Machine Learning

B. N. M. Institute of Technology

Department: Artificial Intelligence and Machine Learning (AI)

Course: Machine Learning (18Al61) effective from the academic year 2018–2019 under VISVESVARAYA TECHNOLOGICAL UNIVERSITY, BELAGAVI

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Machine Learning Landscape

What is Machine Learning?

Few Examples

Spam Email Filter Re

Recommender

Optical Character Recognition

Chatbots

Voice Recognition

Facial Recognition

Forecasting

Autonomous Vehicle

Medical Diagnostics

Fraud Detection

and many more...

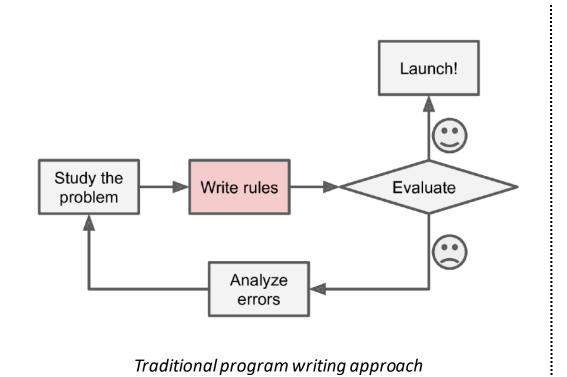
Some Definitions

"Science of programming computer to learn from data

"field of study that gives computers the ability to learn without being explicitly programmed

"A computer program is said to learn from experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, improves with experience E

Why is Machine Learning Required?



Study the problem

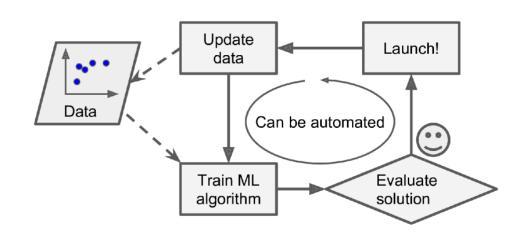
Analyze errors

Launch!

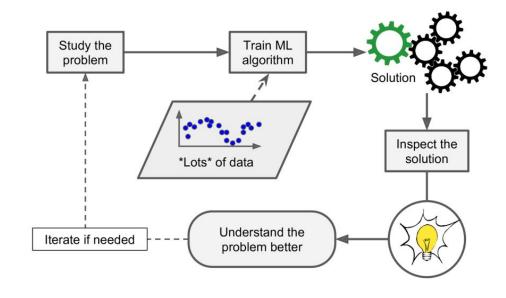
Evaluate solution

The Machine Learning approach

Why is Machine Learning Required? (Cont.)



Automatically adapting change



Discovering patterns for human learning

Machine Learning is a Good Option for

- Problems that require a lot of fine-tuning and long list of rules
- Problems that traditional approaches do not yield good results
- Problems where data distribution changes over time
- Getting insights about complex problems from large volume of data

Types of Machine Learning Systems

Whether learning requires human supervision

- Supervised Learning
- Semi-supervised Systems
- Unsupervised Learning
- Reinforcement Learning

