



Brain Break

16.55



CSE/STAT 416

Course Wrap Up

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University of Washington
Aug 19, 2019

Slides borrowed from Emily Fox

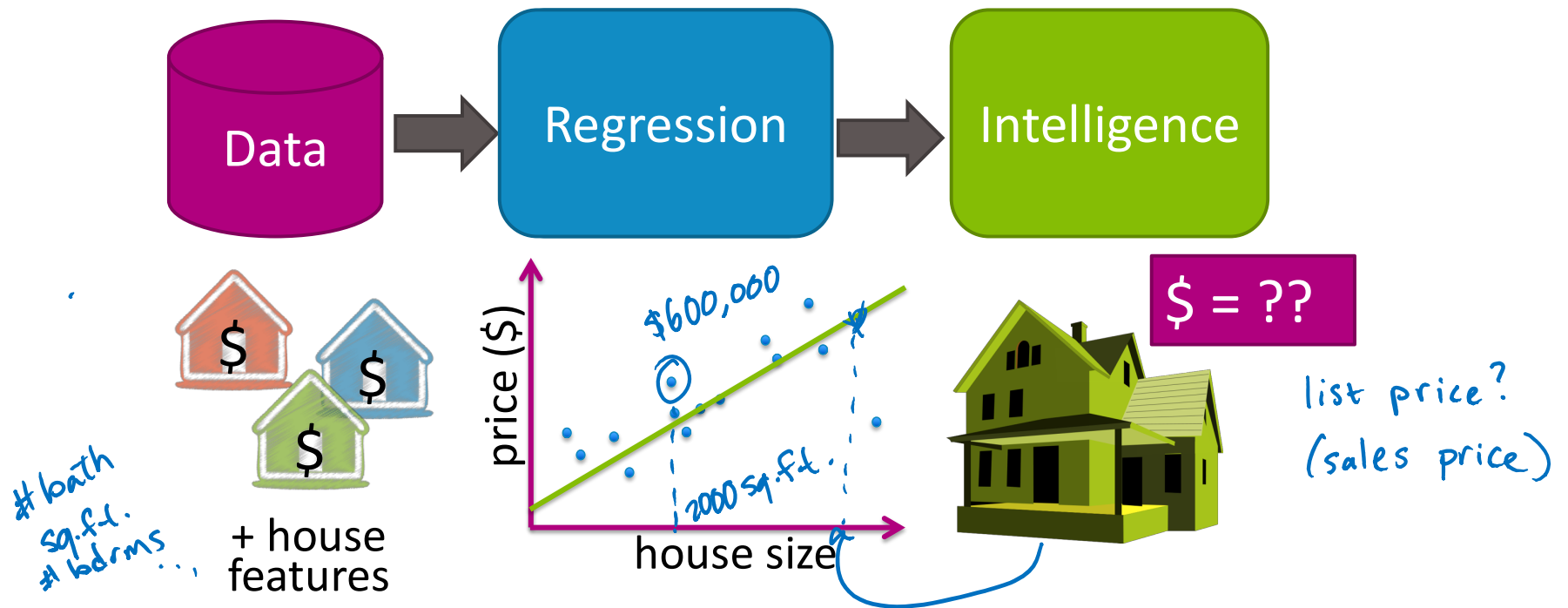


One Slide

- Regression
- Overfitting
- Training, test, and generalization error
- Bias-Variance tradeoff
- Ridge, LASSO
- Cross validation
- Gradient descent
- Classification
- Logistic regression
- Decision trees
- Boosting
- Precision and recall
- Nearest-neighbor retrieval, regression, and classification
- Kernel regression
- Locality sensitive hashing
- Dimensionality reduction, PCA
- k-means clustering
- Hierarchical clustering
- Unsupervised v. supervised learning
- Recommender systems
- Matrix factorization
- Coordinate descent
- Neural networks
- Convolutional neural networks
- Transfer learning for deep learning



Case Study 1: Predicting house prices



Regression

Case study: Predicting house prices

$$y_i = f(x_i) + \epsilon_i \quad \hat{y}_i = \hat{f}(x_i)$$

Models

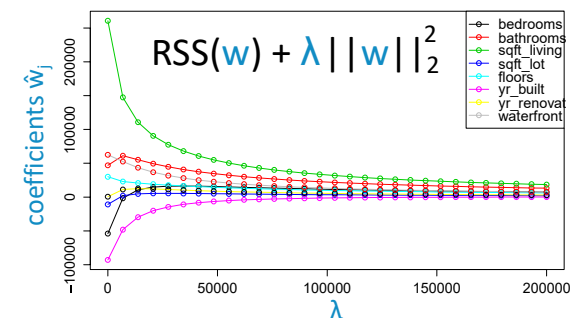
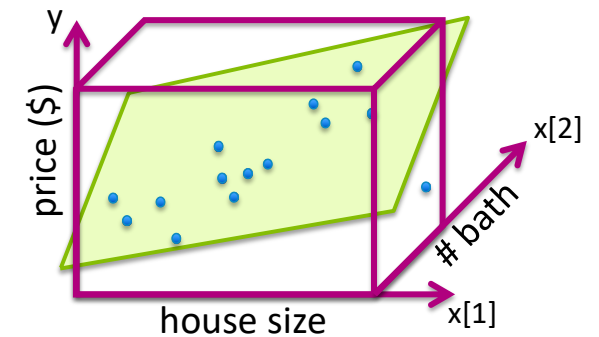
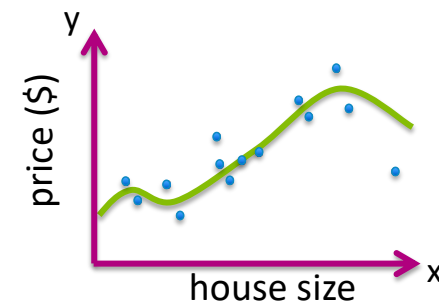
- Linear regression
- Regularization: Ridge (L2), Lasso (L1)

Including many features:

- Square feet
- # bathrooms
- # bedrooms
- Lot size
- Year built
- ...



