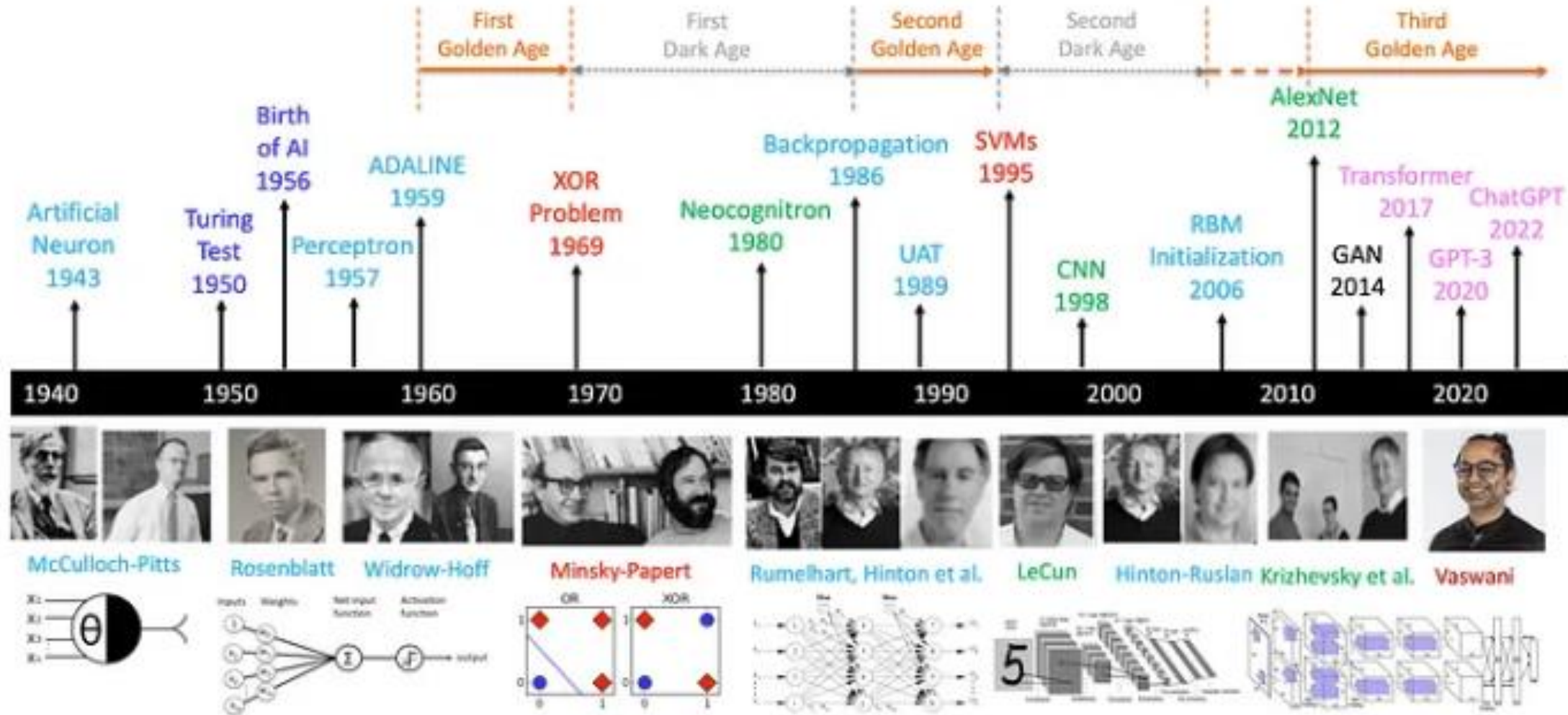


# INTRODUCTION TO MACHINE LEARNING

# Is Machine Learning new?



A Brief History of AI with Deep Learning, LM Po

# How AI is changing the world today

- **Economics:** macroeconomics forecasting, market analysis
- **Finance:** fraud detection, algorithmic trading
- **Retail:** recommendation engines, demand forecasting
- **Transportation:** autonomous vehicles, traffic management
- **Energy:** optimizing wind and solar power generation, smart home systems