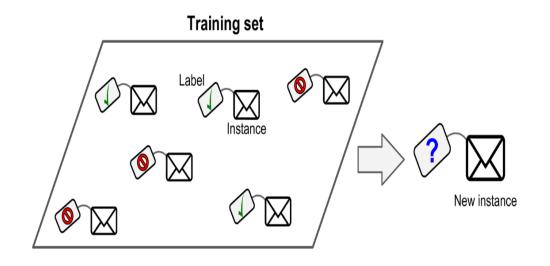
Whether learning is incremental

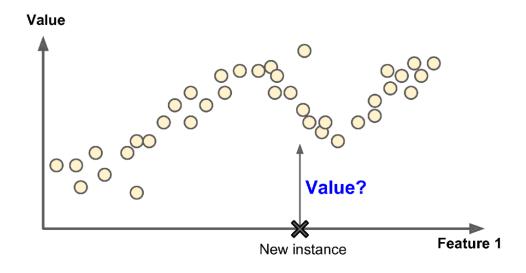
- Batch Learning
- Online Learning

Whether learning is to compare or to detect pattern and/or predict

- Instance-based Learning
- Model-based Learning

Classification & Regression in Supervised Learning

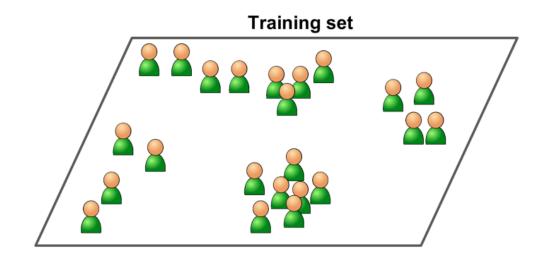


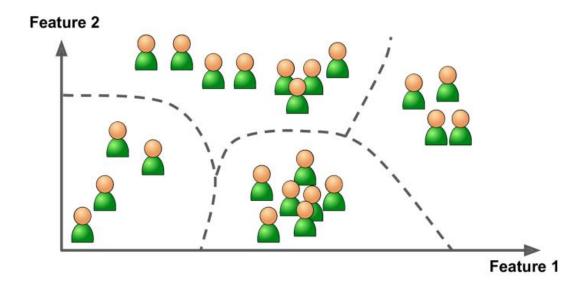


A labeled training set for spam classification

Predicting a continuous value in regression problem

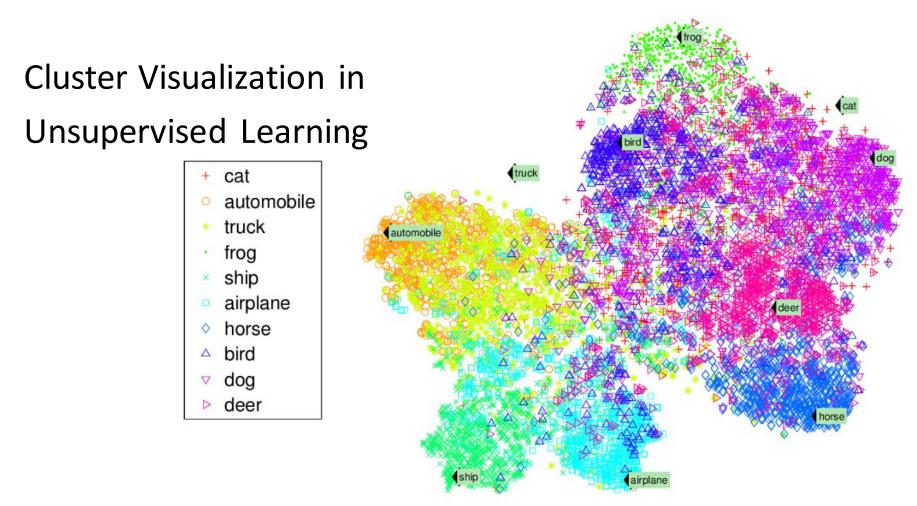
Clustering in Unsupervised Learning





An unlabeled training set for clustering

Detecting groups of similar visitor in clustering



Cluster visualization in unsupervised learning

Dimensionality Reduction in Unsupervised Learning

- Simplifying data without losing too much information
- Merging correlated features into one
- Feature extraction

