



Brain Break

16.55



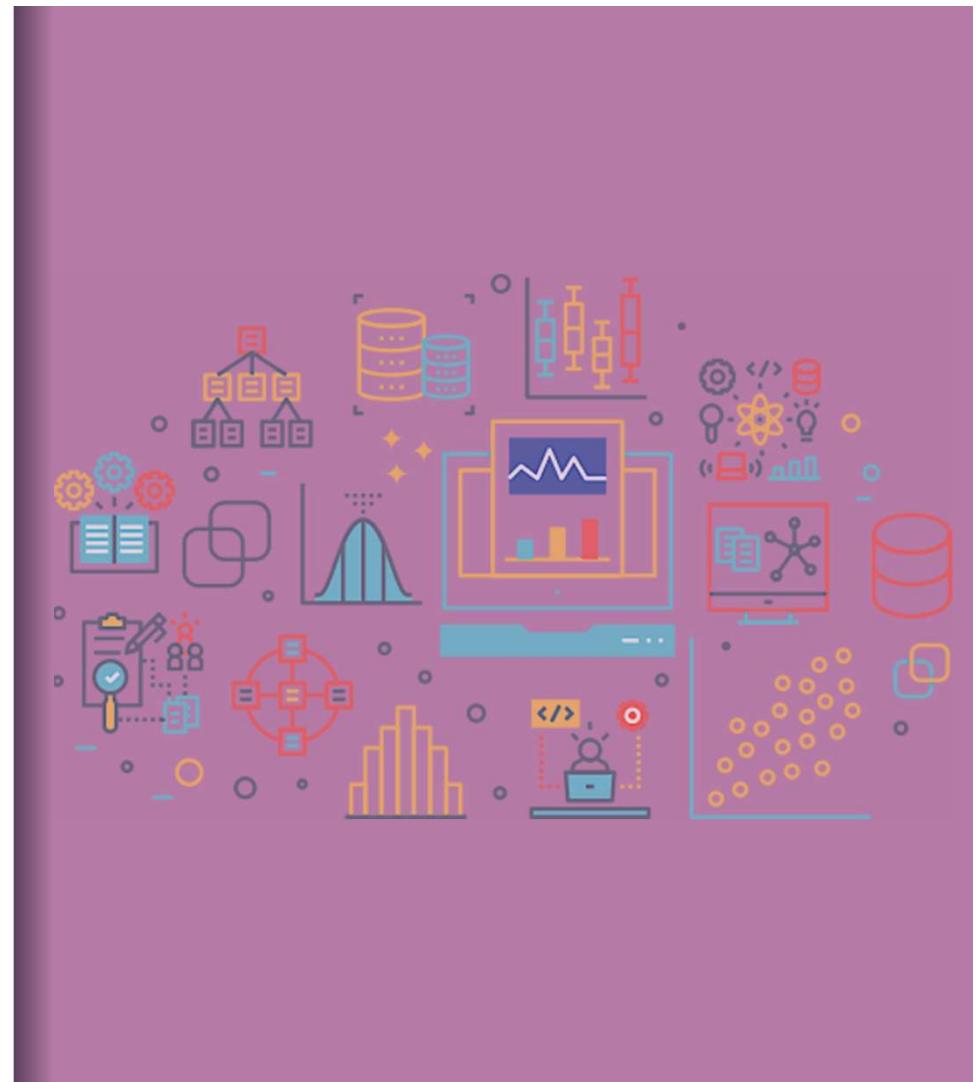
4GIFs
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CSE/STAT 416

Course Wrap Up

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University of Washington
Aug 19, 2019

