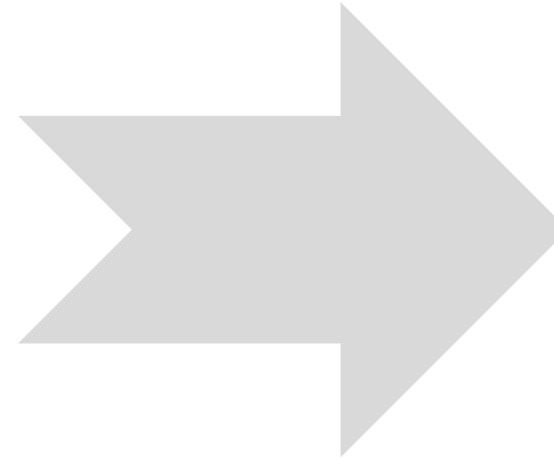
Let's put BI into perspective



(1)  
Metaintroduction



(2)  
Decision Support and  
Business Intelligence



(3)  
Data-analytic thinking

# Decision Support Systems

And variants of problem modelling

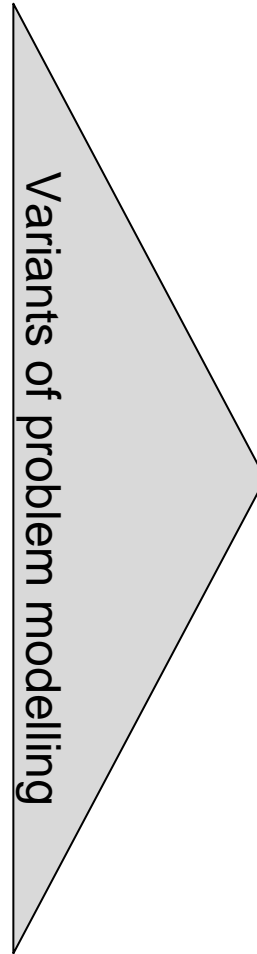
**Decision Support Systems** in the broadest sense can be defined as

*Computer technology solutions that can be used to support complex decision making and problem solving.*

(Shim et al. 2002)

Broad definition that encompasses many areas

- Application systems
- Mathematical modeling
- Data driven modeling



Common **mathematical modelling**

- All the relevant variables and relations can be identified
- The nature of the problem can adequately be caught by mathematical models
- Optimal solution for the underlying decision problem can be derived from the model (e.g., linear programming)

**Data driven (empirical) modelling**

- Problem too complex to identify all the relevant variables and relationships
- Gain insight into problem structure by analysis of historical (transactional) data (e.g., data mining)
- Entails trial-and-error experiments and oftentimes black-boxing

# Business Intelligence: Definition

There is no unique or mathematical definition of Business Intelligence.

The Data Warehousing Institute defines Business Intelligence as...

*... the process, technologies and tools needed to **turn data into information, information into knowledge and knowledge into plans** that drive profitable business action. Business intelligence encompasses data warehousing, business analytics tools, and content/knowledge management.*

(<http://www.tdwi.org/>)



Ref.

Another (similar) definition:

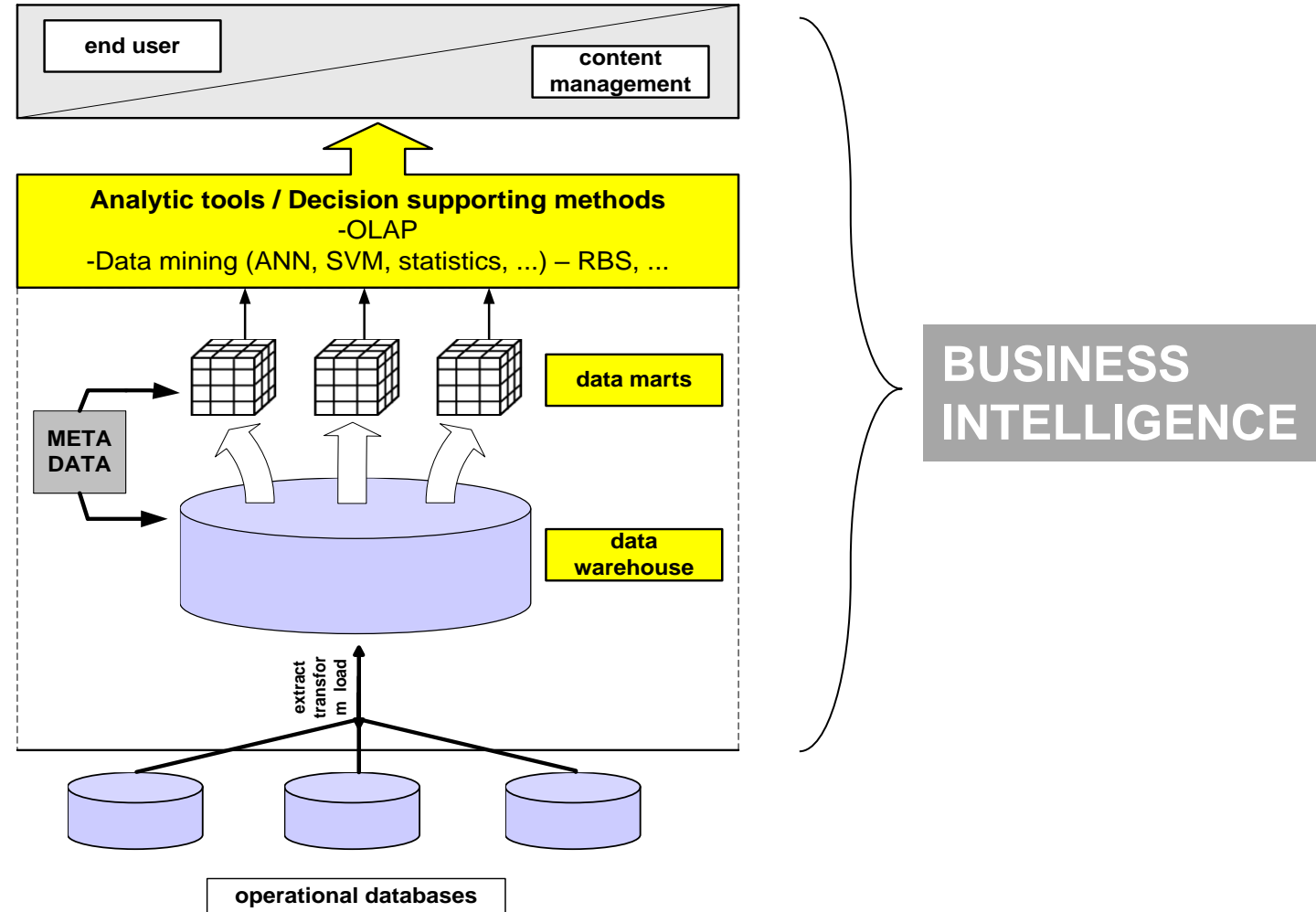
*Business intelligence is the conscious, methodical **transformation of data** from any and all sources into new forms to provide **information** that is **business driven and results oriented**. It will often encompass a mixture of tools, databases and vendors.*