# Future employment in the era of AI

## World Economic Forum's Future of Jobs Report 2025: Fastest growing and declining jobs by 2030

↑ Top fastest growing jobs	↓ Top fastest declining jobs
1 Big data specialists	1 Postal service clerks
2 FinTech engineers	2 Bank tellers and related clerks
3 AI and machine learning specialists	3 Data entry clerks
4 Software and applications developers	4 Cashiers and ticket clerks
5 Security management specialists	5 Administrative assistants and executive secretaries
6 Data warehousing specialists	6 Printing and related trades workers
7 Autonomous and electric vehicle specialists	7 Accounting, bookkeeping and payroll clerks
8 UI and UX designers	8 Material-recording and stock-keeping clerks
9 Light truck or delivery services drivers	9 Transportation attendants and conductors
10 Internet of things specialists	10 Door-to-door sales workers, news and street vendors, and related workers
11 Data analysts and scientists	11 Graphic designers
12 Environmental engineers	12 Claims adjusters, examiners and investigators
13 Information security analysts	13 Legal officials
14 DevOps engineers	14 Legal secretaries
15 Renewable energy engineers	15 Telemarketers

# AI vs DATA SCIENCE vs MACHINE LEARNING

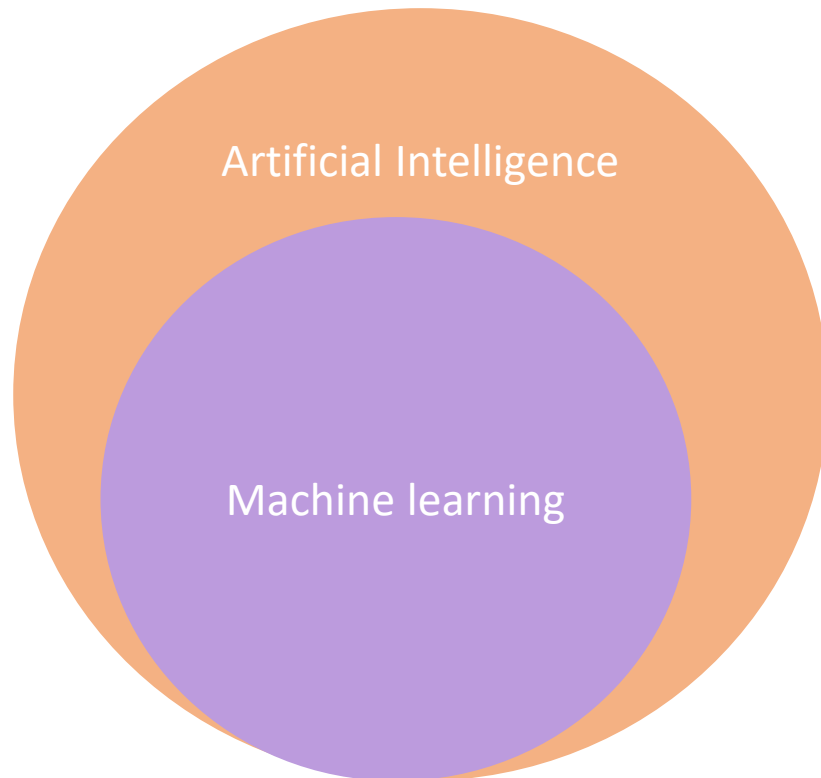
# What is Artificial Intelligence



Artificial Intelligence

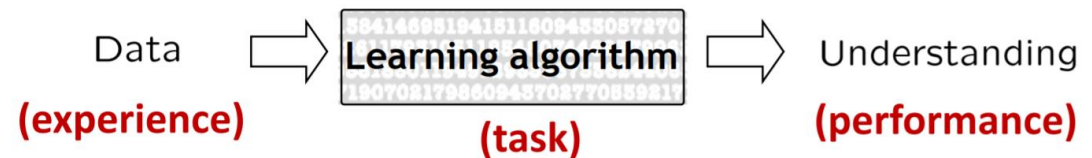
Ability of a computer to perform tasks  
commonly associated with intelligent being