Anomaly and Novelty Detection in Unsupervised Learning



Detection of an anomaly

Association Rule Learning in Unsupervised Learning

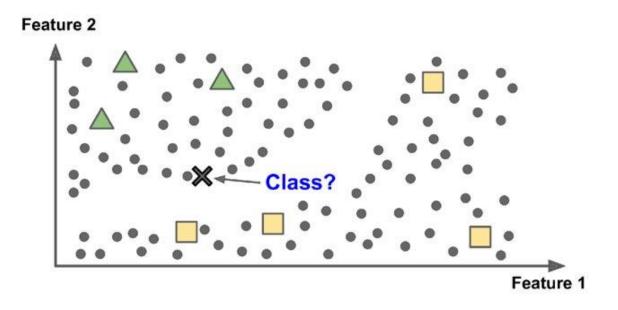
{Potatoes, Onions} => {Burger}

{Barbeque sauce, Potato chips} => {Steak}

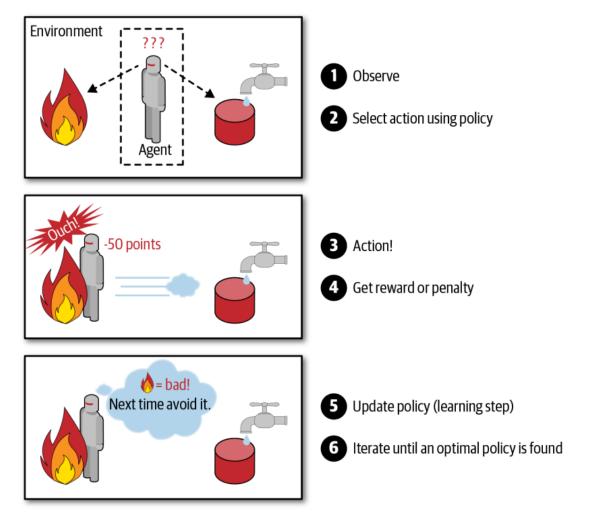


Results in Promotional Pricing and/or Appropriate Product Placement

Classification in Semisupervised Learning

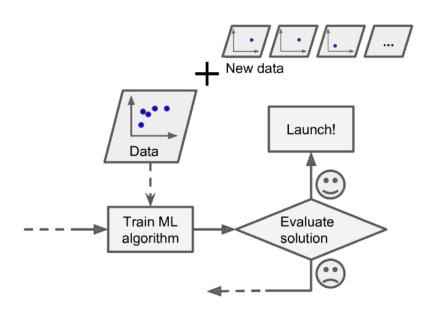


Building Strategy in Reinforcement Learning

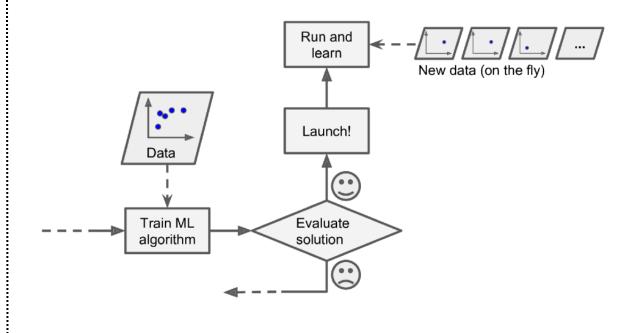


Receiving reward or penalty against its actions and fine-tuning its strategy