Slides borrowed from Emily Fox



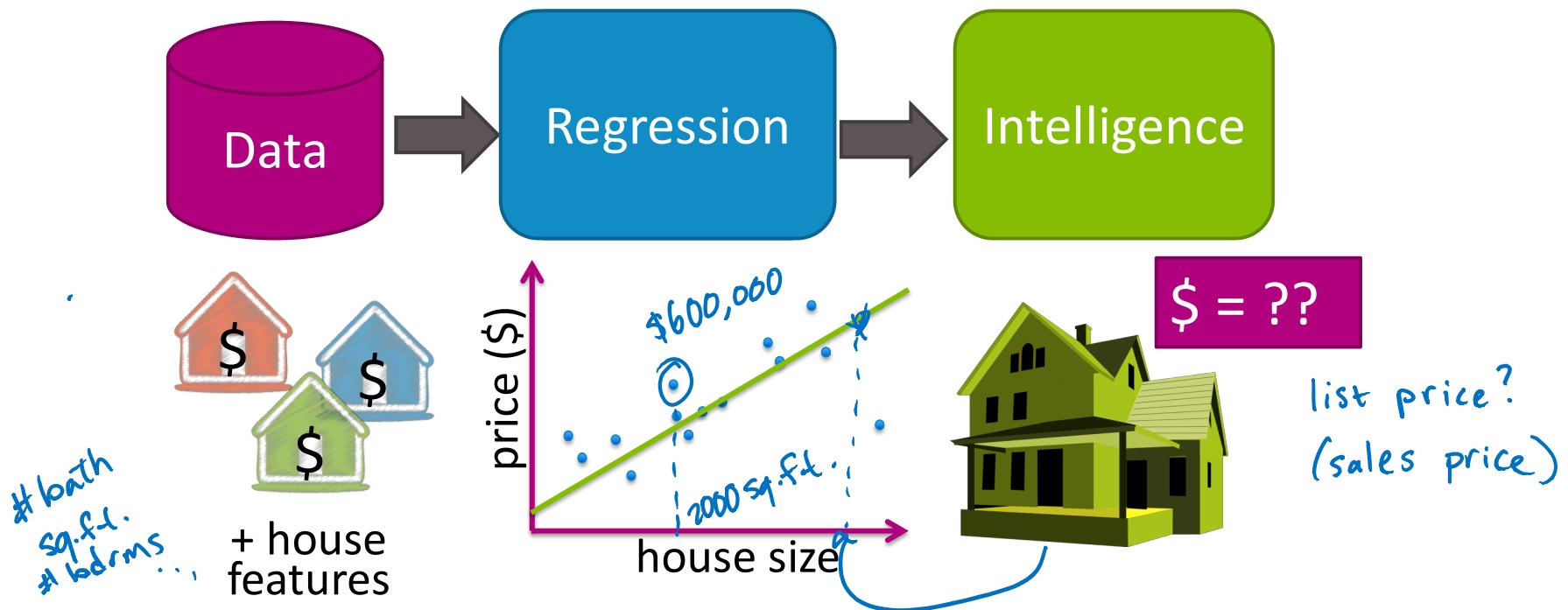
One Slide

- 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

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

Case study: Predicting house prices

Models

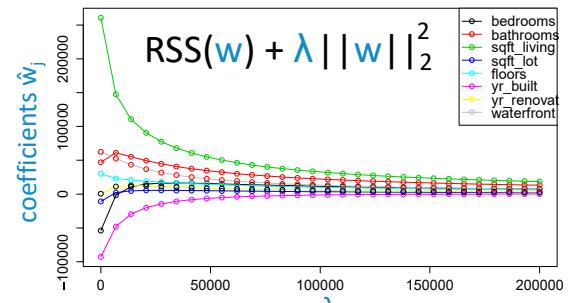
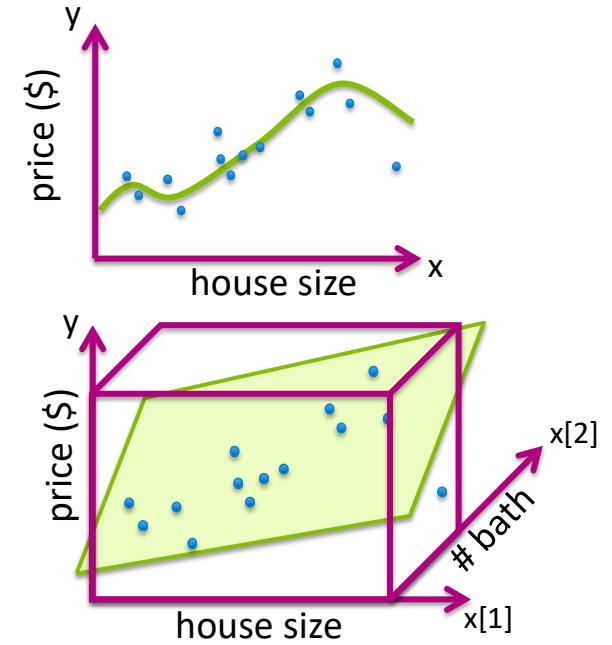
- Linear regression
- Regularization:
Ridge (L2), Lasso (L1)

Including many features:

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



LASSO \Rightarrow Sparse



Regression

Case study: Predicting house prices

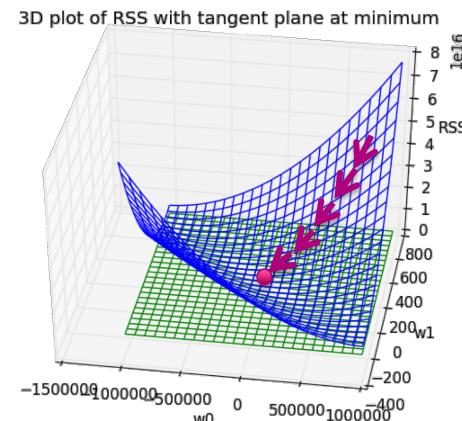
Algorithms

- Gradient descent

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



$$\text{RSS}(\omega) = \sum_{i=1}^n (\omega^\top h(x_i) - y_i)^2$$

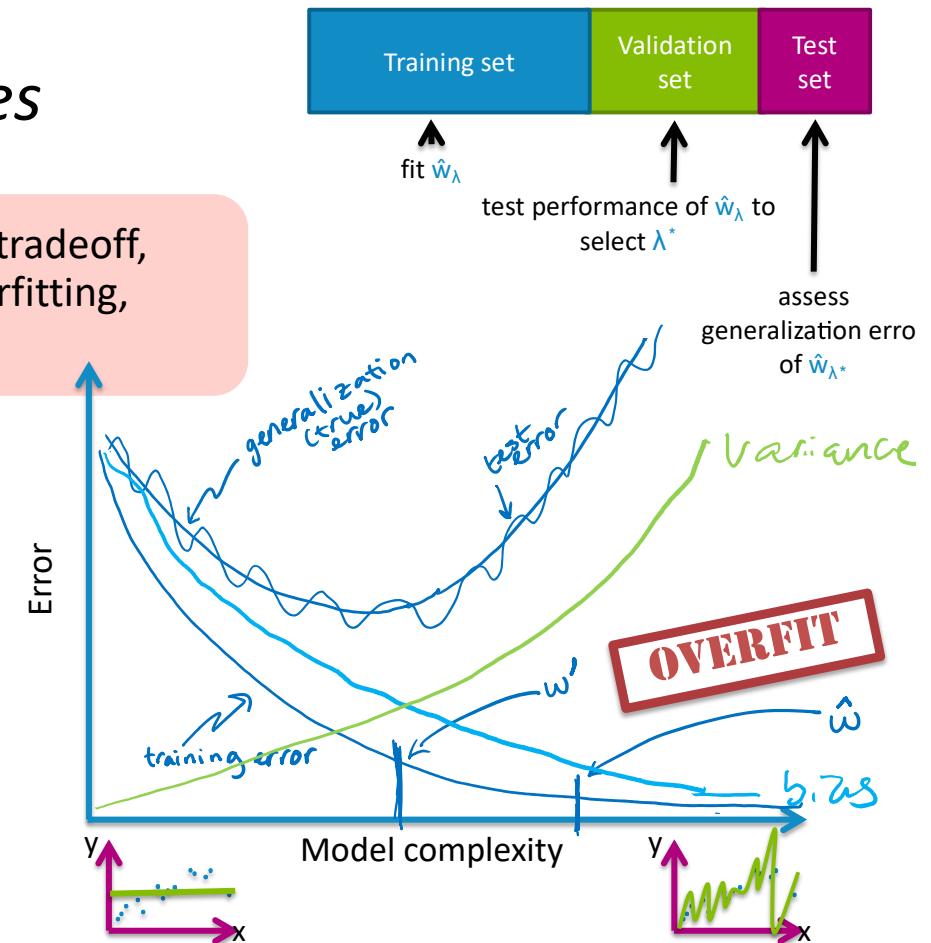
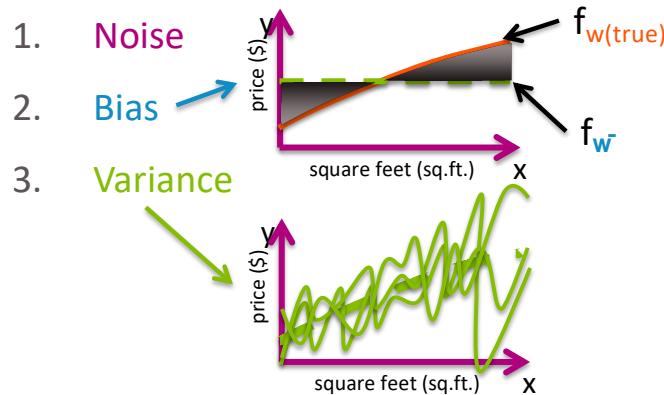