(Mike Biere, specialist for business intelligence analytic tools, IBM)



# Other view on Business Intelligence

A holistic view on BI



## Increased profitability

Distinguish between profitable and non-profitable customers

## Improved Customer-Relationship-Management

Analysis of aggregated customer information to provide better customer service, increase customer loyalty



## Decreased costs

Lower operational costs, improve logistics management

## Decreased risk

Apply Business Intelligence methods to credit data can improve credit risk estimation

Business Intelligence can help improve businesses in a variety of fields:

Customer analysis

→ customer profiling

Sales channel analysis

Behavior analysis

→ fraud detection, shopping trends, web activity, social network analysis

Business productivity analysis

→ defect analysis, capacity planning and optimization, risk management, increase sustainability

Supply chain analysis

→ supply and vendor management, shipping, distribution analysis, sustainable supplier management



Business Intelligence supports decision makers with the required information at the right time and location, and with sufficient quality (format, visualization, validity etc.)

# Discerning „Data“ from „Information“

## Excursus: Semiotics

Semiotics is a field of research in Epistemology („meaning making“) and studies signs (icons, characters, symbols) and sign systems.

### Layer of observation

### Object of observation

### Example

#### Syntax

*“Relations among signs in formal structures”*

Is the communicated sign (orthographically/ grammatically) correct?

#### Semantics

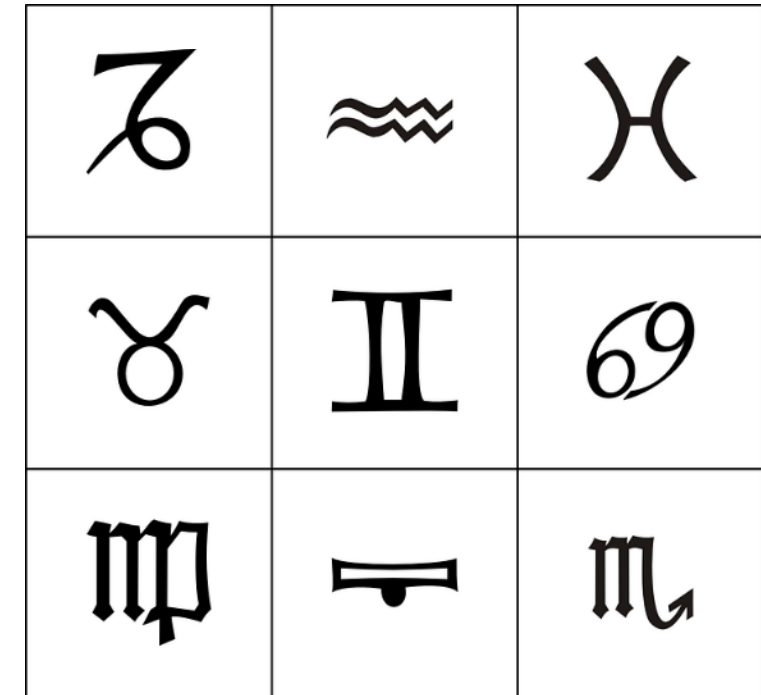
*“Relation between signs and the things to which they refer; their [...] meaning”*

What is the meaning of the sign?  
What is the meaning of „display“ or „windows“?

