

## **Enterprise Al** -

Wertschöpfung in Unternehmen mit künstlicher Intelligenz

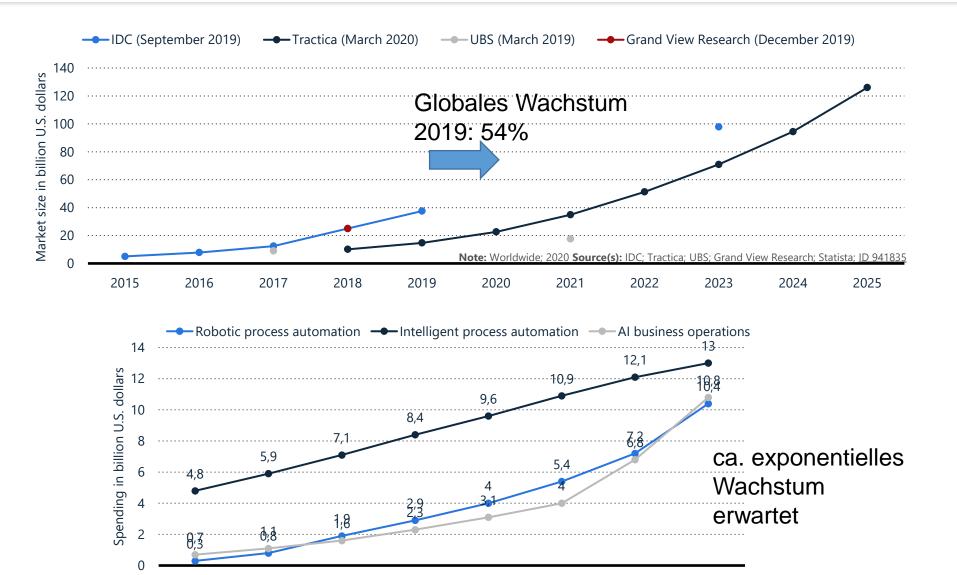
2. Thüringer KI-Forum, 7. Dez. 2020











2020\*

2021\*

2022\*

2023\*

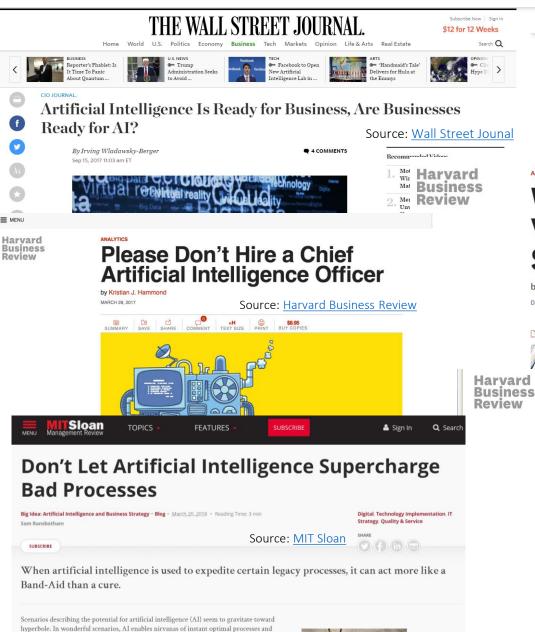
2017

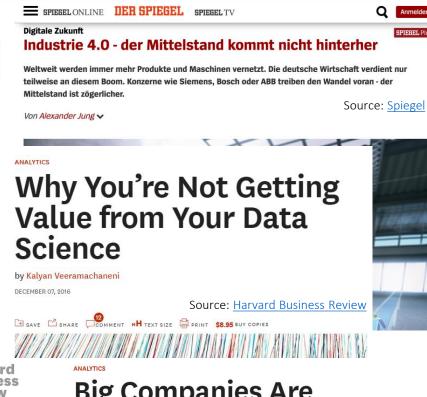
2018

2019

2016







Big Companies Are Embracing Analytics, But Most Still Don't Have a Data-Driven Culture