Regression

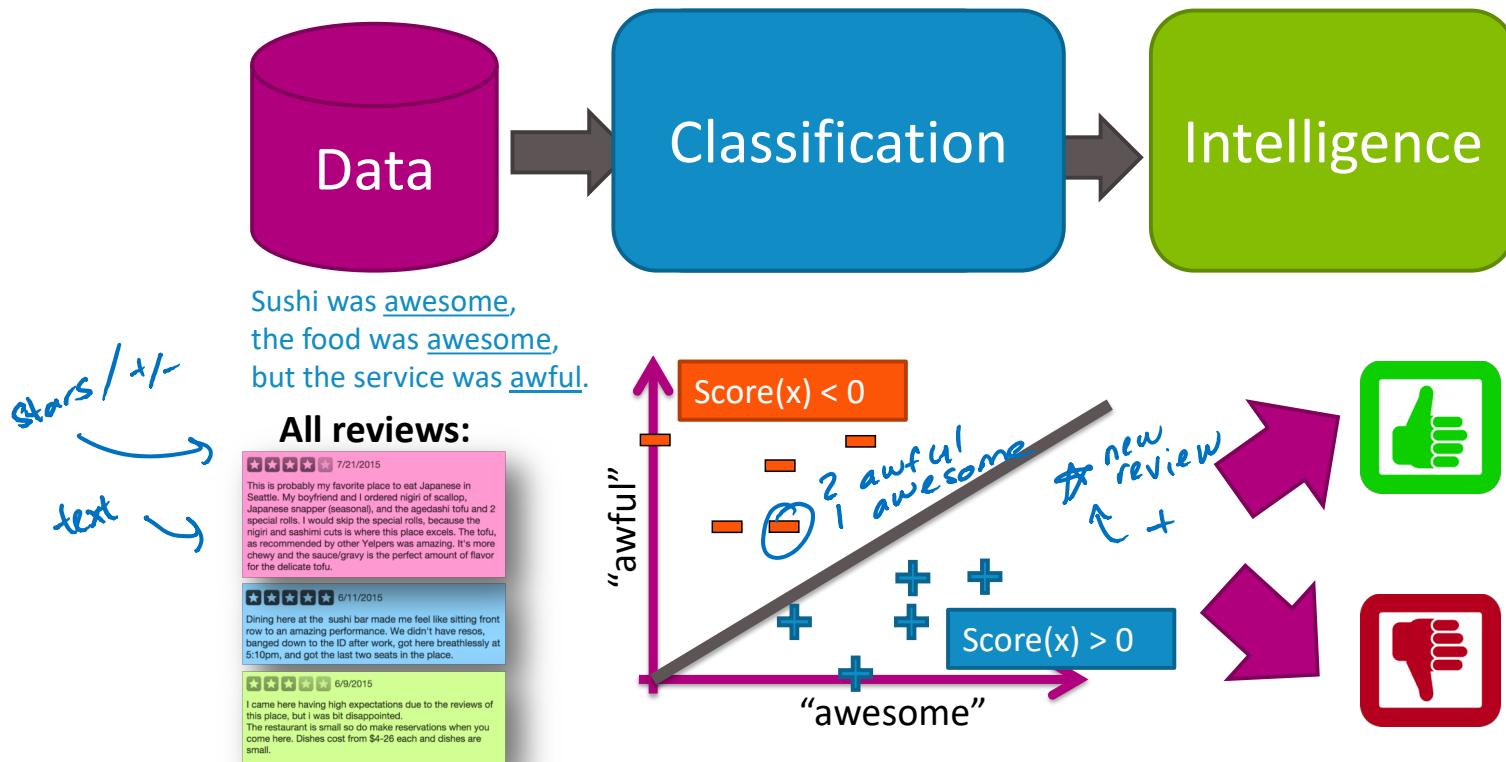
Case study: Predicting house prices

Concepts

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



Case Study 2: Sentiment analysis



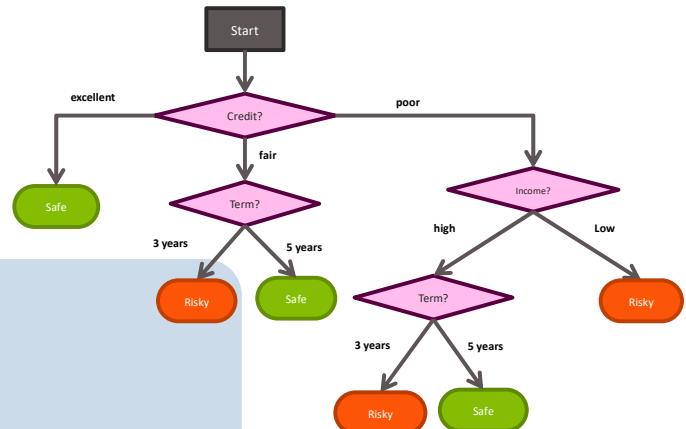
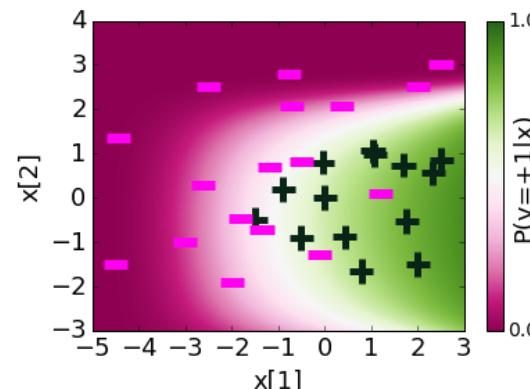
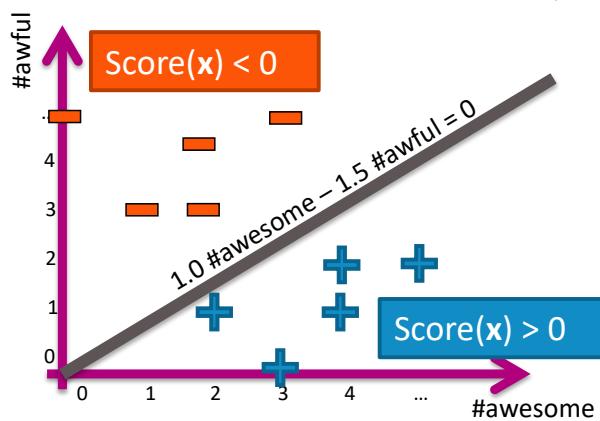
Classification

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

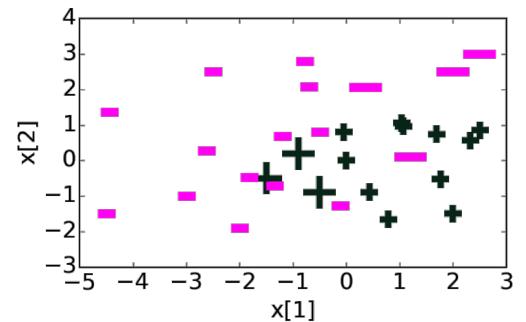
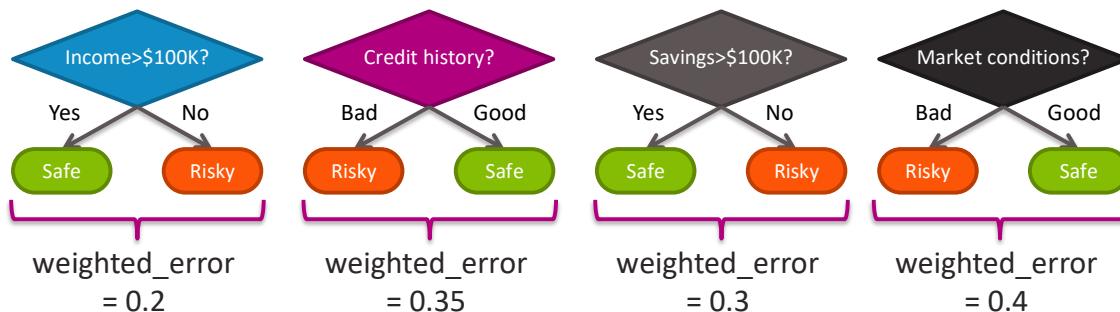
$\alpha_i \leftarrow \text{dataset weight}$