LASSO \Rightarrow Sparse



Regression

Case study: Predicting house prices

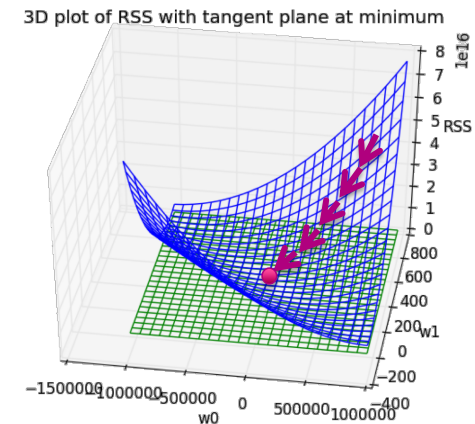
$$RSS(w) = \sum_{i=1}^n (w^T h(x_i) - y_i)^2$$

Algorithms

- Gradient descent

$$\begin{aligned} RSS(w_0, w_1) = & (\$_{\text{house 1}} - [w_0 + w_1 \text{sq.ft.}_{\text{house 1}}])^2 \\ & + (\$_{\text{house 2}} - [w_0 + w_1 \text{sq.ft.}_{\text{house 2}}])^2 + \\ & (\$_{\text{house 3}} - [w_0 + w_1 \text{sq.ft.}_{\text{house 3}}])^2 + \dots \\ & \text{[include all houses]} \end{aligned}$$

↓
 \hat{w}

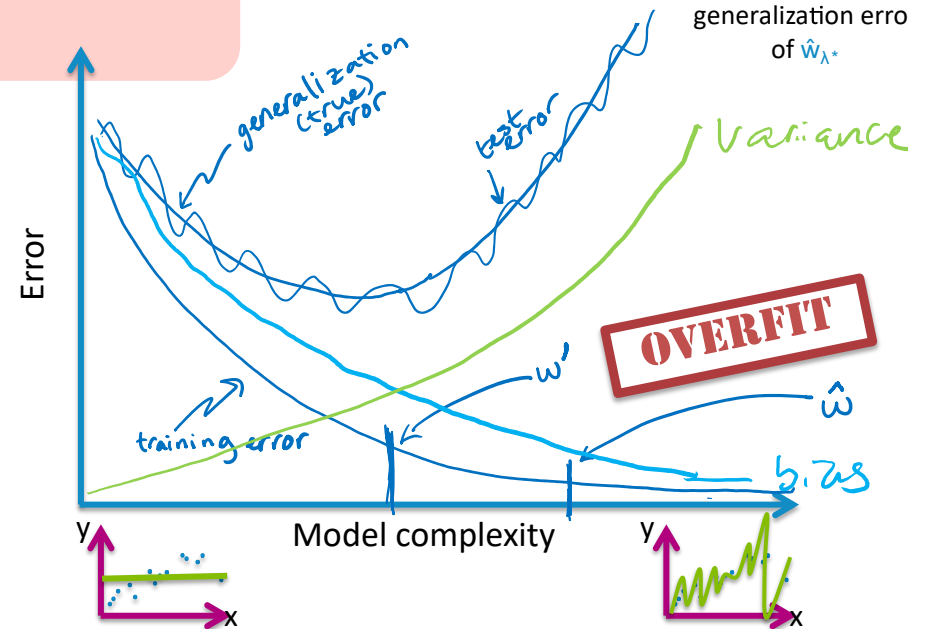
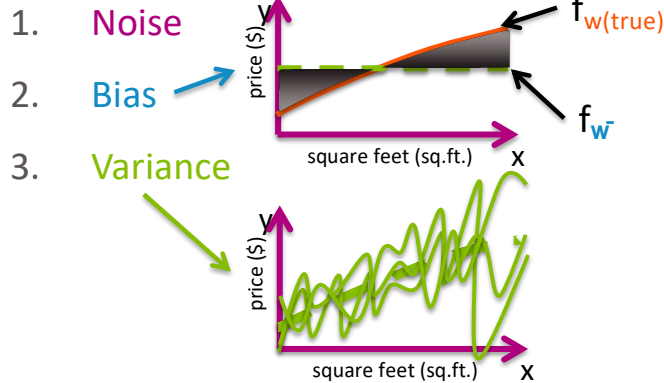
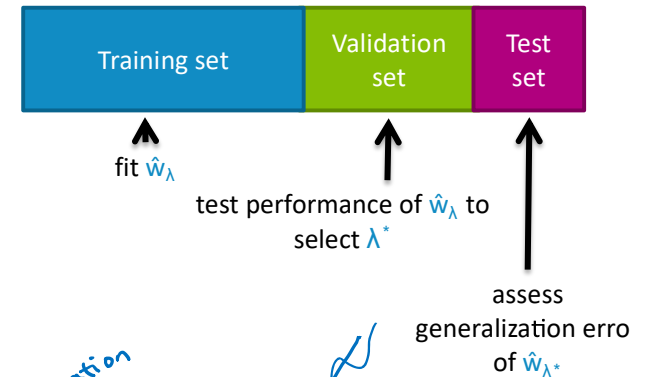


Regression

Case study: Predicting house prices

Concepts

- Loss functions, bias-variance tradeoff, cross-validation, sparsity, overfitting, model selection



Case Study 2: Sentiment analysis



Sushi was awesome,
the food was awesome,
but the service was awful.

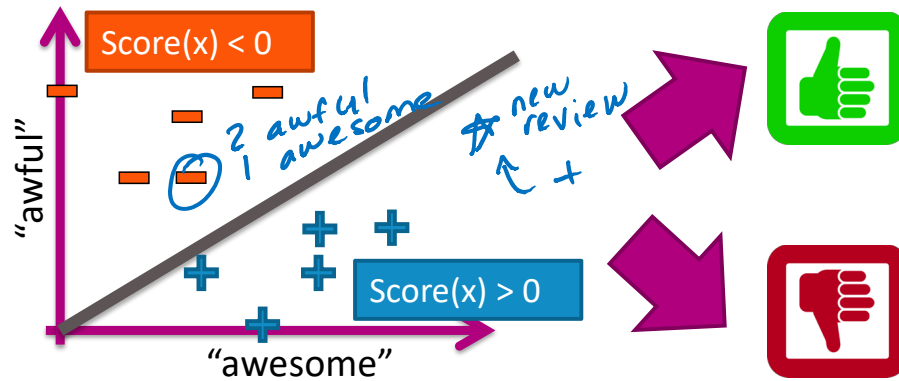
stars / +/-
text

All reviews:

★★★★★ 7/21/2015
This is probably my favorite place to eat Japanese in Seattle. My boyfriend and I ordered nigiri of scallop, Japanese snapper (seasonal), and the agedashi tofu and 2 special rolls. I would skip the special rolls, because the nigiri and sashimi cuts is where this place excels. The tofu, as recommended by other Yelpers was amazing. It's more chewy and the sauce/gravy is the perfect amount of flavor for the delicate tofu.

★★★★★ 6/11/2015
Dining here at the sushi bar made me feel like sitting front row to an amazing performance. We didn't have resos, banged down to the ID after work, got here breathlessly at 5:10pm, and got the last two seats in the place.

★★★★☆ 6/9/2015
I came here having high expectations due to the reviews of this place, but I was bit disappointed. The restaurant is small so do make reservations when you come here. Dishes cost from \$4-26 each and dishes are small.