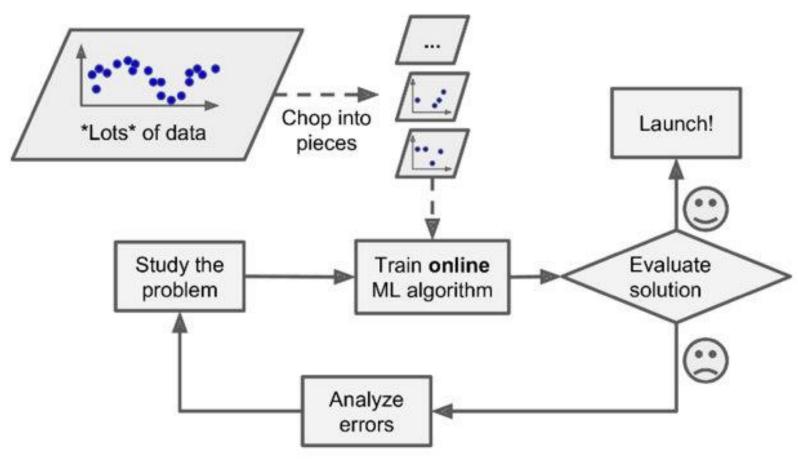
Batch and Online Learning



Relearning from scratch on full dataset (old and new data) in Batch Learning

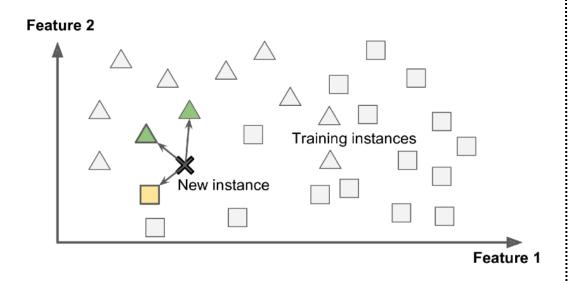


Learning is incremental for new data in Online Learning

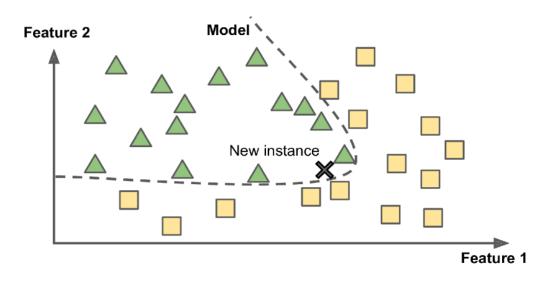


Handling huge dataset in Online Learning

Instance-based or Model-based Learning systems on how they generalized

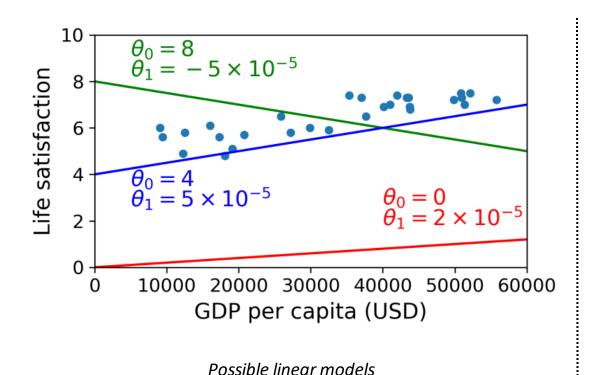


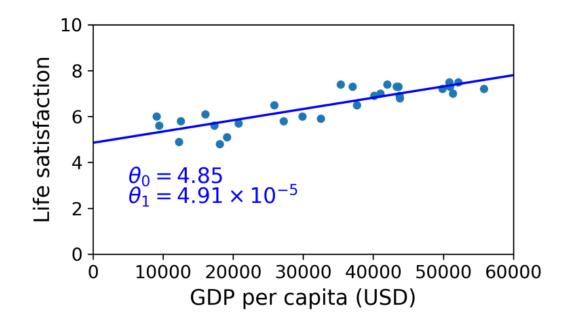
Instance-based Learning is by heart and it then generalizes to new cases by using similarity measures