Classification

Case study: Analyzing sentiment

Algorithms

- Boosting
- Learning from weighted data

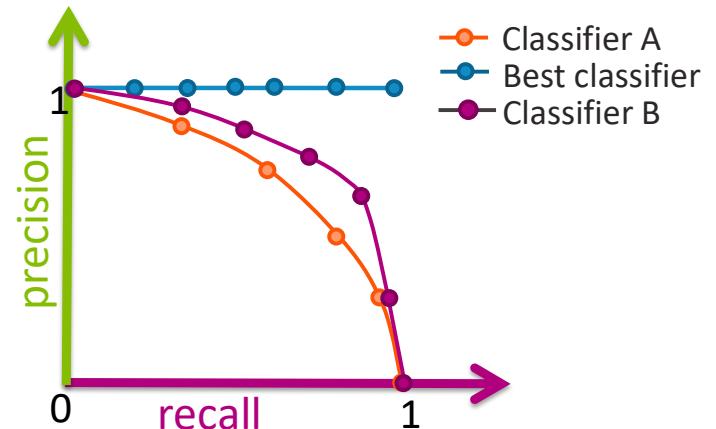
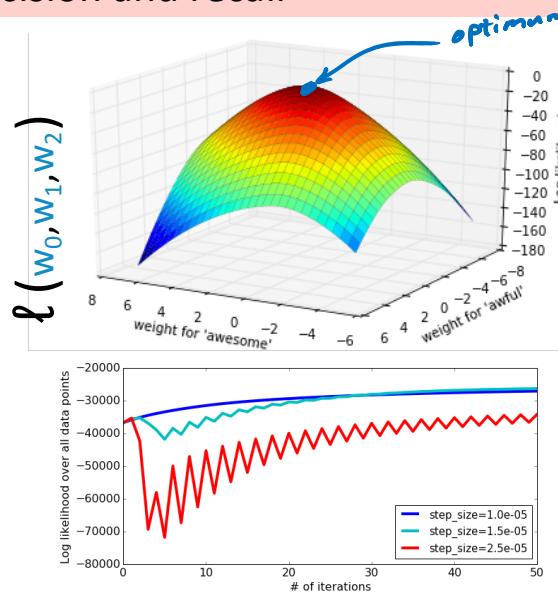
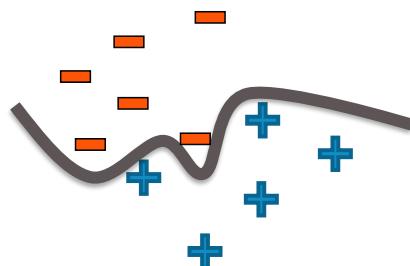


Classification

Case study: Analyzing sentiment

Concepts

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



Accuracy
vs class imbalance

Case Study 3: Document retrieval



Case Study 3+:

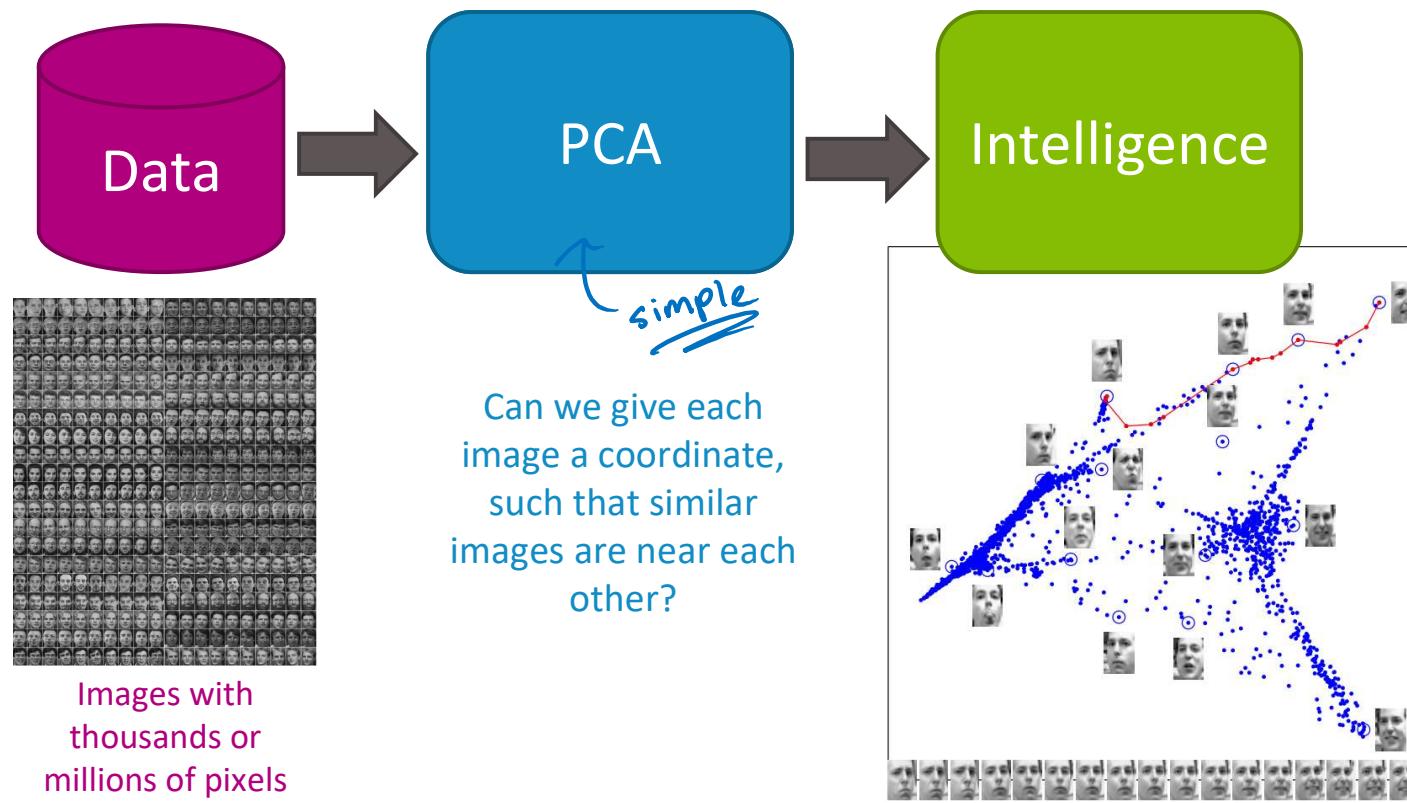
Document structuring for retrieval

Bag of Words
TF-IDF



Euclidean
Manhattan
Cosine

Case Study 3++: Dimensionality reduction



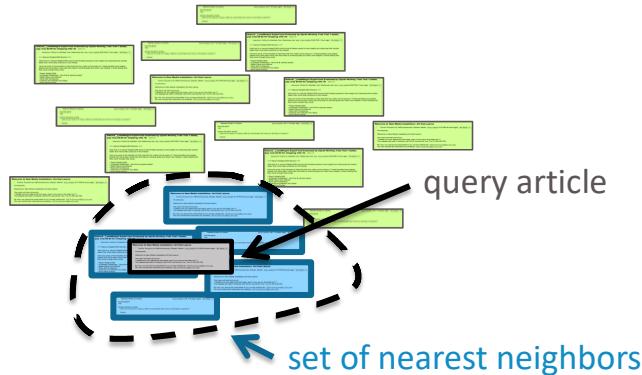
Clustering & Retrieval

Case study: Finding documents

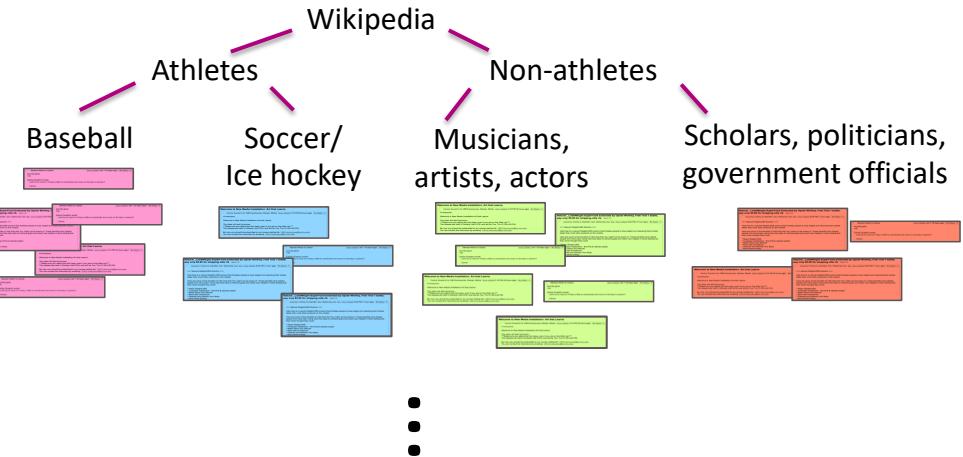
Models

- Nearest neighbors
- Clustering
- Hierarchical clustering

Agglomerative



/ Kernel methods

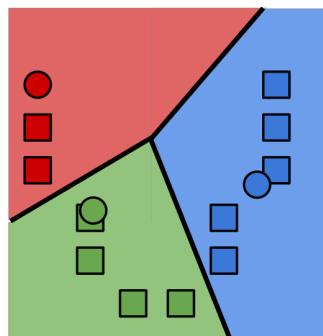


Clustering & Retrieval

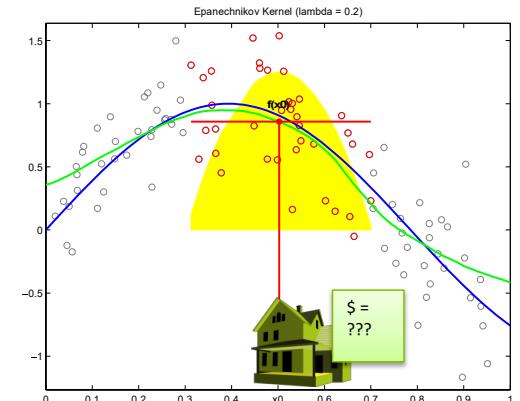
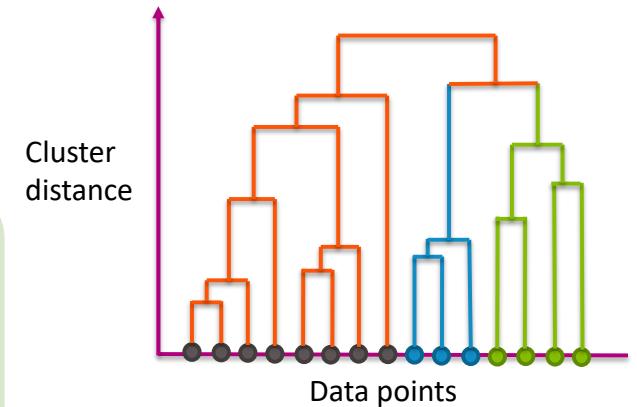
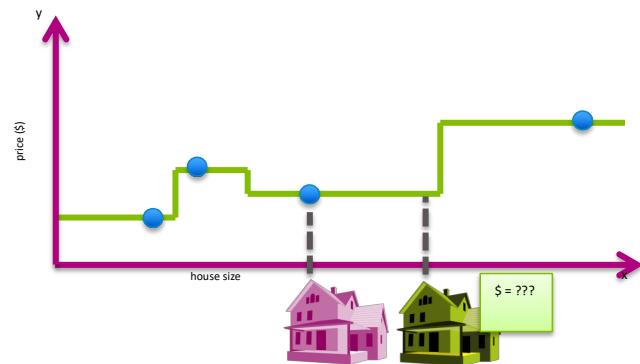
Case study: Finding documents

Algorithms

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



Lloyd's
Algorithm



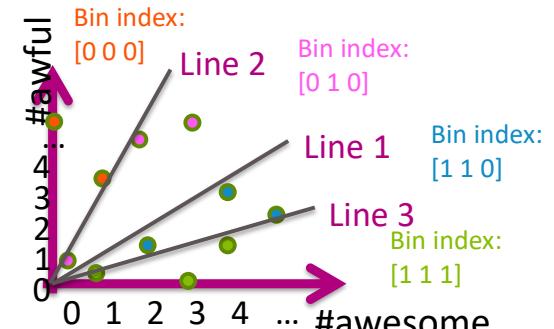
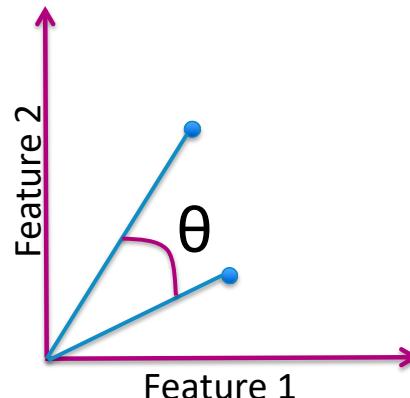
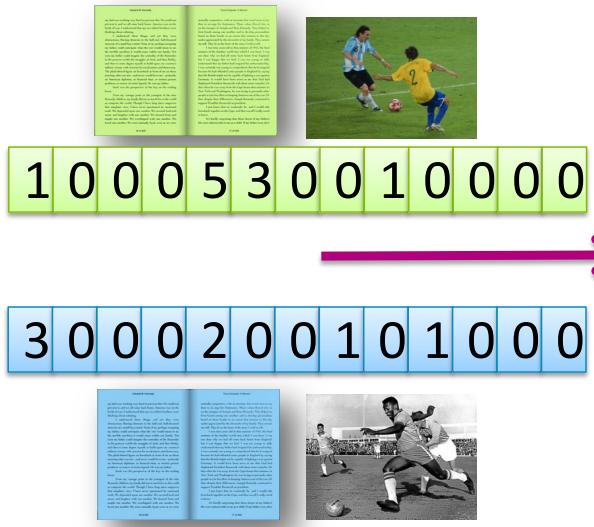
STAT/CSE 416: Intro to Machine Learning