# What is Machine Learning

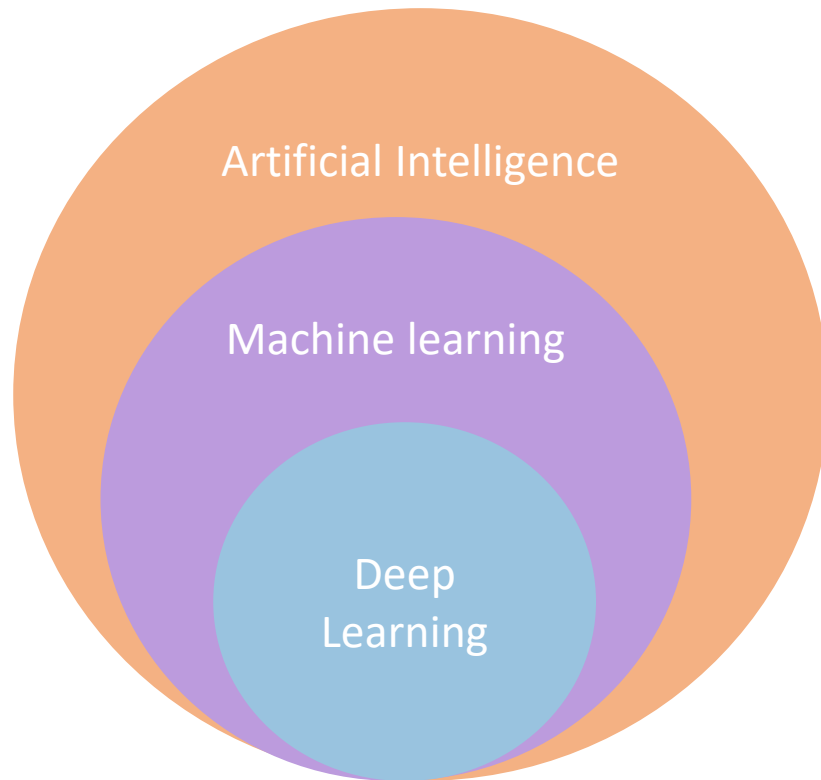


Science and art of programming computers so they can *learn from data*.

Study of algorithms that improve their performance at some task with experience.



# What is Deep Learning



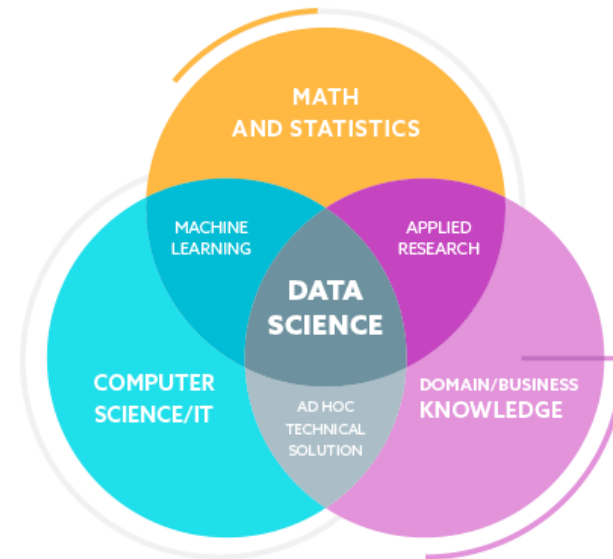
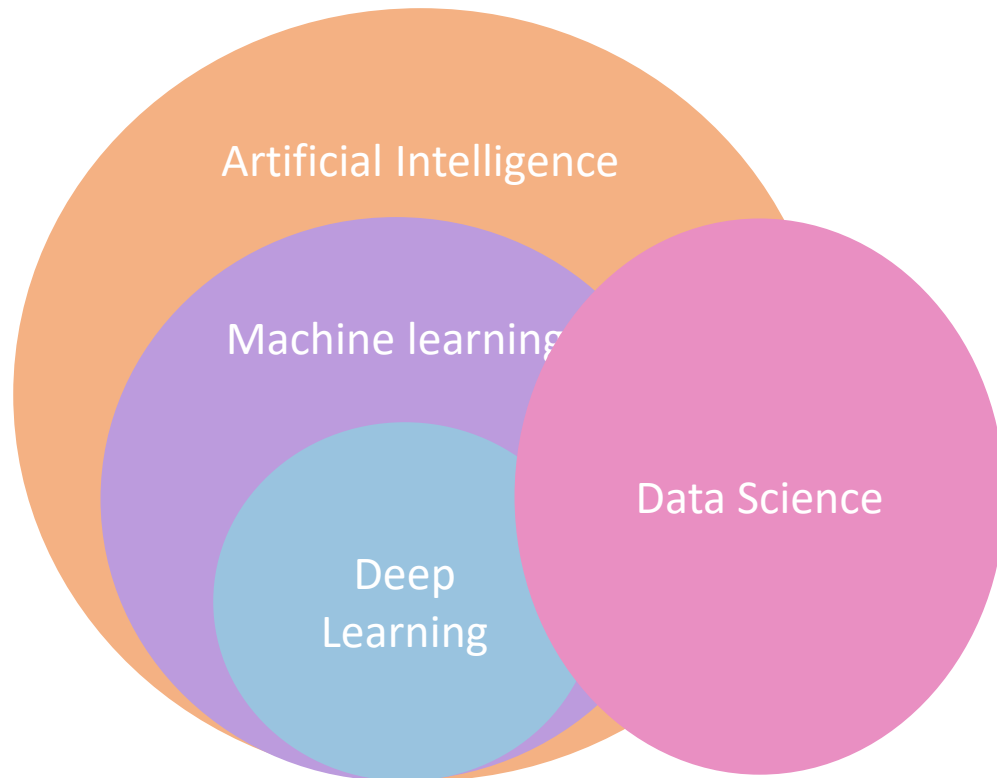
In comparison to ML, Deep Learning:

- can solve more complex problems
- is more difficult to implement
- requires more time to train the model
- requires more computational resources.



# What is Data Science

Set of fundamental principles that guide the extraction of knowledge from data.



# Key Takeaways: Difference between Data Science, Machine Learning and AI

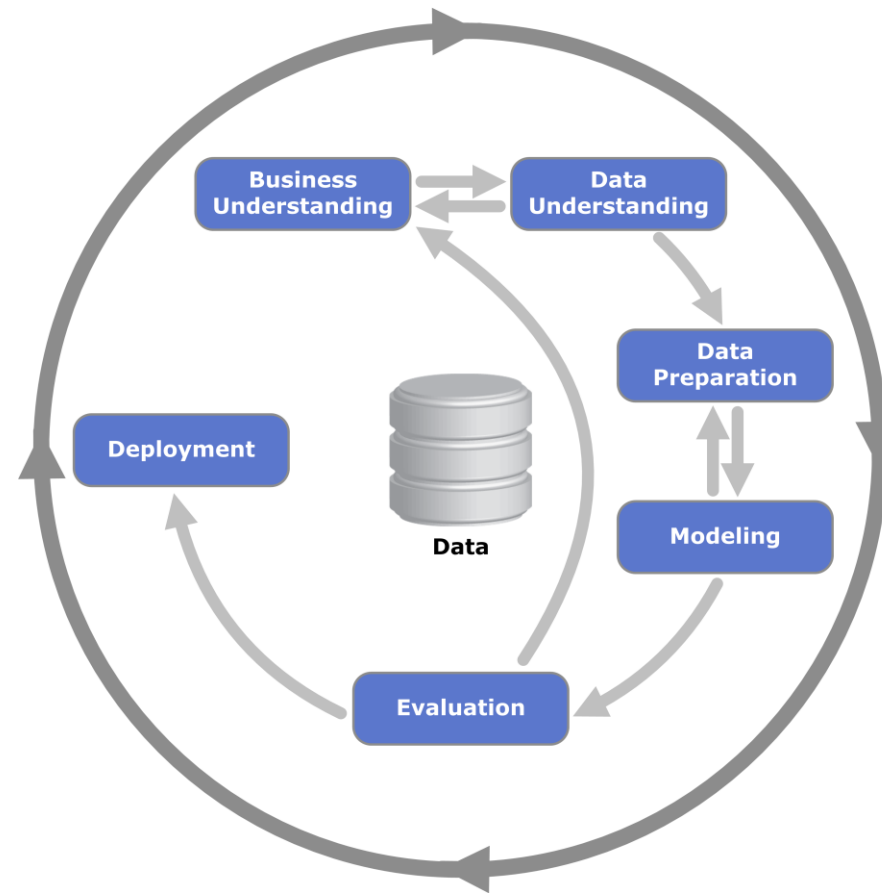
The fields do have a great deal of overlap, but they are **not interchangeable**.

- 1 Data science produces actionable **insights**.
- 2 Machine learning produces **predictions**.
- 3 Artificial intelligence produces **actions**.