Model is created in Model-based Learning and then it make predictions

Fitting Model to Data in Model-based Learning

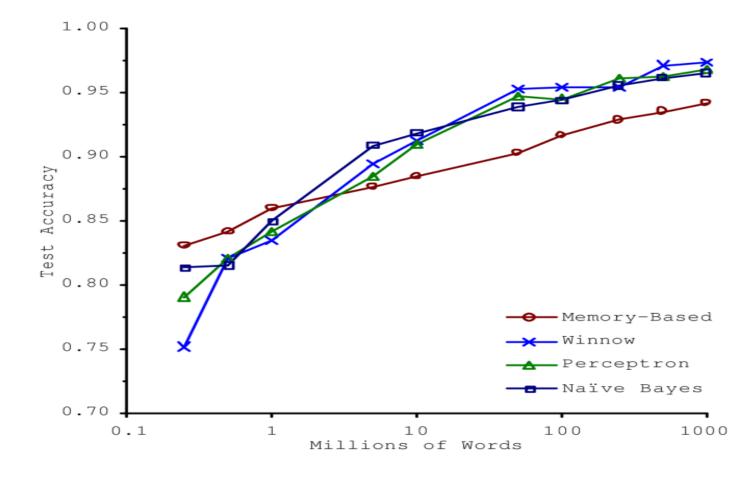




Fitted linear model

Main Challenges in Machine Learning

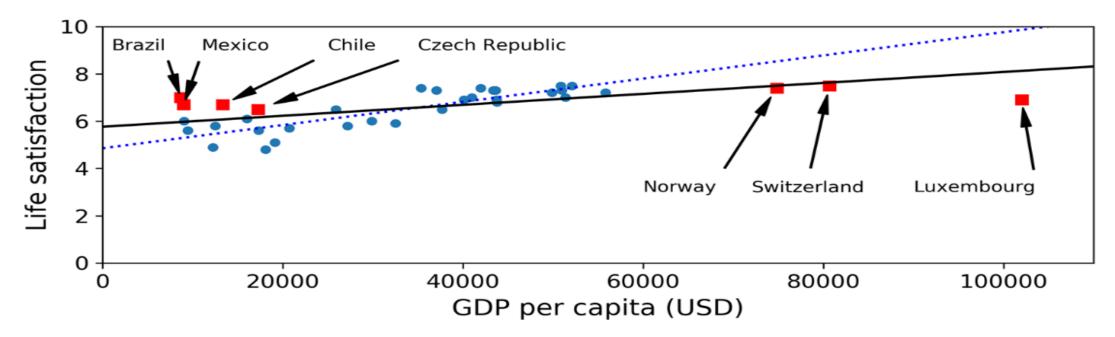
Insufficient Quantity of Training Data



Trade-off between spending time and money on algorithm development and spending these on corpus development

Nonrepresentative Training Data

- Sampling Noise
- Sampling Bias



A better model fitted over more representative training data

Poor-Quality Data

- Errors
- Missing values
- Outliers

Irrelevant Features

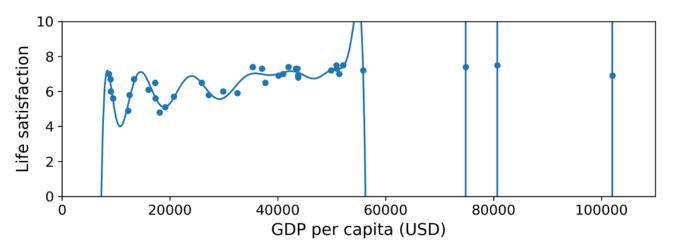
Applying Feature Engineering involving the following steps

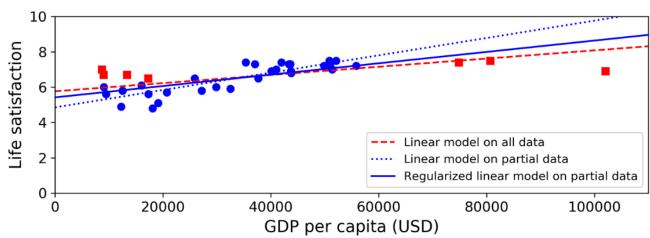
- Feature selection
- Feature extraction
- Creating new features

Overfitting the Training Data

Resolving by

- Simplifying model
- Applying constraints (regularization)
- Gathering more training data
- Fixing data quality issues