Clustering & Retrieval

Case study: Finding documents

Concepts

- Distance metrics, kernels, approximation algorithms, dimensionality reduction



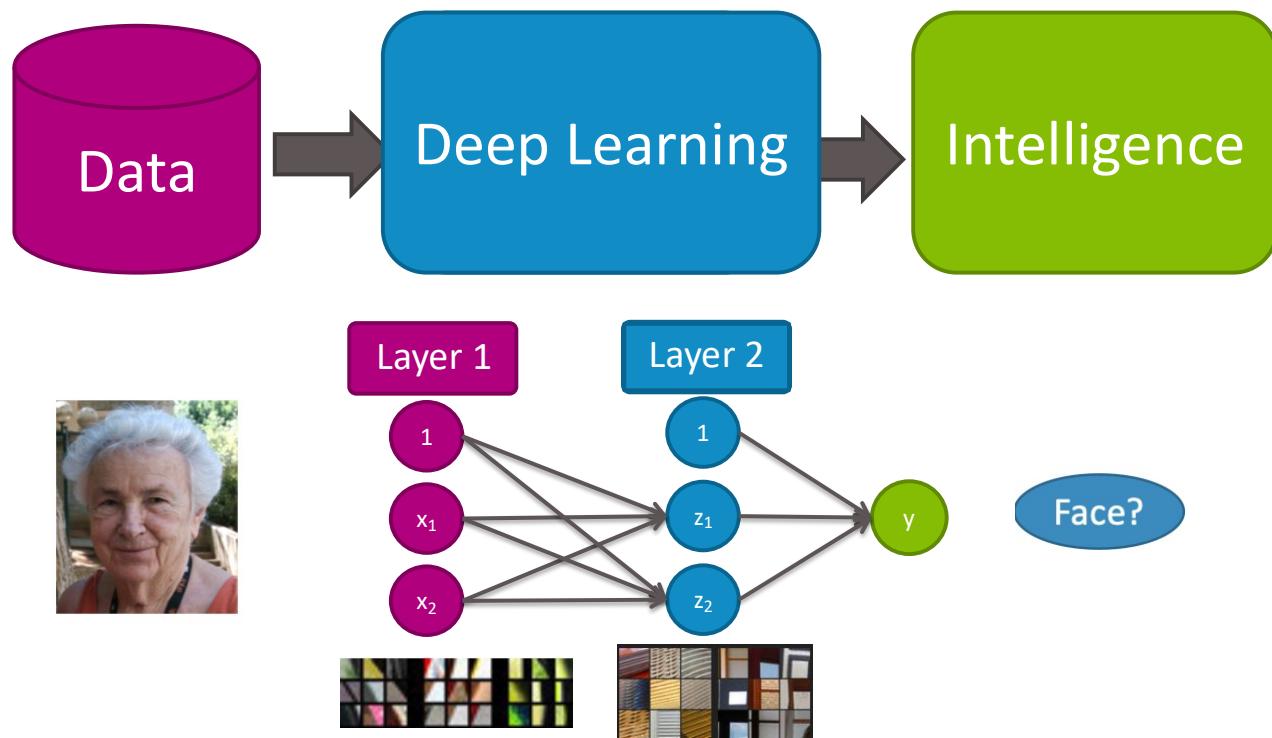
Principal components:



Reconstructing:



Case Study 4: Image classification



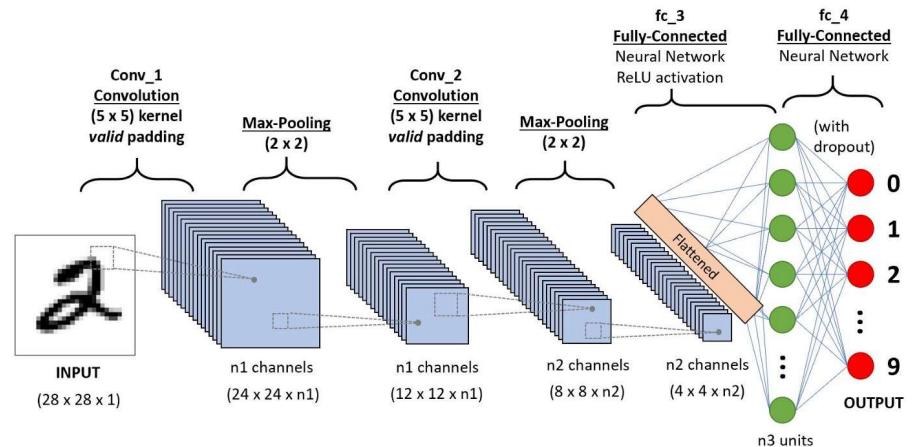
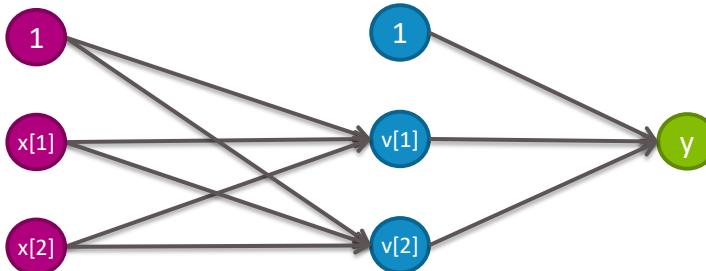
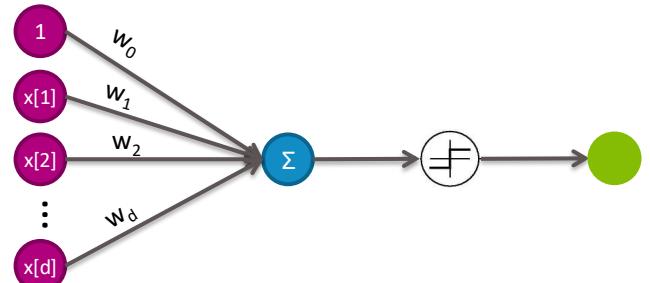
Deep Learning

Case study: Image classification

Models

- Perceptron
- General neural network
- Convolutional neural network

Fully Connected

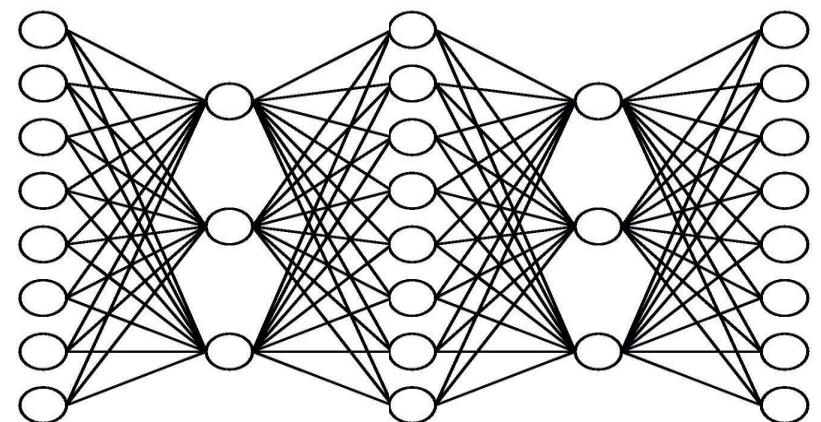
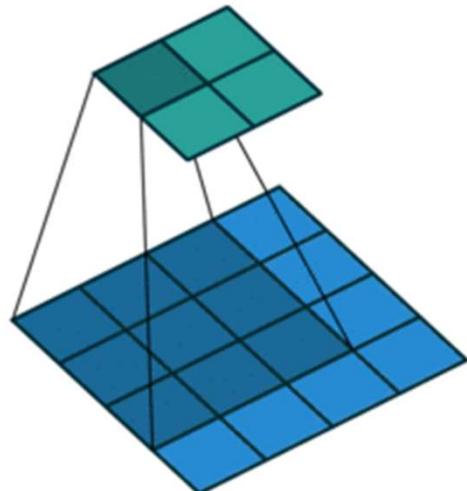