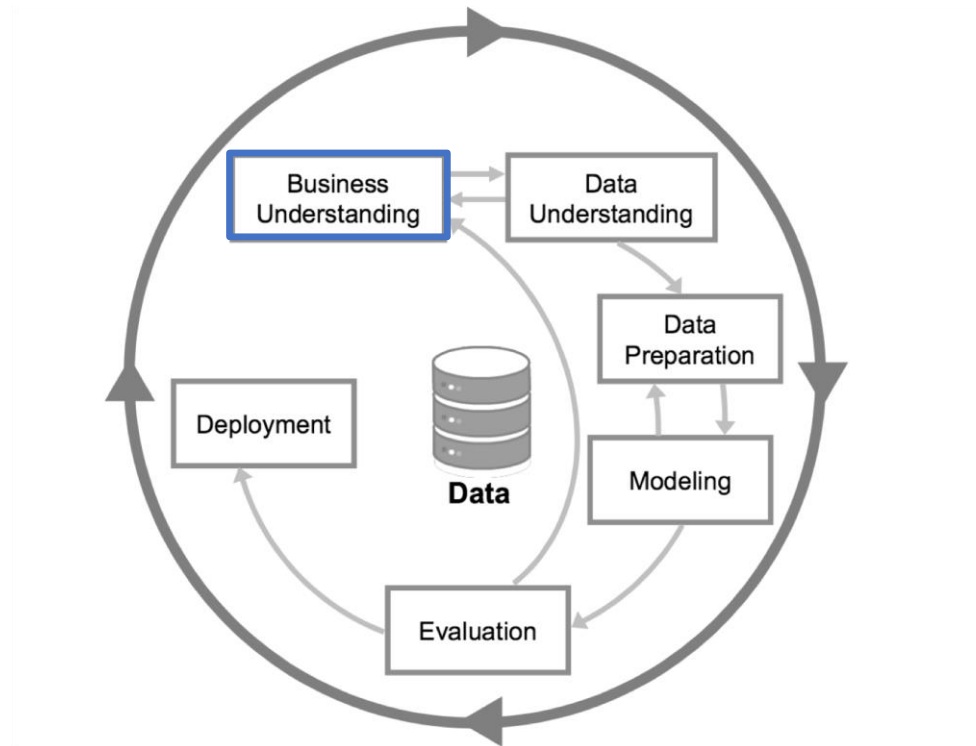
# HOW AND WHEN TO USE ML FOR BUSINESS

# Data mining process

Cross Industry Standard Process for Data Mining (CRISP-DM; Shearer, 2000)