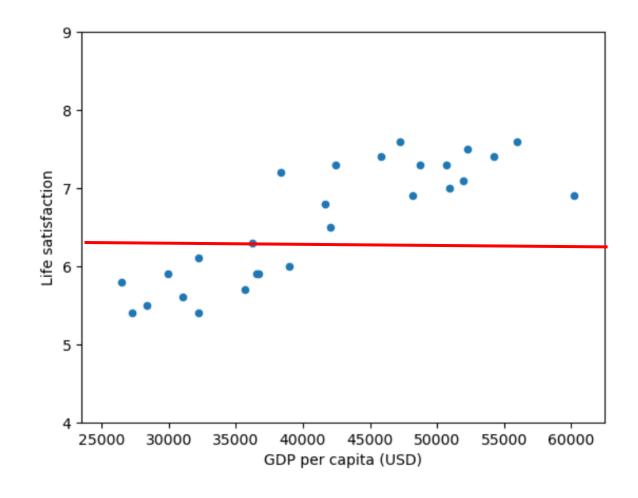


Overfitting and applying regularization to avoid it

Underfitting the Training Data

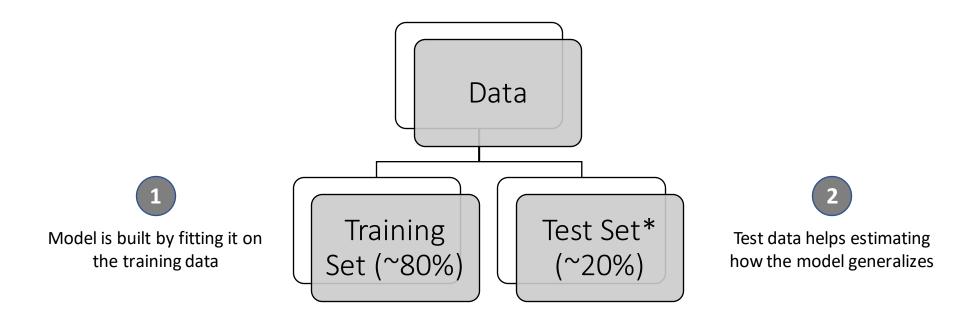
Resolving by

- Selecting more powerful model
- Feeding better features
- Reducing constraints or regularization



Testing and Validating

Ensuring Model Generalizes Well

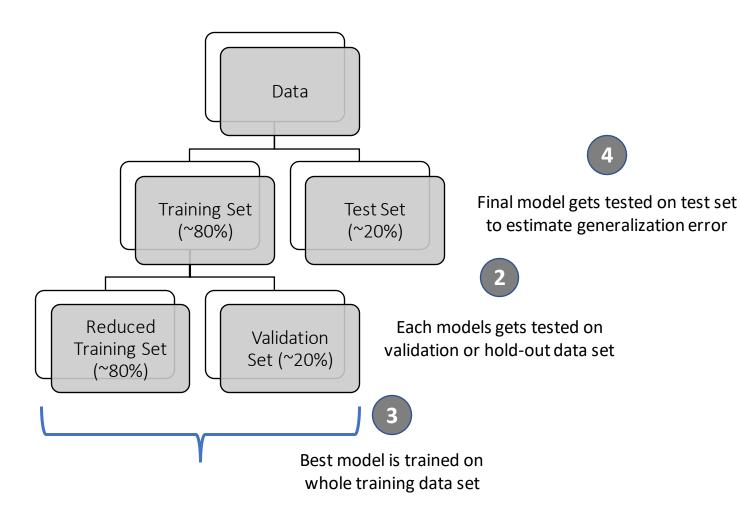


Testing and Validating (Cont.)

- Hyperparameter Tuning
- Model Selection
- Validation
- Cross-validation

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Multiple models with different hyperparameters are created on reduced training data set



No Free Lunch (NFL) Theorem

- No model that is a priori guarantees to be a better one
- Evaluating all models is only way to know which one works best Not practical
- Hence, making reasonable assumption about data and evaluating few reasonable models is better option

The Pipeline