Deep Learning

Case study: Image classification

Algorithms

- Convolutions
- Backpropagation (high level only)

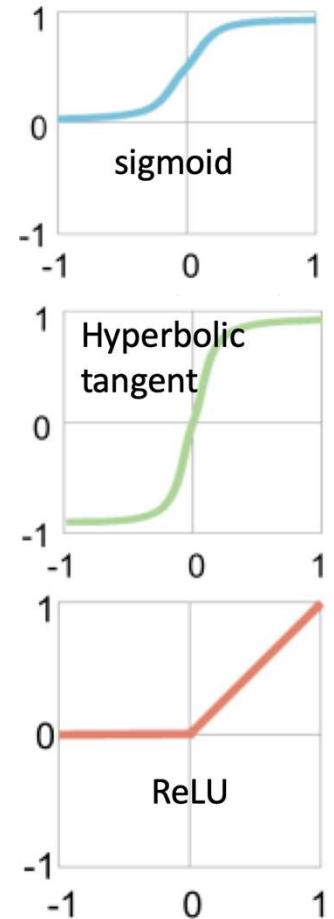
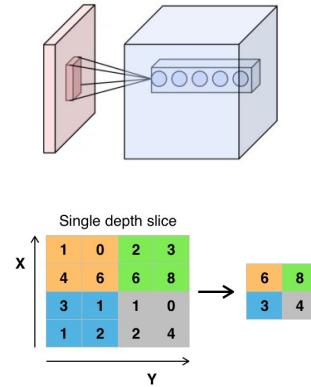
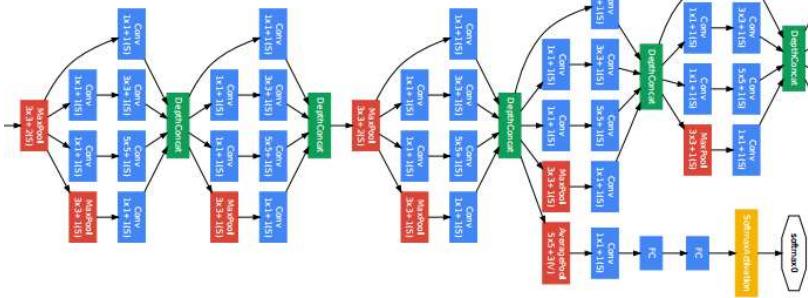


Deep Learning

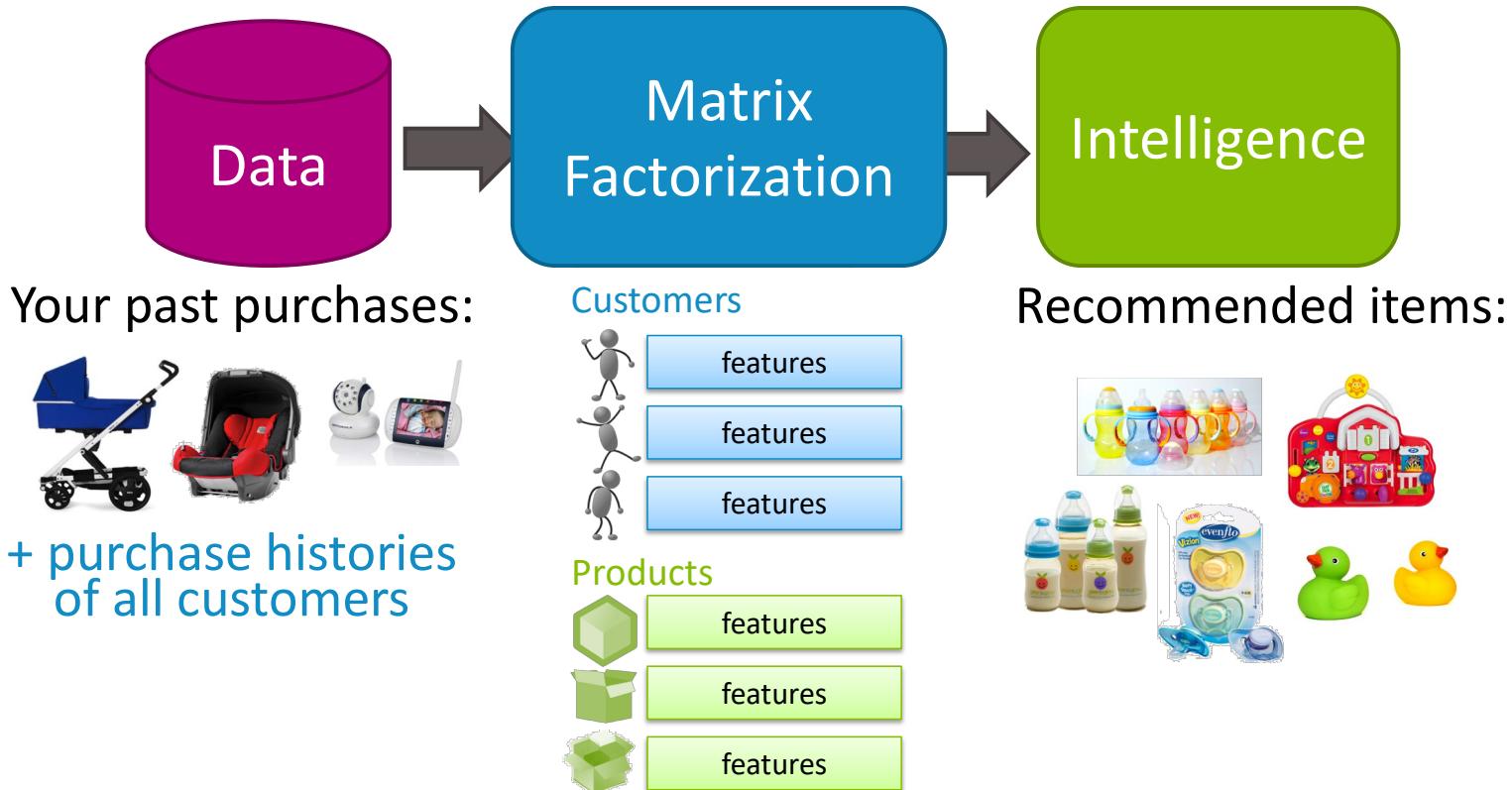
Case study: Image classification

Concepts

- Activation functions, hidden layers, architecture choices



Case Study 5: Product recommendation



Recommender Systems & Matrix Factorization

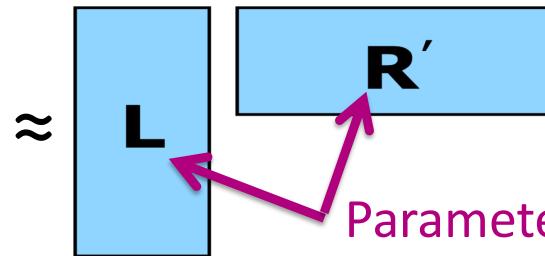
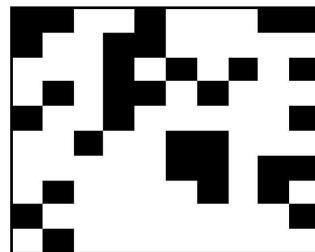
Case study: Recommending Products