Source: Harvard Business Review

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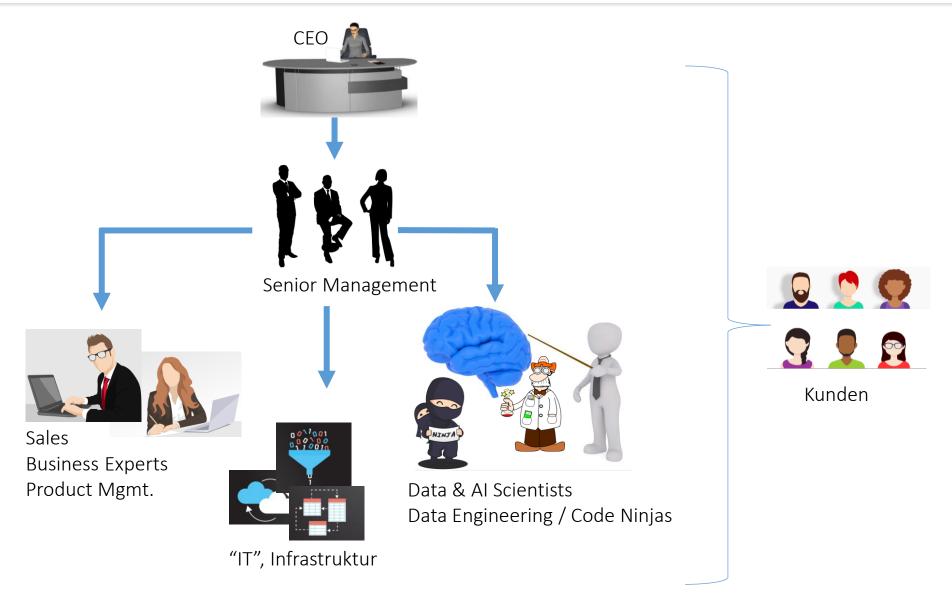
Source: Harvard Business Review

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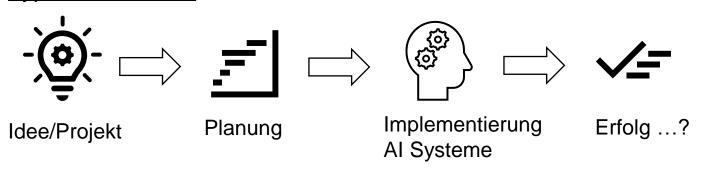


### Conway's law:

Any organization that designs a system [...] will produce a design whose structure is a copy of the organization's communication structure.

(Conway, 1968)

## **Typischer Ansatz:**

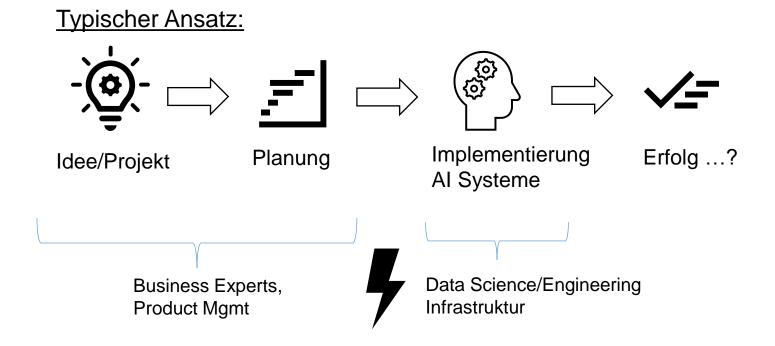




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(Datamation, 1968)