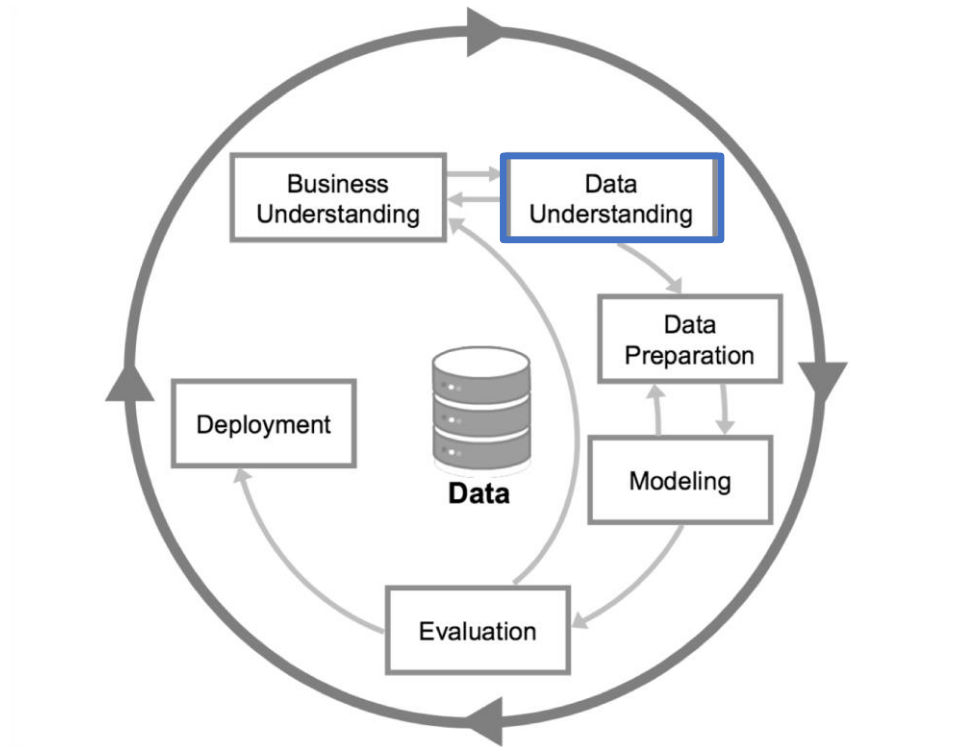


# Business understanding



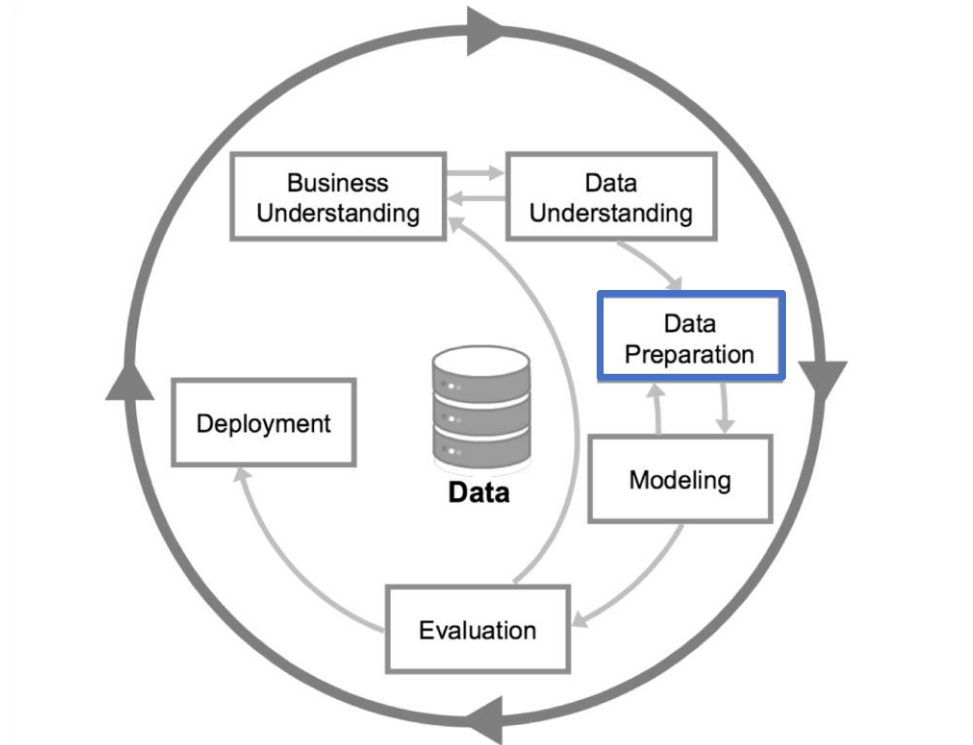
- Identify our business goals
- Assess our situation
- Define our data science goals
- Produce our project plan

# Data understanding



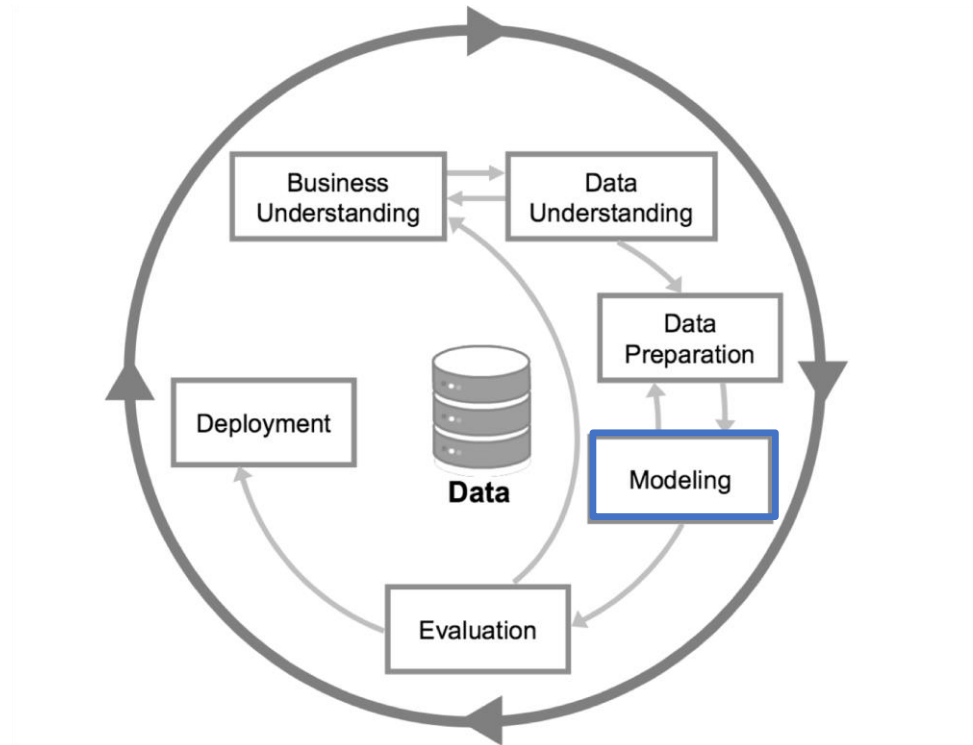
- Collect initial data
- Describe data
- Explore data
- Verify data quality