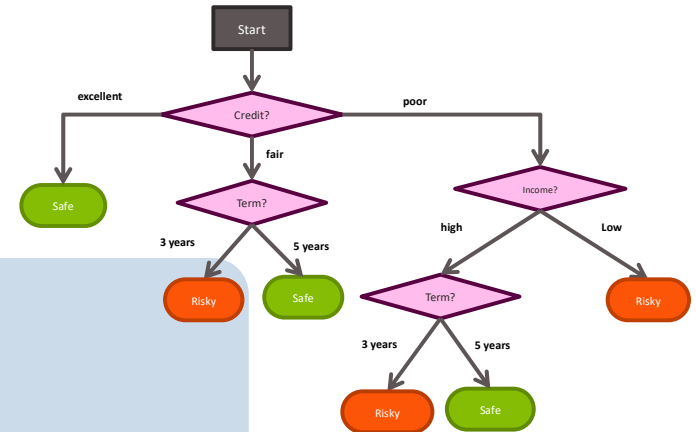


Classification

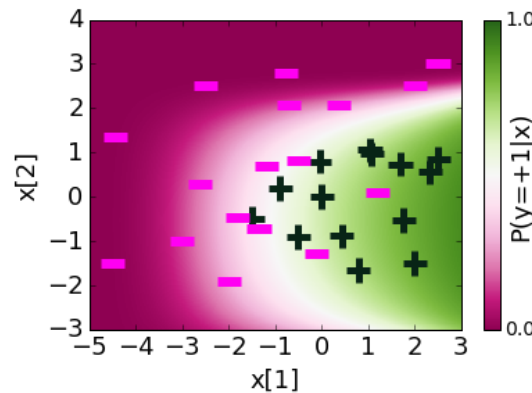
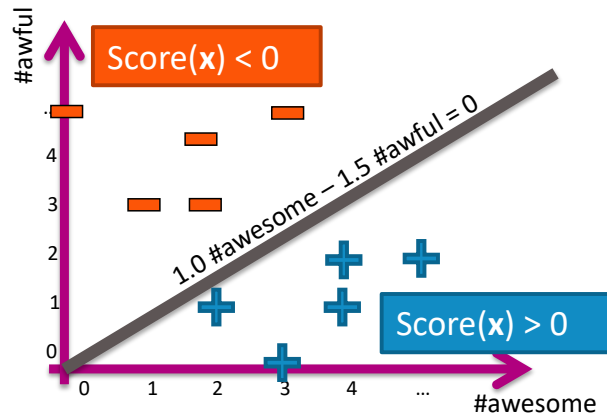
Case study: Analyzing sentiment

Models

- Linear classifiers (logistic regression)
- Multiclass classifiers
- Decision trees
- Boosted decision trees and random forests



Ensemble Method



AdaBoost

$$\hat{f}(x) = \text{sign} \left(\sum_{t=1}^T \hat{w}_t \hat{f}_t(x) \right)$$

↑
model weight

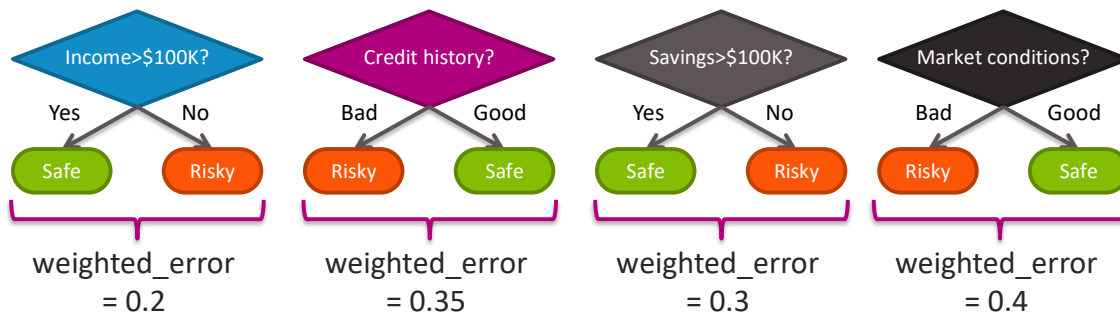
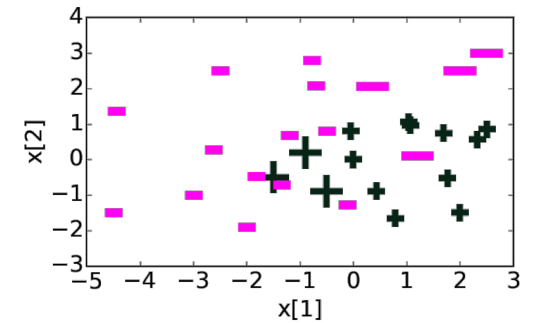
$$\alpha_i \leftarrow \text{dataset weight}$$

Classification

Case study: Analyzing sentiment

Algorithms

- Boosting
- Learning from weighted data



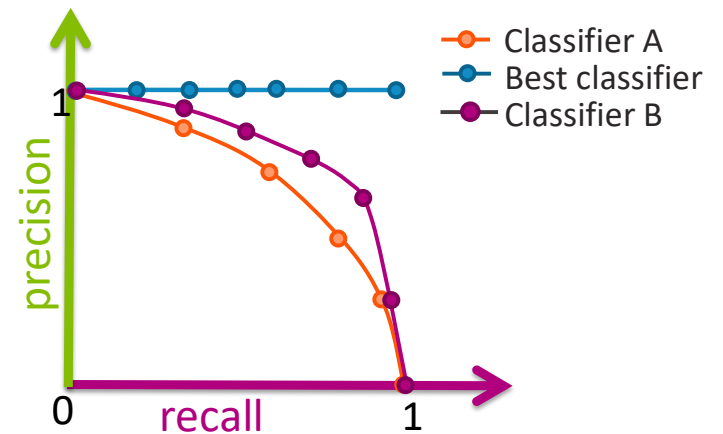
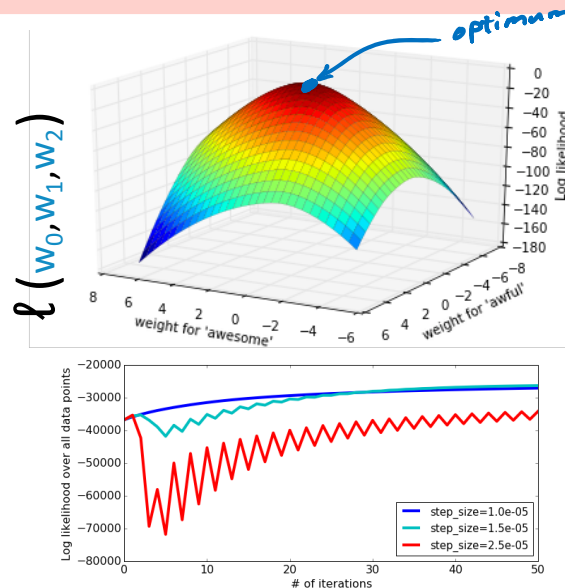
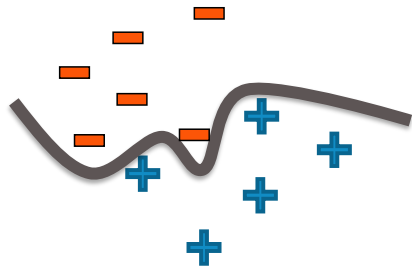
Classification

Case study: Analyzing sentiment

Accuracy vs class imbalance

Concepts

- Decision boundaries, maximum likelihood estimation, ensemble methods, random forests
- Precision and recall