#### Pragmatics

*“Relation between signs and the effects they have on the people who use them”*

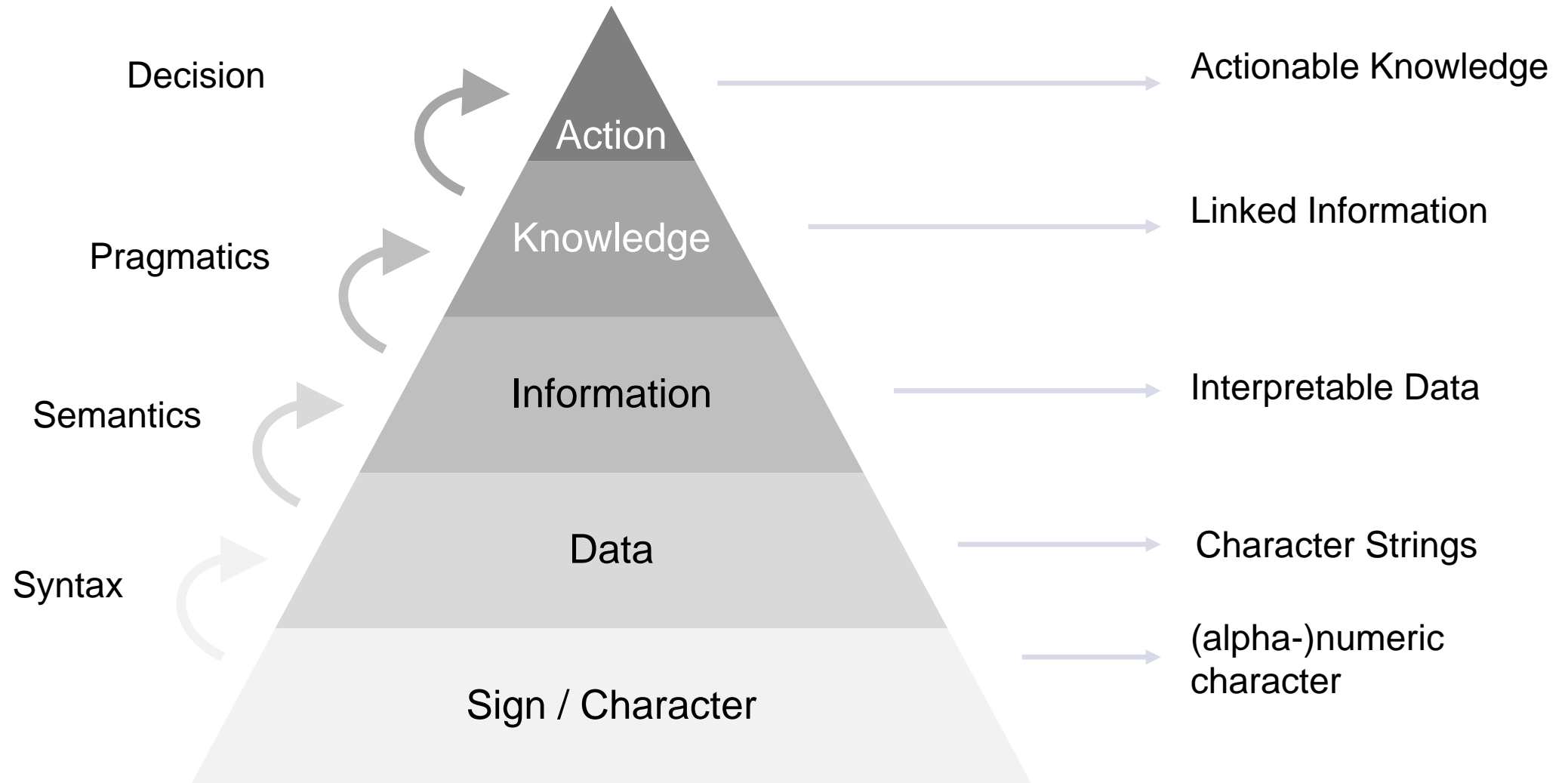
What is the use of this sign?





# Data transformation (1/2)

## Knowledge Pyramid



# Data transformation (2/2)

Transforming Data into information; into knowledge; into action

## From data to information

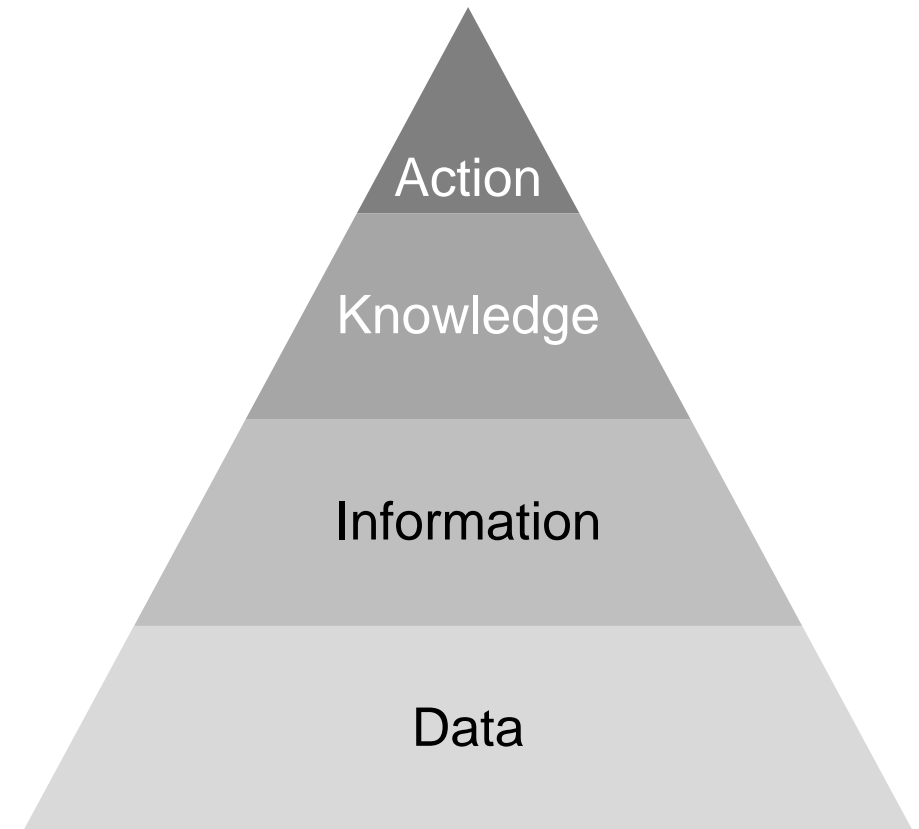
- **Select relevant data** and figure out which configuration makes the data more significant