# Data preparation



- Data preparation/curation/preprocessing
  - Invalid values
  - Formats
  - Uniqueness
  - Missing values
  - Misspellings

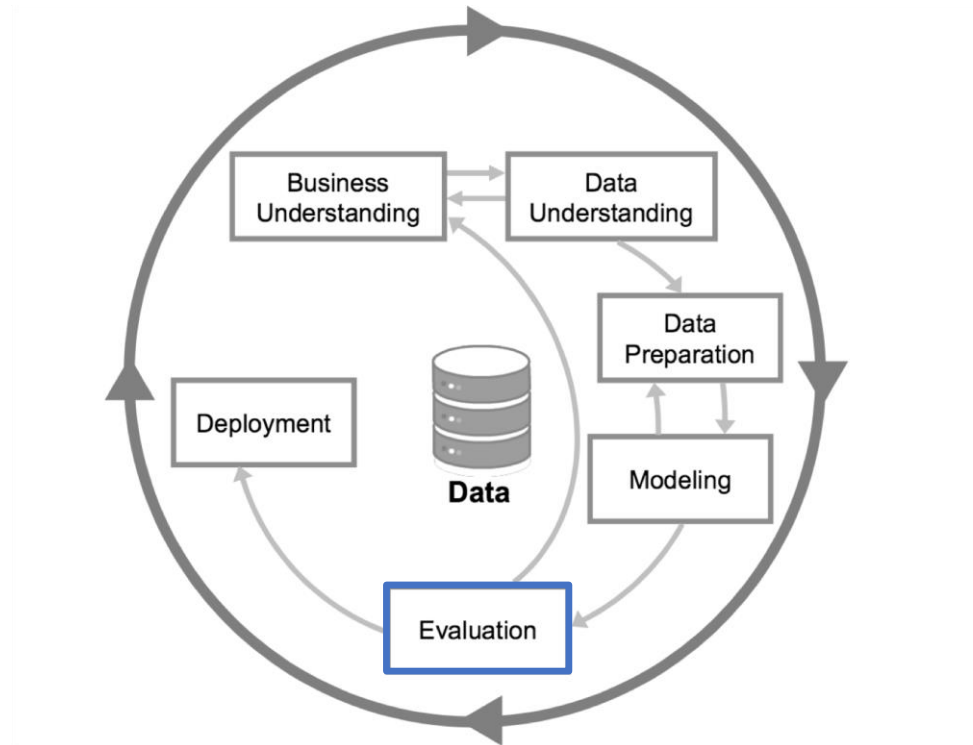
# Modeling



Build a model that captures regularities of data.