Case Study 3: Document retrieval

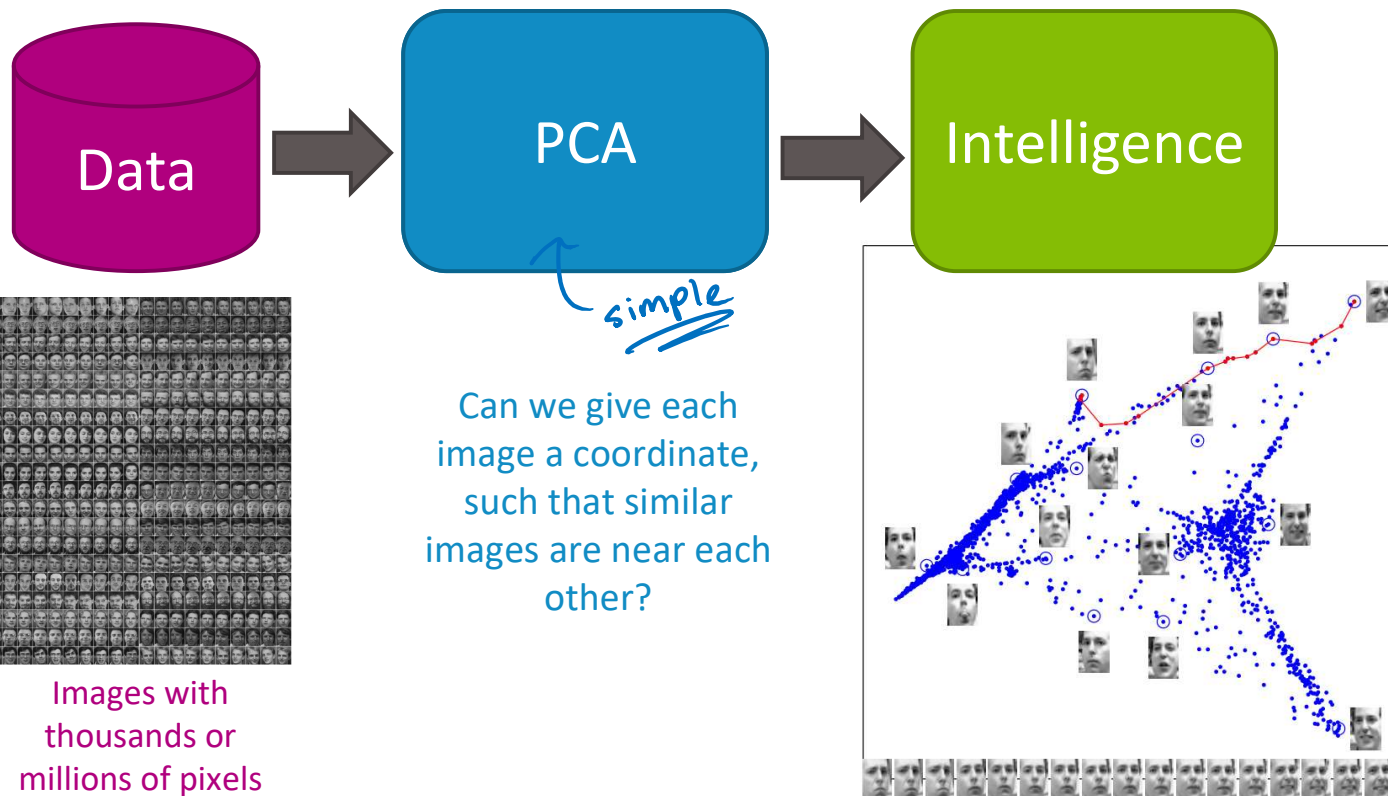


Case Study 3+: Document structuring for retrieval

Bag of Words
TF-IDF



Case Study 3++: Dimensionality reduction



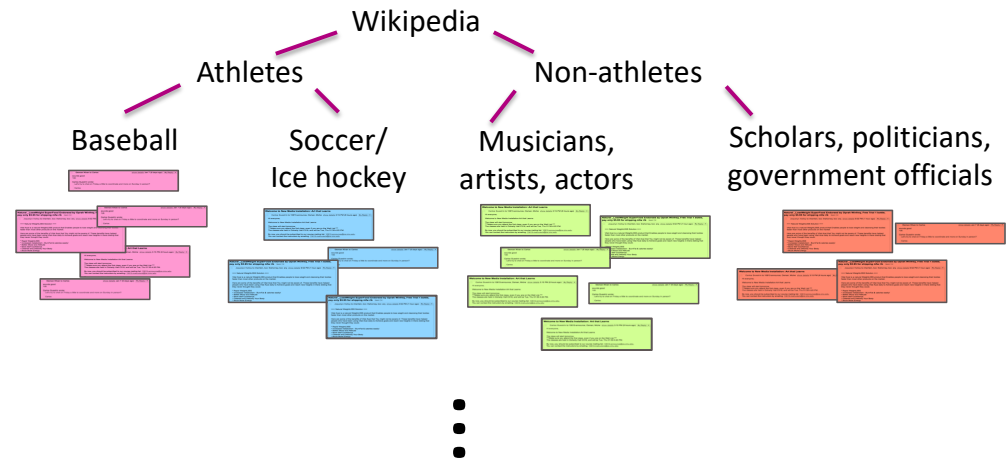
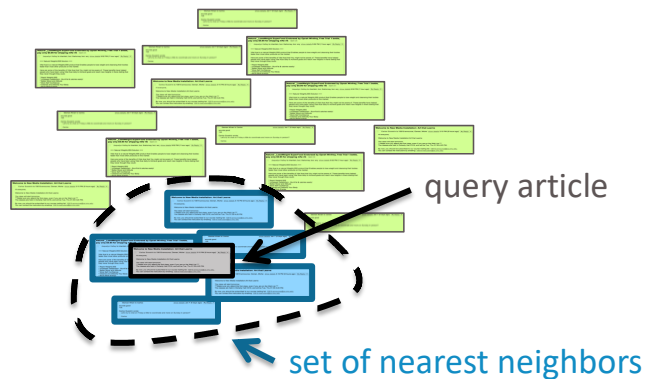
Clustering & Retrieval