Models

- Collaborative filtering
- Matrix factorization

Popularity
Co-occurrence matrix
Featureized MF

Rating =



Parameters of model

Recommender Systems & Matrix Factorization

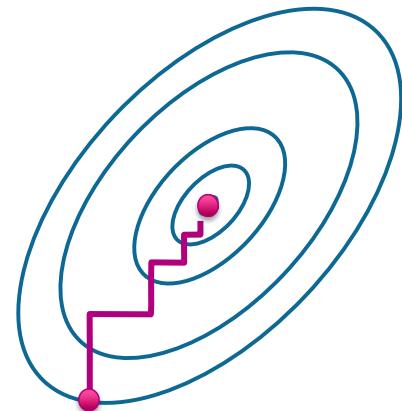
Case study: Recommending Products

Algorithms

- Coordinate descent

$$\text{Rating} = \begin{matrix} \text{Matrix} \\ \approx \begin{matrix} L & R' \end{matrix} \end{matrix}$$

Form estimates \hat{L}_u and \hat{R}_v

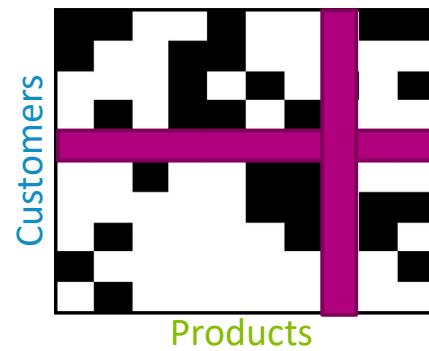
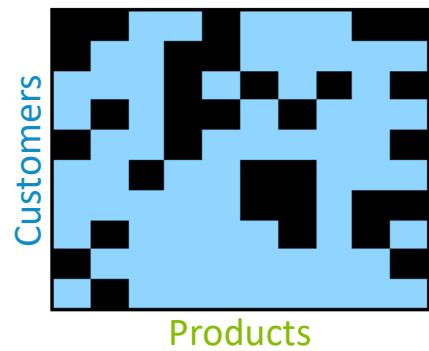


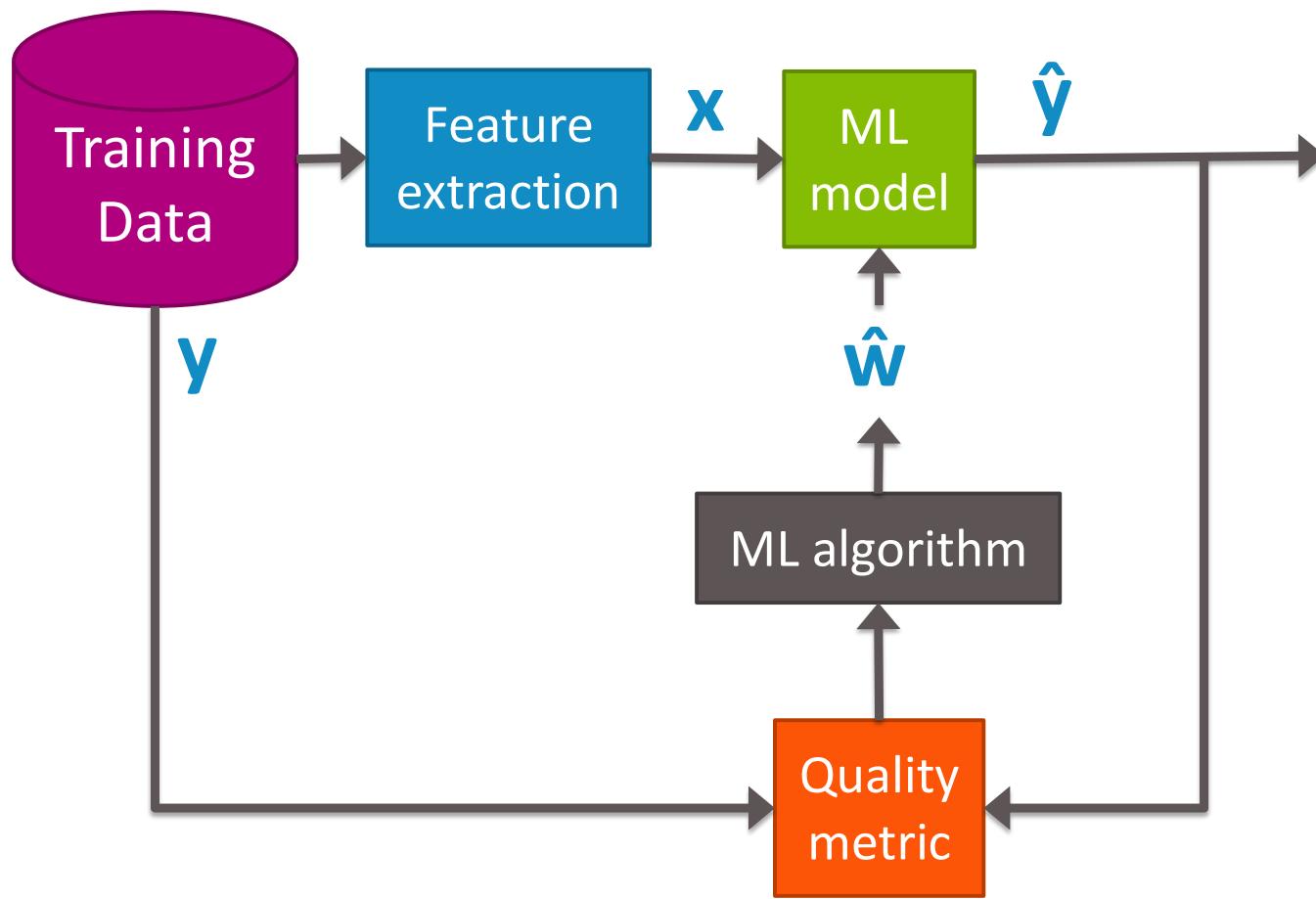
Recommender Systems & Matrix Factorization

Case study: Recommending Products

Concepts

- Matrix completion, cold-start problem





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?

From what
experience?

What model?

What loss function
are you optimizing?

With what
optimization algorithm?

Are there
any guarantees?

How will you
evaluate the model?

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