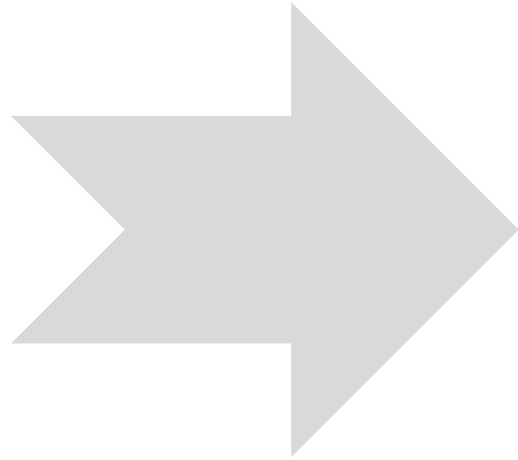
## From information to knowledge

- Piles of information are accumulated and analyzed in different ways
- In this process **analytical tools** are involved

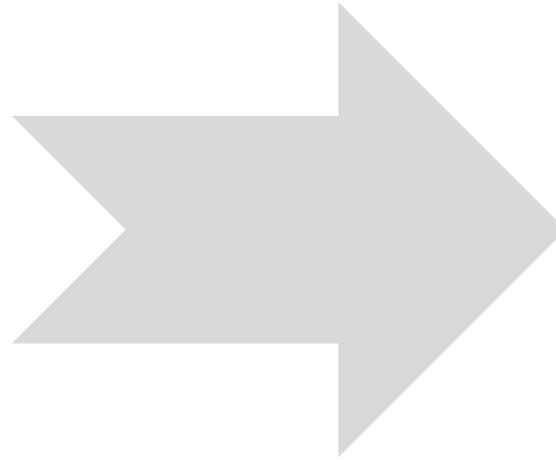
## From knowledge to actionable (critical) knowledge

- Knowledge is considered to be critical if it can be used to form a plan of action for **solution of a business problem**

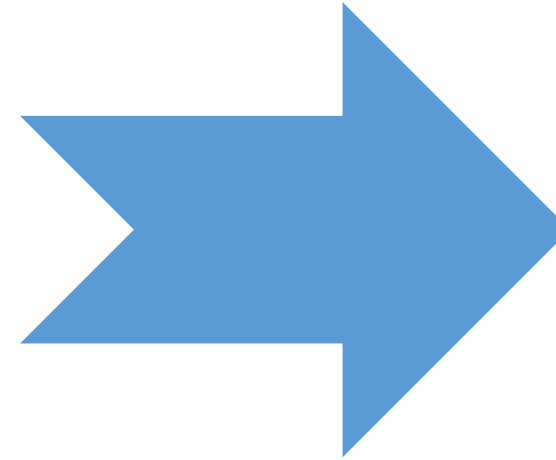




(1)  
Metaintroduction



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(3)  
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# Data Driven Decision-making (DDD)

Legitimizing decisions on the basis of data

Data-driven Decision Making (DDD) is a *“practice of basing decisions on the analysis of data rather than purely on intuition.”*

Provost & Fawcett (2013)

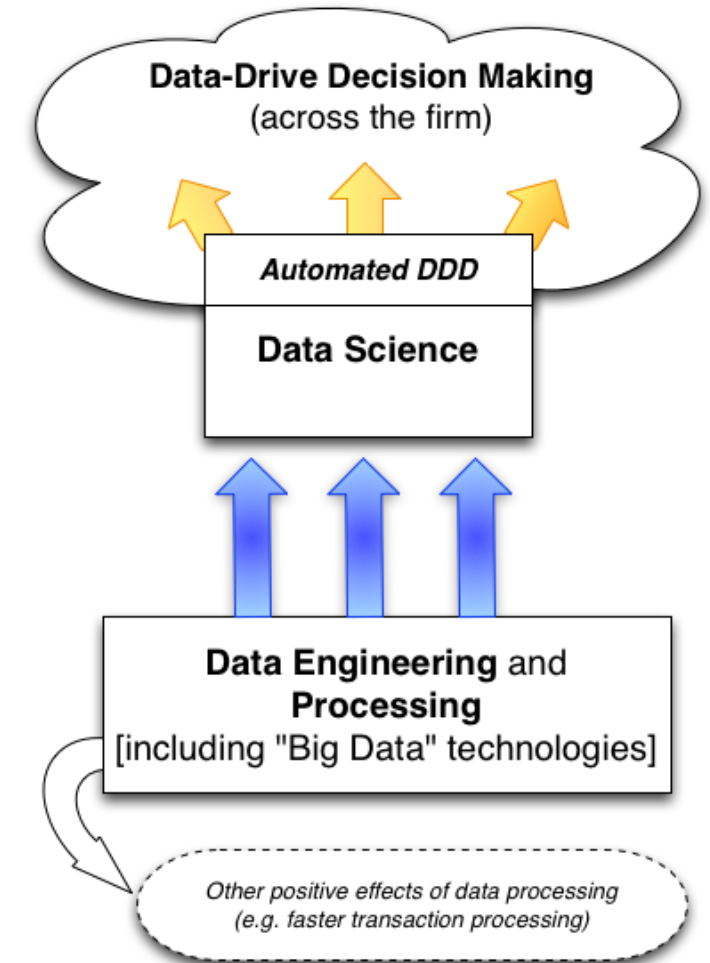
## Type-1 decision:

“discover” something new within your data

## Type-2 decision:

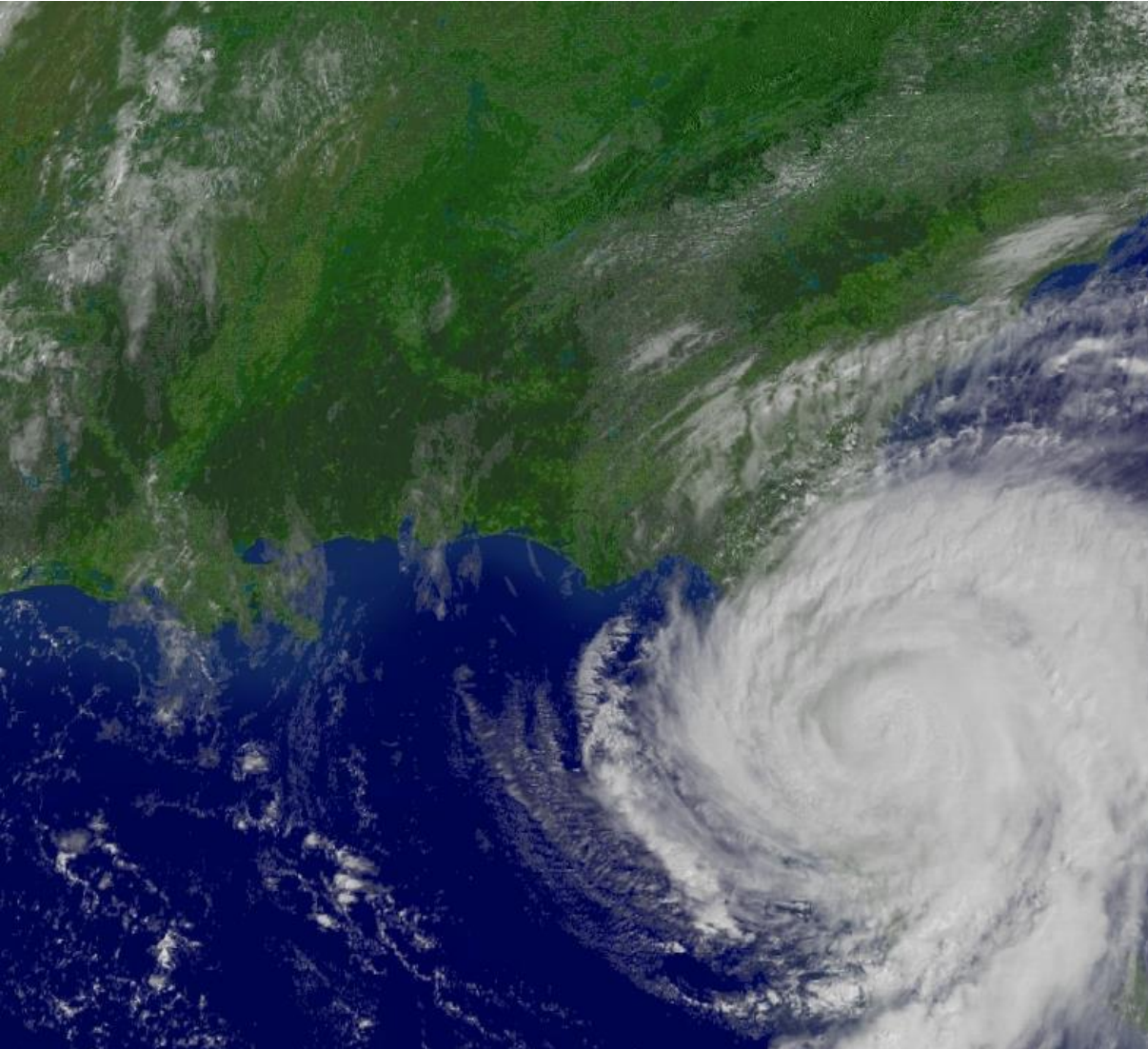
repeat decisions at massive scale  
(automatic decision making)

Ref. Provost & Fawcett, 2013 ([Link](#))



# Type 1 decision

Discover something new: Hurricane Frances (1/2)



*Hurricane Frances was on its way, barreling across the Caribbean, threatening a direct hit on Florida's Atlantic coast. Residents made for higher ground, but far away, in Bentonville, Ark., executives at Wal-Mart Stores decided that the situation offered a great opportunity for one of their newest data-driven weapons ... **predictive technology**.*