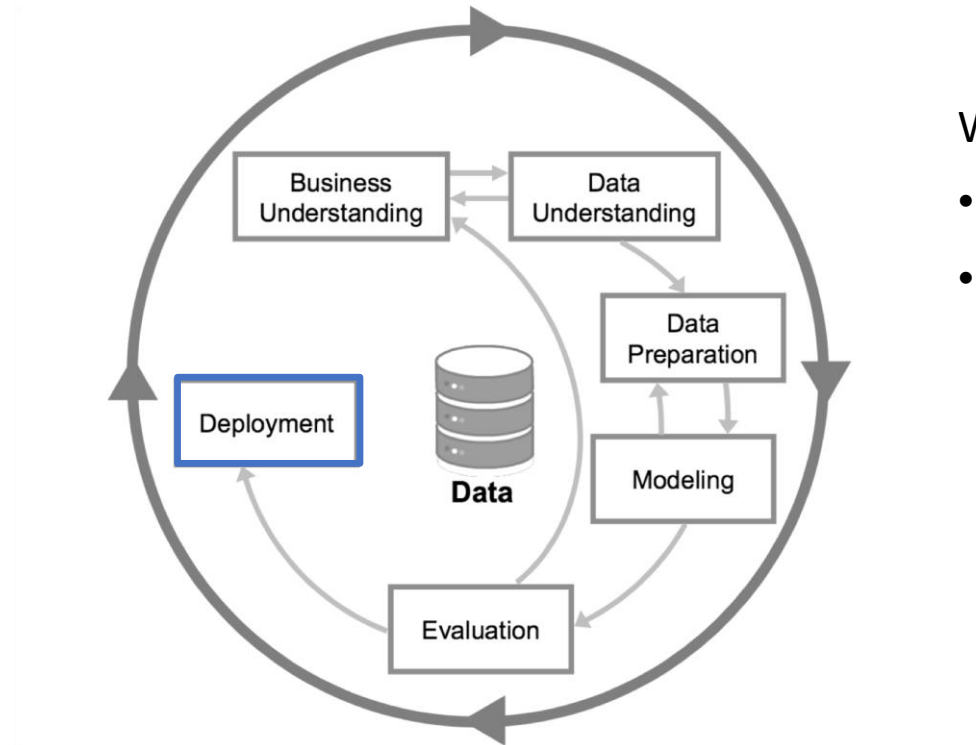


# Model evaluation



- Evaluate performance
- Make sure that the model satisfies initial business goals
- Interpret results

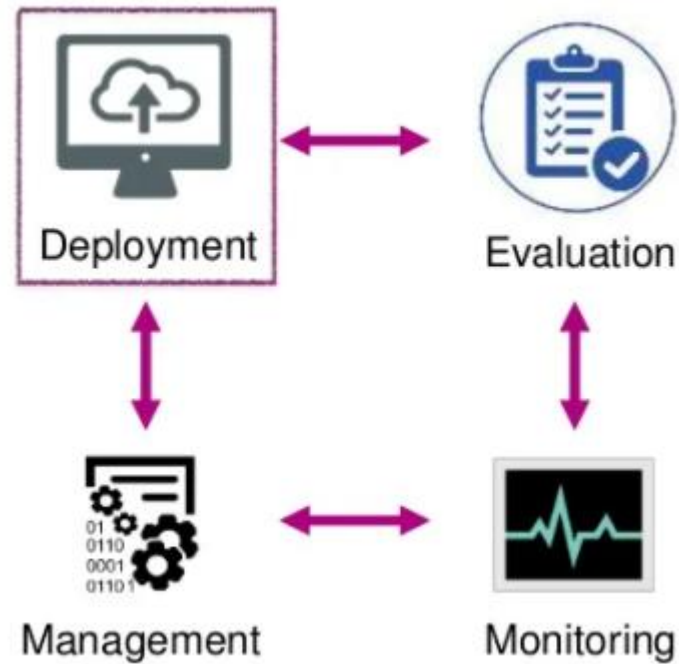
# Model deployment



We might need to solve a business problem:

- Once (Specific campaign)
- Regularly (Monthly campaign of contacting customers). Here, we need to integrate the machine learning model into a production environment.

## After model deployment



**MLOps** is a set of practices that aims to deploy and maintain machine learning models in production reliably and efficiently.

<https://ml-ops.org/content/mlops-principles>

# When should you use ML



Can you formulate your problem clearly?



Do you need to automate your task?



Do you have sufficient data?



Does your problem have a regular pattern?



How do you define success?

# When should you use ML: example of sentiment analysis



Can you formulate your problem clearly?

Given a customer review, infer its sentiment:

- Input: customer review text
- Output: positive, negative, neutral



Do you need to automate your task?

Too many review to read.



Do you have sufficient data?

Millions of customer reviews and ratings from the Web.



Does your problem have a regular pattern?

High rated reviews more positive words (*good, awesome*)  
Low rated reviews more negative words (*bad, disappointed*)

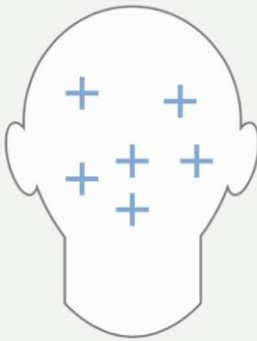


How do you define success?

Percentage of reviews with correctly inferred sentiment.

## More examples of when to use ML

*Hand-written rules and equations are too complex—as in face recognition and speech recognition.*



*The rules of a task are constantly changing—as in fraud detection from transaction records.*



*The nature of the data keeps changing, and the program needs to adapt—as in automated trading, energy demand forecasting, and predicting shopping trends.*