Case study: Finding documents

Models

- Nearest neighbors / Kernel methods
- Clustering
- Hierarchical clustering

Agglomerative

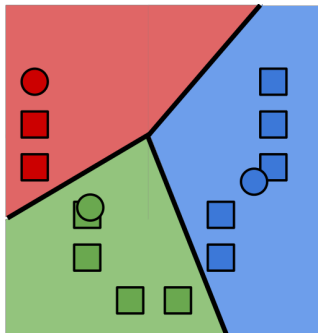


Clustering & Retrieval

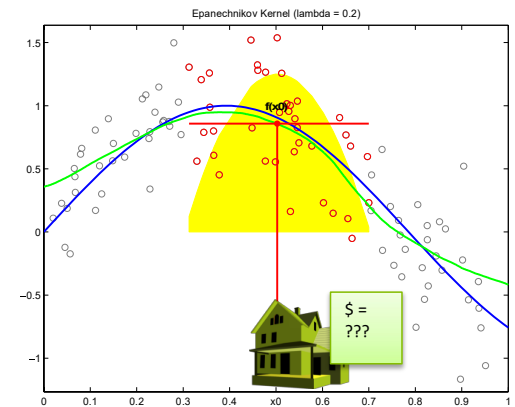
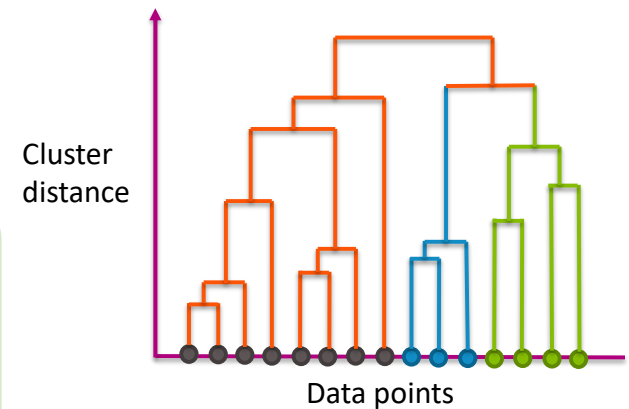
Case study: Finding documents

Algorithms

- k-means , *k-means++*
- Locality-sensitive hashing (LSH)
- NN regression and classification
- Kernel regression
- Agglomerative and divisive clustering
- PCA



*Lloyd's
Algorithm*

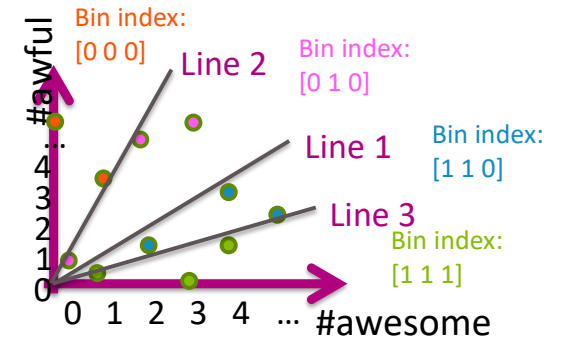


Clustering & Retrieval

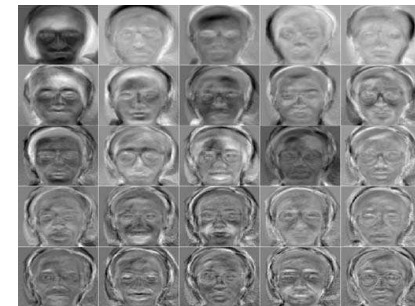
Case study: Finding documents

Concepts

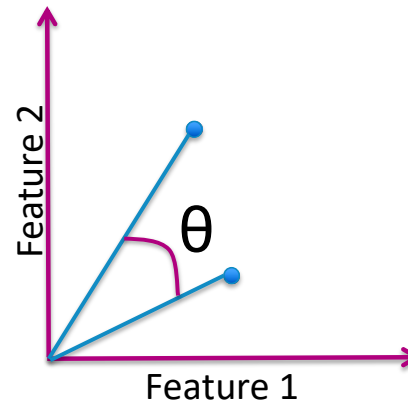
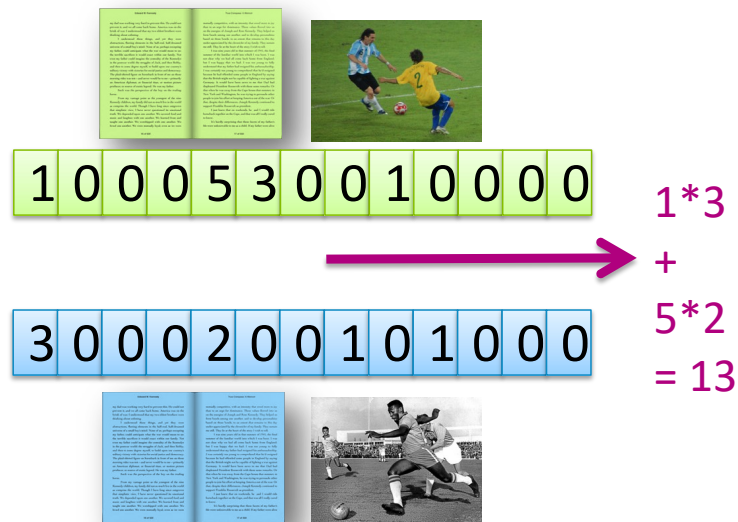
- Distance metrics, kernels, approximation algorithms, dimensionality reduction



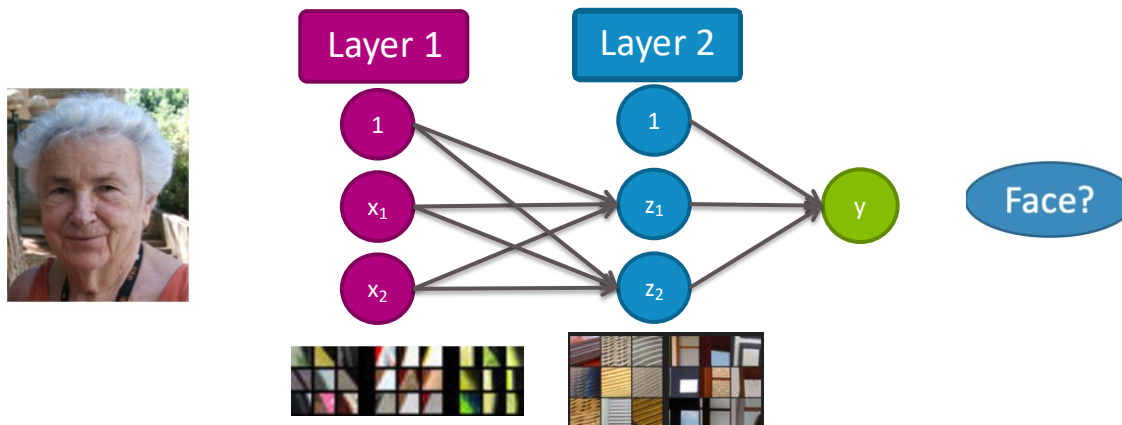
Principal components:



Reconstructing:



Case Study 4: Image classification



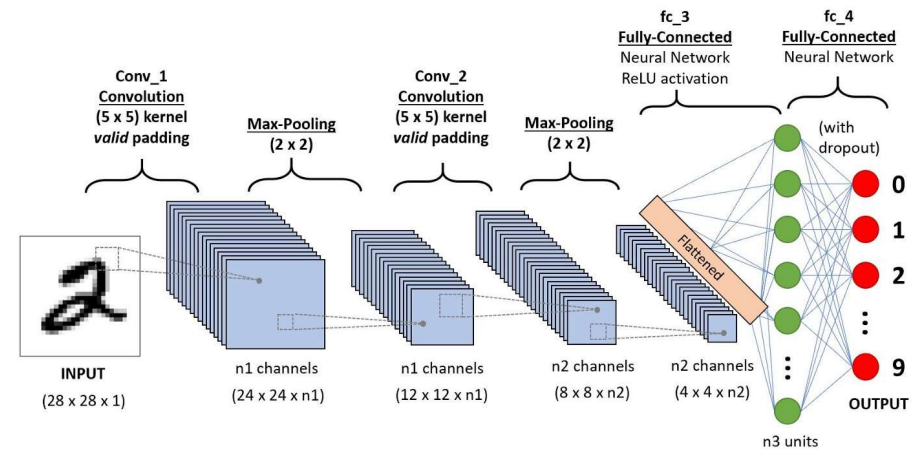
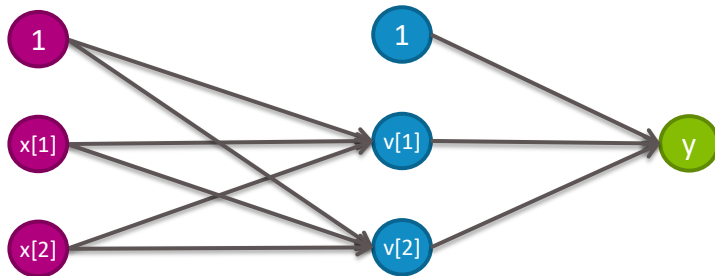
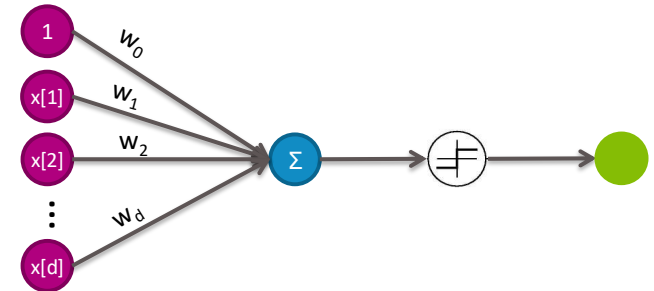
Deep Learning

Case study: Image classification

Models

- Perceptron
- General neural network
- Convolutional neural network

Fully Connected

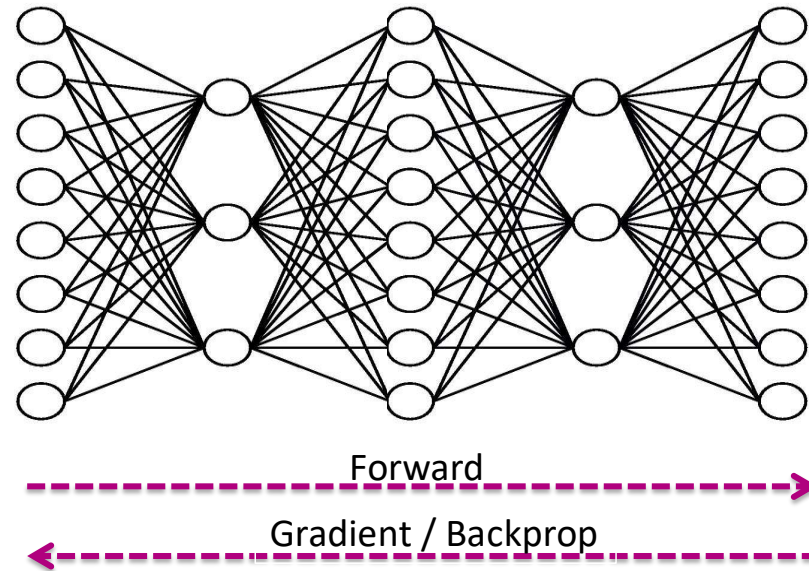
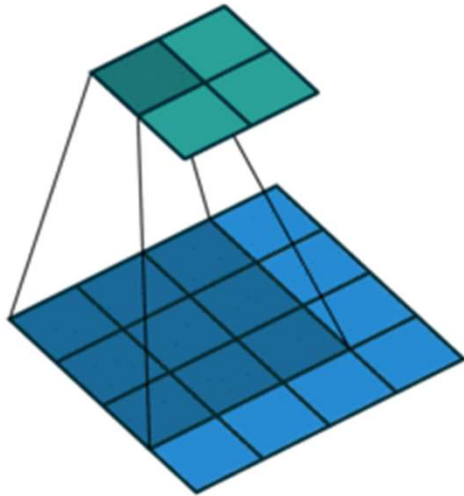


Deep Learning

Case study: Image classification

Algorithms

- Convolutions
- Backpropagation (high level only)

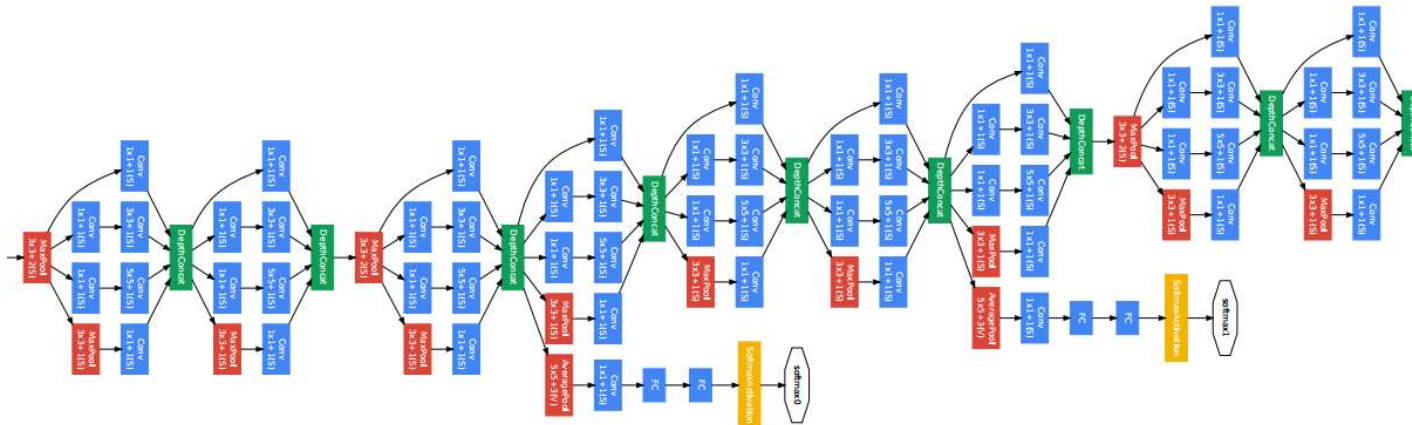
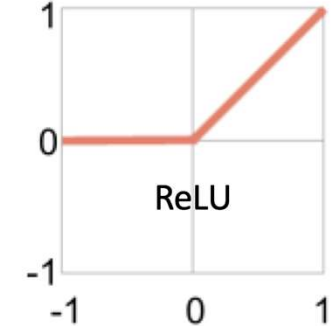
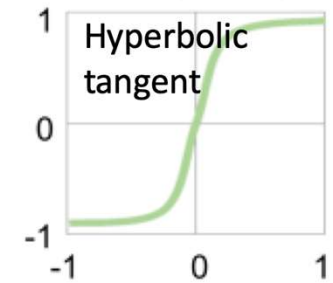
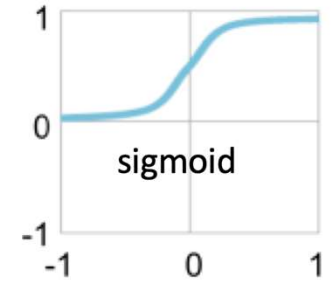
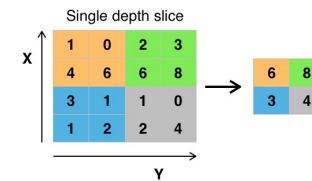
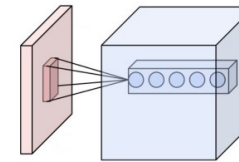