**Business Wissen** 

Datenverständnis/ Stochastik AI/Data Science Infrastruktur/ Operative Systeme











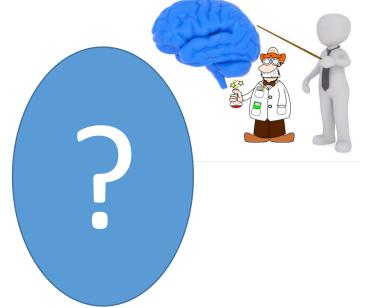
#### **Business Wissen**

## Datenverständnis/ Stochastik

## AI/Data Science

# Infrastruktur/ Operative Systeme





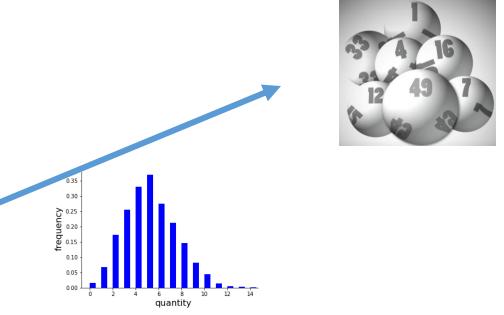




- Datenqualität als Aufgabe für jeden im Unternehmen:
   Daten Sammeln/Aufbereiten ca. 90% der täglichen Arbeit (Forbes)
- "Data Literacy" als Core Skill



## Zufälliges System – nicht vorhersagbar



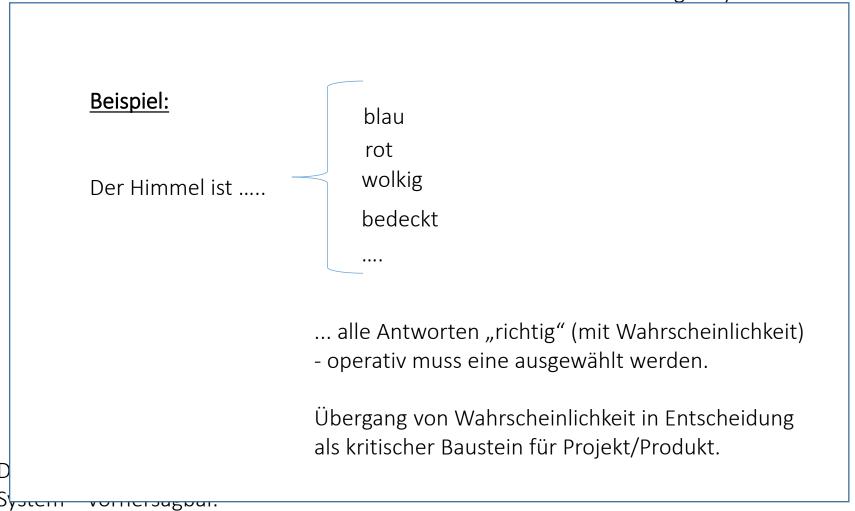
"Typische" Anwendung / Projekt Vorhersagbar, aber als Wahrscheinlichkeit.

→ Operative Entscheidung aus Wahrscheinlichkeit?

Deterministisches System – vorhersagbar. "Unsere Vorstellung"



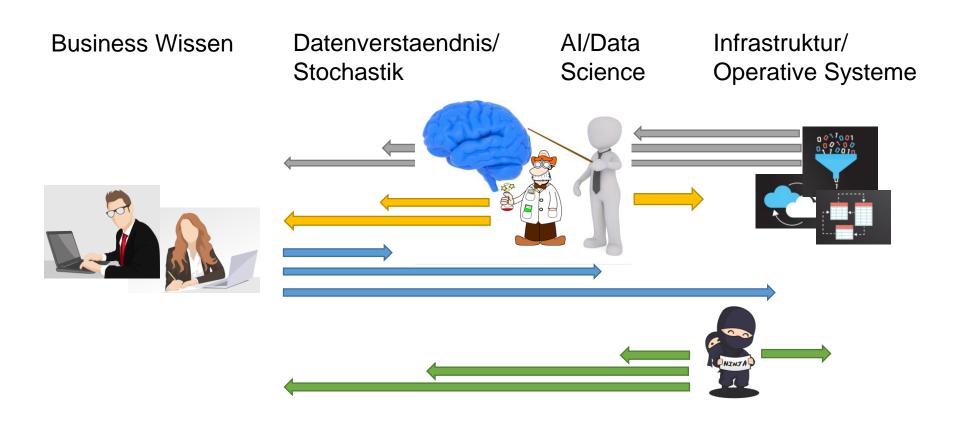
### Zufälliges System –



"Unsere Vorstellung"







Jedes Team muss mehr von den anderen verstehen, insb.

