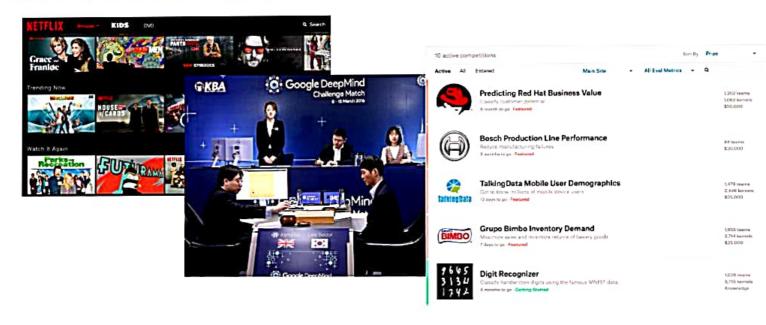
# Machine learning model deployment with ibm cloud watson studio

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# Machine Learning (ML)

- ML is a branch of artificial intelligence:
  - Uses computing based systems to make sense out of data
    - Extracting patterns, fitting data to functions, classifying data, etc
  - ML systems can learn and improve
    - · With historical data, time and experience
  - Bridges theoretical computer science and real noise data.

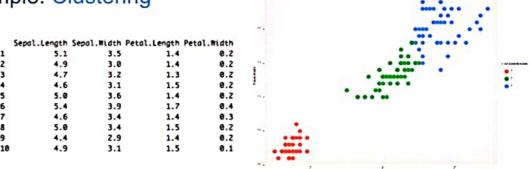
# ML in real-life



### Supervised and Unsupervised Learning

- Unsupervised Learning
  - There are not predefined and known set of outcomes
  - · Look for hidden patterns and relations in the data



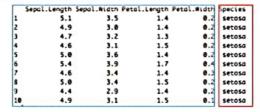


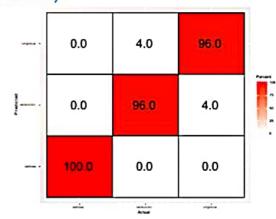
## Supervised and Unsupervised Learning

- Supervised Learning
  - For every example in the data there is always a predefined outcome
  - Models the relations between a set of descriptive features and a target (Fits data to a function)
  - 2 groups of problems:
    - Classification
    - Regression

#### Supervised Learning

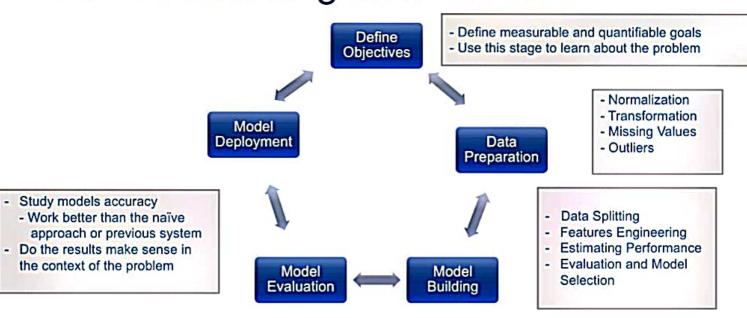
- Classification
  - Predicts which class a given sample of data (sample of descriptive features) is part of (discrete value).





- Regression
  - Predicts continuous values.

#### Machine Learning as a Process



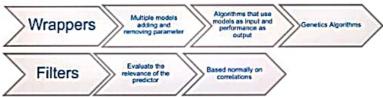
# ML as a Process: Data Preparation

- Needed for several reasons
  - Some Models have strict data requirements
    - · Scale of the data, data point intervals, etc
  - Some characteristics of the data may impact dramatically on the model performance
- Time on data preparation should not be underestimated



#### ML as a Process: Feature engineering

- Determine the predictors (features) to be used is one of the most critical questions
- Some times we need to add predictors
- Reduce Number:
  - Fewer predictors more interpretable model and less costly
  - Most of the models are affected by high dimensionality, specially for non-informative predictors



Binning predictors

#### ML as a Process: Model Building

- Data Splitting
  - Allocate data to different tasks
    - model training
    - · performance evaluation
  - Define Training, Validation and Test sets
- Feature Selection (Review the decision made previously)
- Estimating Performance
  - Visualization of results discovery interesting areas of the problem space
  - · Statistics and performance measures
- Evaluation and Model selection
  - · The 'no free lunch' theorem no a priory assumptions can be made
  - Avoid use of favorite models if NEEDED