Deep Learning

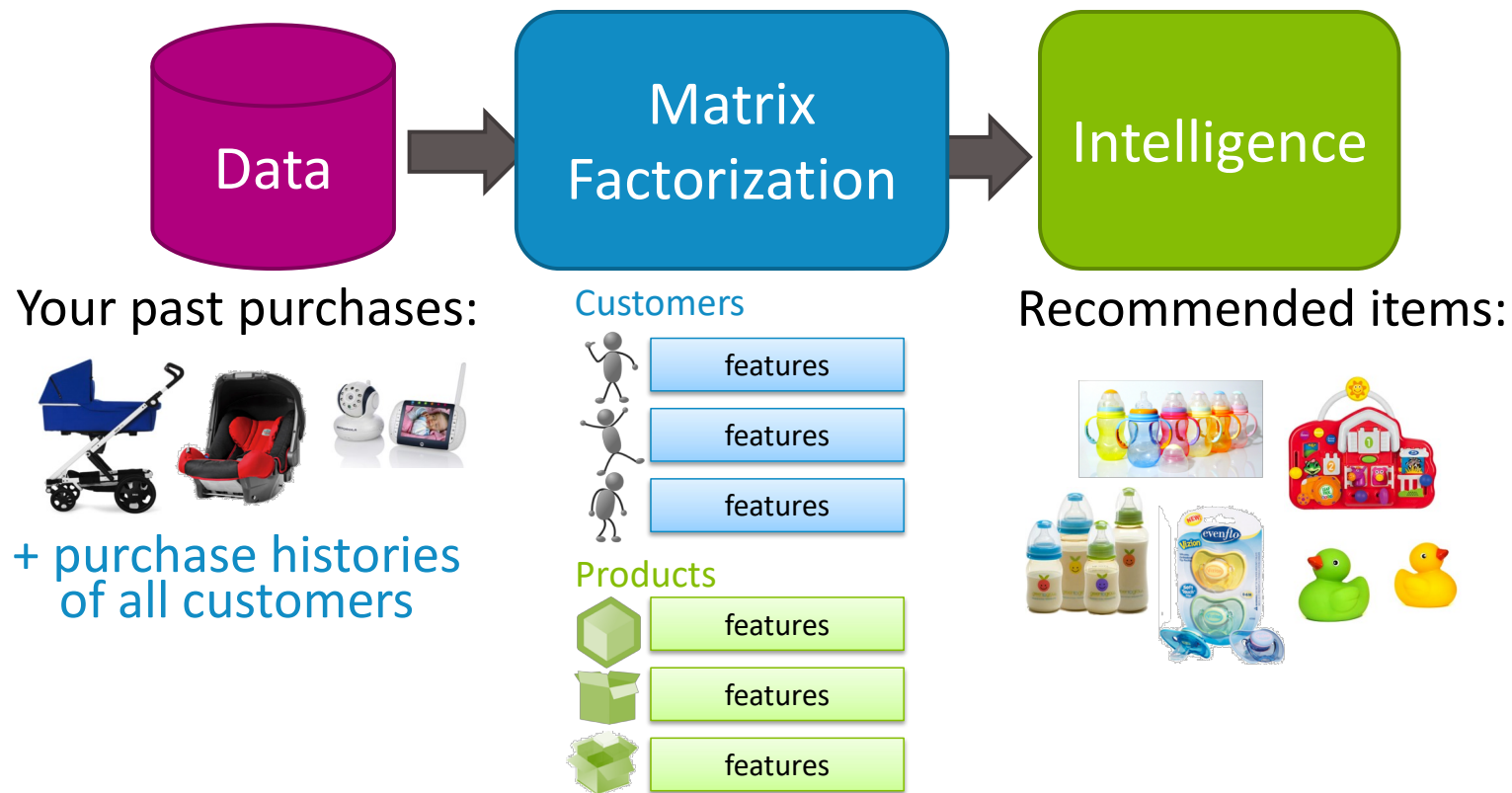
Case study: Image classification

Concepts

- Activation functions, hidden layers, architecture choices



Case Study 5: Product recommendation



Recommender Systems & Matrix Factorization

Case study: Recommending Products

Models

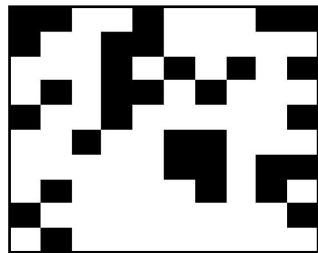
- Collaborative filtering
- Matrix factorization