Bussiness : Was kann AI, was kann AI nicht? Bewertung von "Erfolg, Value"

Data/AI Scientists : Was sind Business Anforderung/Constraints? Bewertung von "Erfolg, Value"

Alle : Daten, Datenqualität



## Warum Gehen Projekte Schief?



### Erfolg

- Keine klaren Metriken
- Nur technische Metriken
- Metriken beziehen sich nicht auf "value".
- Optimieren des Al Modells kein Check auf Business Impact.



### Statistik / Stochastik

- Übergang Wahrscheinlichkeit (Prognose)
   → operative Entscheidung
- Zu wenig Kenntnis nicht-deterministischer Systeme (Business & AI)



#### Daten

- Datenqualität, insb. Inhaltliche Fehler (Domänenwissen)
- Infrastruktur / Zugang zu Daten



- AI Projekt als technisches Projekt
- Unzureichende Kopplung an Business
- "Value" nicht klar definiert



Al Modelle

- Keine Baseline
- Was ist realistisch erreichbar?
- Was ist "gut genug"?
- Schwer wartbarer Code
- Kein Produktivcode

Jupyter Notebook...



## Enterprise Al Canvas – Strukturierte Vorgehensweise zum Aufsetzen eines Al Projekts.

### **Business View**

- Value
- Decision & Optimization
- Success
- Domain
- Organization / Change
- Sponsor





Prediction & Action





Domains & Data Quality

Constraints

Data Source & Processing











Enterprise Al Canvas – Part 1: Business V	iew					
Decision & Optimization: How are decisions optimized? By whom?		Value: How does the use-case generate value? What can be offered? Which problem can be solved?		Success:  How is success (metrics, KPIs) defined? What makes the use-case "good" or "bad"?  Who decides this?		Evaluation von Al Use-Cases.
						Betrachte beides Business & Data Science/Al
Organization:			Dor	mains:		Dala Science/Ai
How are decisions & actions currently formed? How will this change			Whic	Which domain expertise is needed? Which business units are involved?		
with this use-case? Change Management? Additional training?						
	Enterprise Al Canvas – Part 2: Model & D				1	
Domains and Da		ata Quality:		Prediction & Action:	Features:	
'		ertise is needed? How and who can decide whether the		What should be predicted? How are	Which features are likely important?	
data are "good" or "bad"		bad"? How to measure & improve data quality?		predictions (as probabilities) transformed into actions?		
Sponsor: Which senior manager is responsible?						
	Data Sources & Processing:  Which data sources are required, which are already available? Which types of data are needed? Which systems are required and already available to handle		Constraints:			
			es of			e for a prediction? Which granularity is required?
			andle		How often does the model need to be re-trained? How are the models served?	
	the data?				Edge, on-premise or Cloud?	
	Evaluation & Monitoring: Which metrics are used? How will the project be monitored? What happens in case of incidents?					



 Artificial Intelligence ist einer der Wachstumsmotoren der Zukunft.



- Herausforderungen an Unternehmen
  - KI Projekte sind nicht "nur" technische Projekte
  - Business Experten müssen mehr von KI und Statistik lernen
  - KI Experten müssen mehr von Unternehmensprozessen und –strukturen lernen.
- Enterprise Al Canvas
  - Strukturierte Methodik, um neue KI Projekte zu evaluieren.
  - Schnelle Auswertung : technischer Machbarkeit & betrieblicher Umsetzung