*A week ahead of the storm's landfall, Linda M. Dillman, Wal-Mart's chief information officer, pressed her staff to come up with **forecasts** based on what had happened when Hurricane Charley struck several weeks earlier. Backed by the trillions of bytes' worth of shopper history that is stored in Wal-Mart's data warehouse, she felt that the company could 'start predicting what's going to happen, instead of waiting for it to happen,' as she put it.*

*(Hays, New York Times, 2004)*



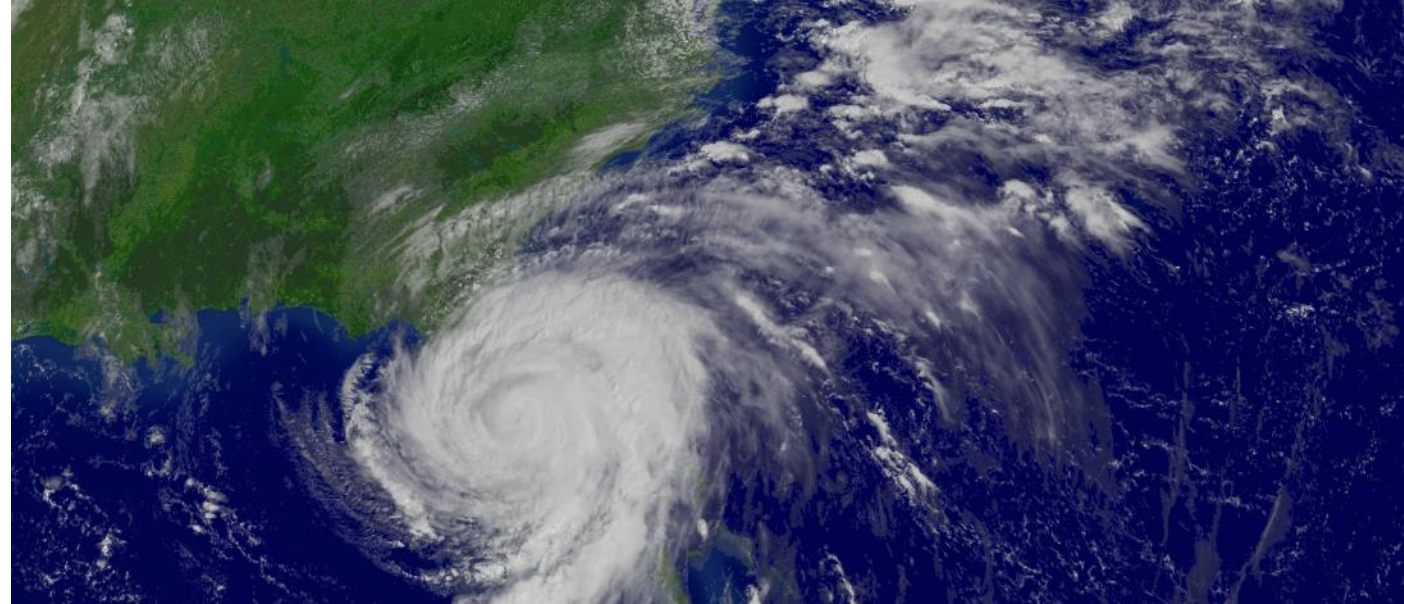
# Type 1 decision

Discover something new: Hurricane Frances (2/2)

Why might data-driven prediction be useful?

Analytic thinking: discover sales patterns due to the hurricane that are not obvious!

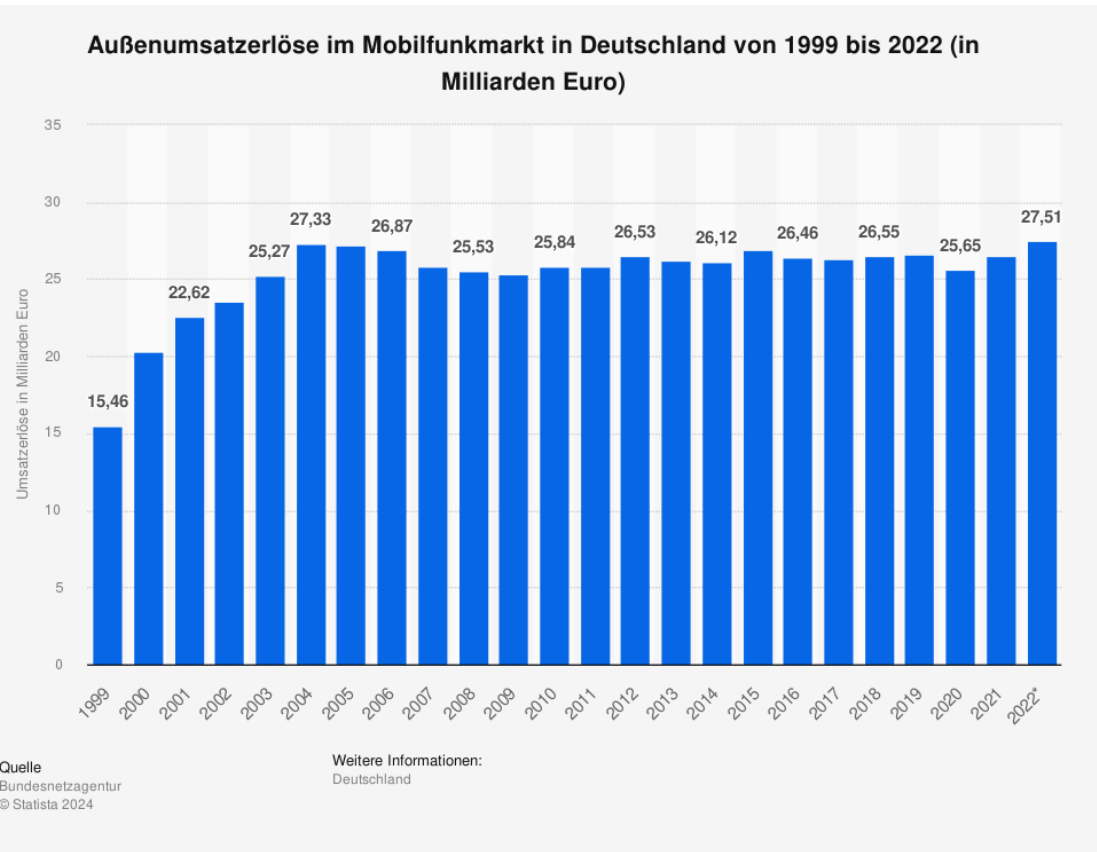
Identify unusual local demand for products (not through hypothesis testing, but data exploration)