Popularity

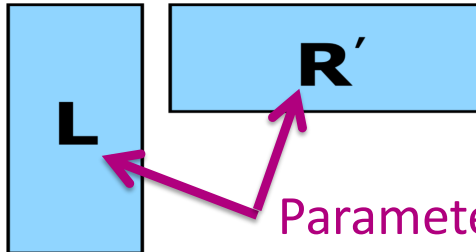
Co-occurrence matrix

Featured MF

Rating =



≈



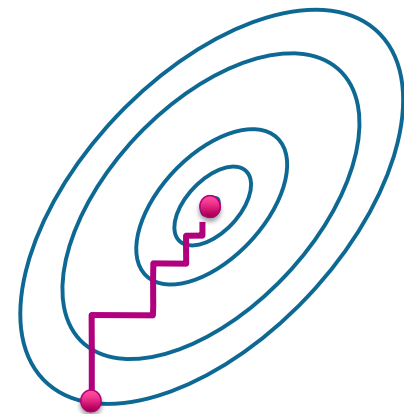
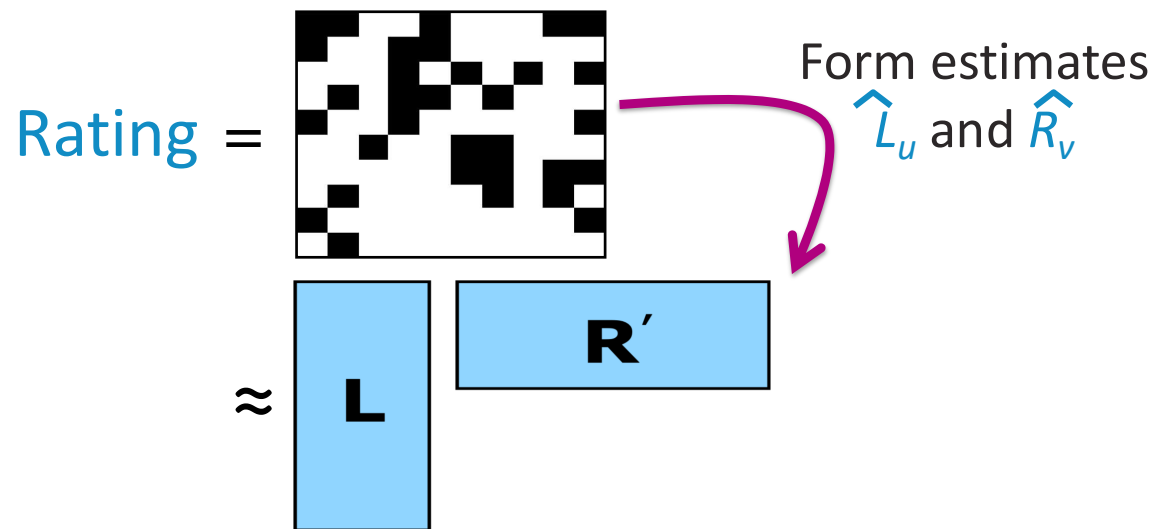
Parameters of model

Recommender Systems & Matrix Factorization

Case study: *Recommending Products*

Algorithms

- Coordinate descent

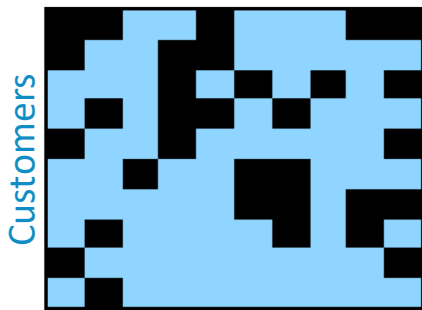


Recommender Systems & Matrix Factorization

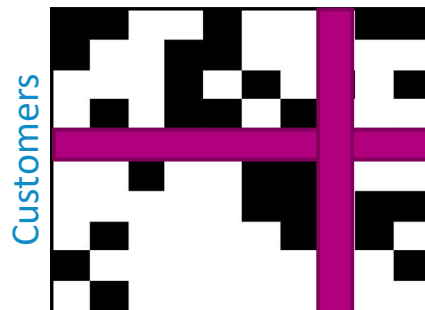
Case study: Recommending Products

Concepts

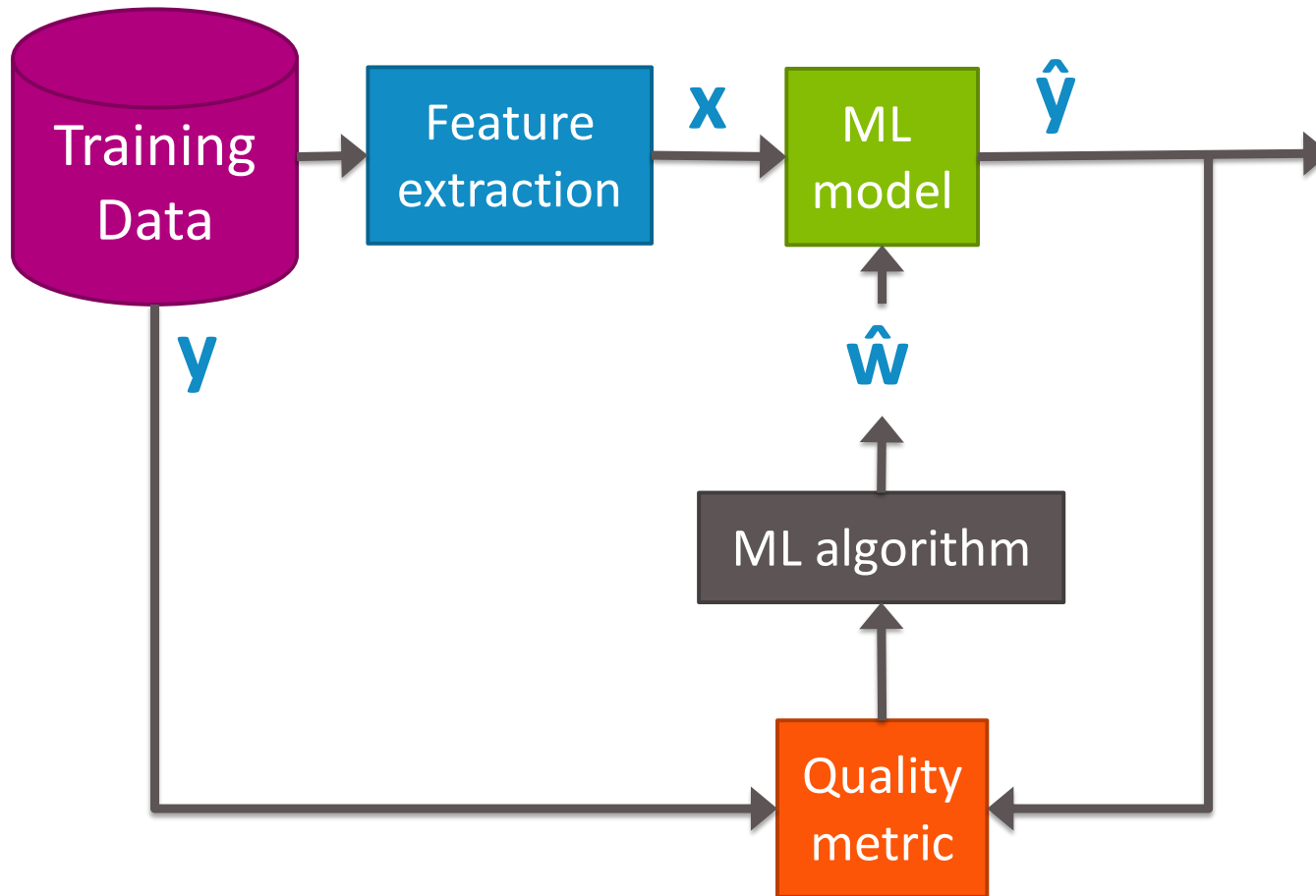
- Matrix completion, cold-start problem



Products



Products



Big Picture

Improving the performance at some task through experience!

- Before you start any learning task, remember fundamental questions that will impact how you go about solving it

What is the learning problem?

What model?

With what optimization algorithm?

How will you evaluate the model?

From what experience?

What loss function are you optimizing?

Are there any guarantees?

Congrats on finishing CSE/STAT 416!
Thanks for the hard work!