# Type 2 decision

## Automatic decision making: Predicting Customer Churn



Mobile TeleCom market is highly saturated.

Ref. [Churn rates from Statista](#)



Many TeleCom companies have great issues with customer retention.

(current churn rates between 1 and 3 % for Deutsche Telekom mobile)

Customer churn is expensive.

Who should get a retention offer?

Raising accuracy of a prediction has huge effects on profitability

[www.o2online.de/mein-o2/vertrag-verlaengern/](http://www.o2online.de/mein-o2/vertrag-verlaengern/)



# Knowing about DDD, how do we proceed?

We follow the path, described by semiotics

Data

Information



# Data processing and “Big Data”

Where's the challenge?

**Data engineering** and data processing are critical to support data science

But: data engineering is not data science

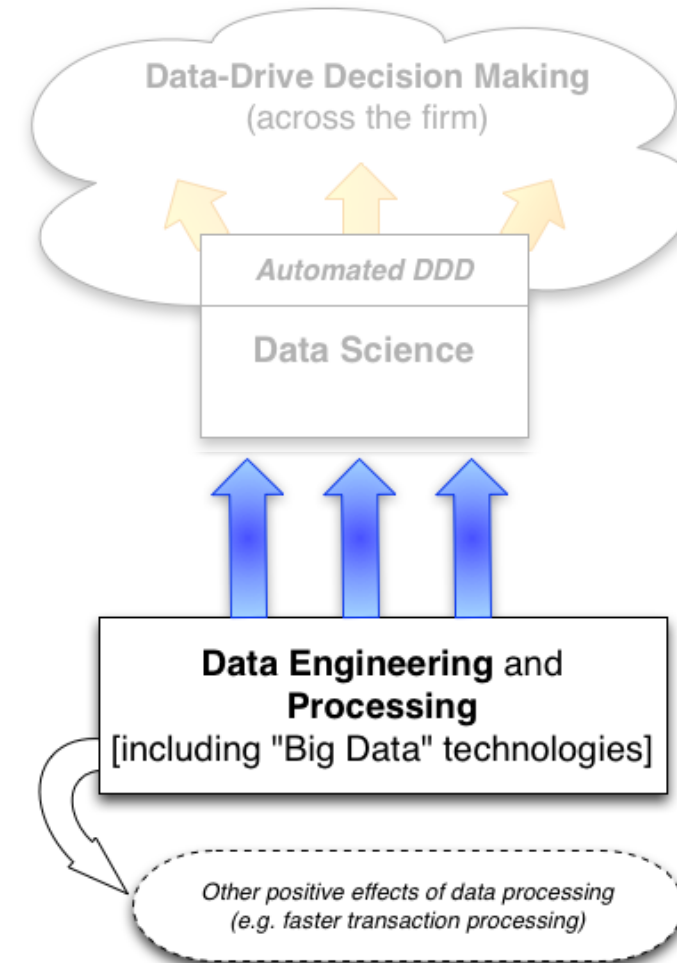
Data engineering handles for instance Data Warehouses, which

- collect data from operational databases
- accumulate **historical data**
- provide the basis for Business Intelligence applications

*Where is the catch?*

“**Big data**” means data sets that are too large for traditional data processing systems

Using big data technologies is associated with additional **productivity growth**



# Data Science vs. Data Mining

## Data science:

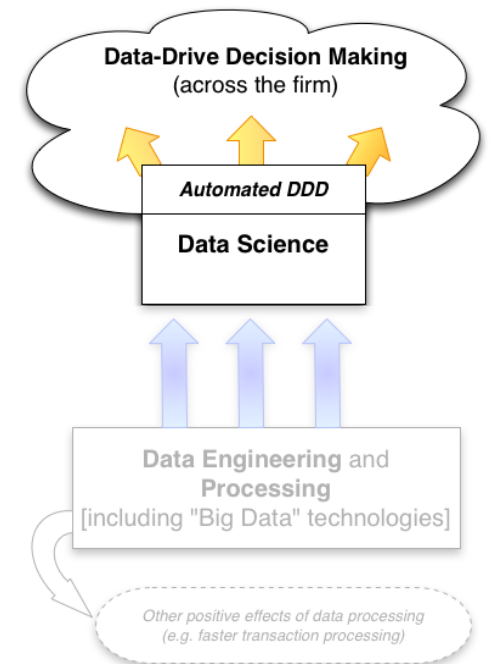
A set of *fundamental principles* that guide the extraction of knowledge from data

*“Data science is an interdisciplinary field aiming to turn data into real value [...] Value may be provided in the form of predictions, automated decisions, models learned from data, or any type of data visualization delivering insights. Data science includes data extraction, data preparation, data exploration, data transformation, storage and retrieval, computing infrastructures, various types of mining and learning, presentation of explanations and predictions, and the exploitation of results taking into account ethical, social, legal, and business aspects”*

(Van der Aalst [2016](#))

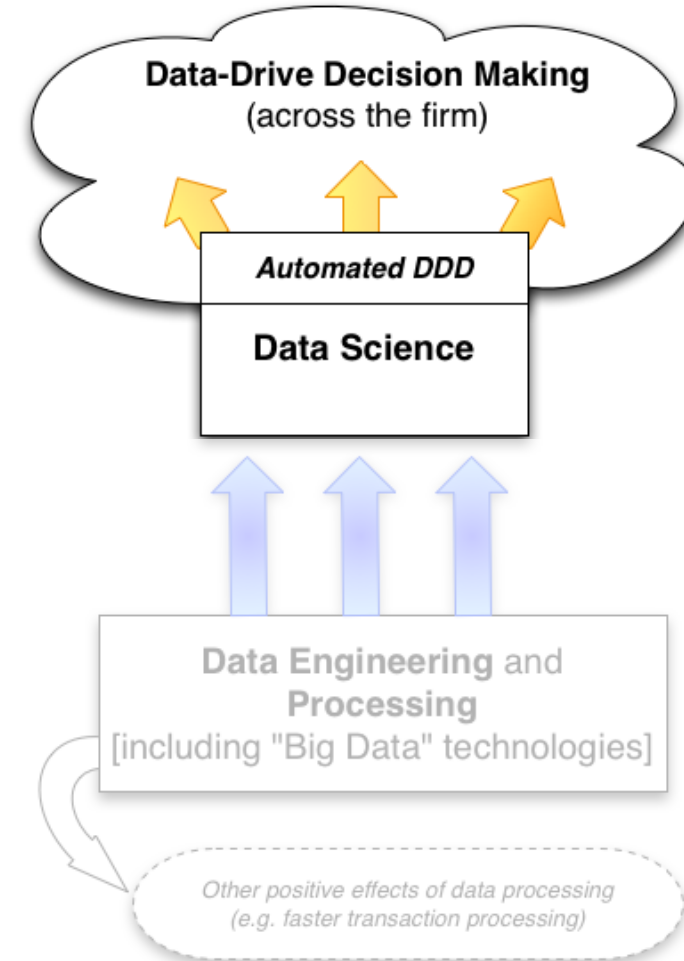
## Data mining:

Extraction of knowledge from data via tools/technologies that *incorporate* the principles



**In this class, we do both!**

- ... use **well-defined** stages for analysis
- ... find informative **descriptive attributes**
- ... be careful with **overfitting**
- ... think about the **context** when evaluating results
- ... data & data science capability as **strategic asset**
- ... *and many more*



# Example: Data & data science capability as strategic asset

- Data and the capability to make decisions from data are complementary assets.

1990s: Signet Bank aims at modeling of profitability of credit card customers.

However, there was only data for the terms they had offered in the past.

They conducted experiments in order to build predictive models from the data (charge-offs!).

Today one of the largest credit-card issuers.





# Some examples

In which fields do we use data for gaining competitive advantages?



## Marketing (Advertising)

Online advertising

Recommendations for cross-selling

Customer relationship management

## Production

Robotics and automation

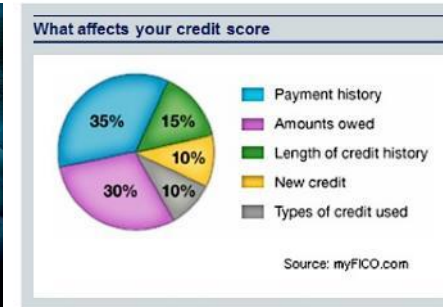


## Finance

Credit scoring and  
(high-frequency) trading

Fraud detection

Workforce management



## Retail

Supply Chain Management  
(Wal-Mart, Amazon etc.)



## Services

Voice-activated services

Human-machine communication  
(Siri, Alexa, Google Assistant)

Augmented Reality Services



## Fragen?

- ✓ Metaintroduction
- ✓ Decision Support and Business Intelligence
- ✓ Data-analytic thinking

1. Read short example about data & data science capability as strategic asset (case of Signet Bank)

[Kursmaterial > Readings/Übungen](#)

2. Read Goes (2014): How does the current acknowledgement of data as a valuable resource influence our research domain?

[Kursmaterial > Readings/Übungen](#)

## EDITOR'S COMMENTS

### Big Data and IS Research

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Publication history  
Frequency  
Ranking (Jourqual)  
Impact factor (2015)  
Cited half-life

1977 - present  
Quarterly  
A+  
5.384  
>10.0 years

3. Start your Python journey

[Kursmaterial > Readings/Übungen > Python Übungen > Jupyter Notebooks](#)

4. Remember your basics on data base design (including normalization)

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# Data - Why should companies care?

Data changes business models

*“[C]ompanies in the top third of their industry in the use of data-driven decision making were, on average, 5% more productive and 6% more profitable”*

(McAfee und Brynjolfsson, 2012, Result of 330 Interviews with Top-Managers and analysis of annual reports in Northern America)

