Whirlwind Tour of Machine Learning (Pattern Recognition)

- Well established academic fields
- Participation from statisticians, computer scientists, mathematicians and engineers
- \* Many useful, tried-and-true techniques
- Resurgence
  SVM, boosting
  Deep learning



#### Common Statements \* Objects characterized by traits (features) □ You credit score, income, debt in loan application □ Your age, look, hobby in dating services □ Website usage, traffic, geographical loc in intrusion detection □ SIFT, wavelet features in image analysis □ Finger print patterns (ridges, bifurcation, terminations, etc.) in matching

- □ Location, school district, # of rooms, square footage in housing search
- □ Color, options, engine size, # of doors in auto



## General Approach

- \* Supervised  $f(x) \rightarrow y$
- □ Learn with a teacher model □ Classification (discrete) & regression (continuous) □ Training + validation, testing, and deployment \* Unsupervised f(x) only □ Learning without a teacher model □ Abnormality finding, data mining, etc. \* RL

Sequence (e.g., temporal) data
No or sparse training data and feedback
Maximize reward functions

## Supervised Regression

- Curve, surface fitting
- \* Right order





#### Universal Dilemma

◆ Bias
□ Error
□ How fitting confirms to data

 Variance
 How fitting confirms to each other



#### Universal Dilemma

 Small bias and large variance
 Way too complicated models
 Overfitting

Large bias and small variance
Naïve model
Insufficient training data

Under general assumptions there is a subtle trade off – you cannot optimize both!



### Supervised classification

#### \* Separation of different class samples





Figure 14-1. Scattergram of evtoplasm area versus nuclear area for five different common types of white blood cells. The letters denote the different classes, with the centroids underlined. The dashed lines show linear boundaries that best separate the classes. Several samples are misclassified. (Plotted from data in "Automated Leukocyte Recognition" by I.T. Young, Ph.D. thesis, MIT, Cambridge, Massachusetts, 1969.)



#### Various Types of Errors

		Condition (as determined by "Gold standard")		$Precision = \frac{tp}{tp + fp}$
		Condition positive	Condition negative	$\text{Recall} = \frac{tp}{tp + fn}$
Test outcome	Test outcome positive	True positive	False positive (Type I error)	Precision = Σ True positive Σ Test outcome positive
	Test outcome negative	False negative (Type II error)	True negative	Negative predictive value =Σ True negativeΣ Test outcome negative
		$\frac{\text{Sensitivity}}{\Sigma \text{ True positive}}$ $\overline{\Sigma \text{ Condition positive}}$	Specificity = Σ True negative Σ Condition negative	Accuracy



### Precision vs. Recall

- A very common measure used in PR and MI community
- One goes up and the other HAS to go down
  A range of options (Receiver operating characteristic curves)
  A rea under the curve

as a goodness measure





## Simple Tool

 $X_2$ 









Y



#### More Advanced Math Tools



## "Massage" data



- Nonlinear data transformation
- Combined SVM + Data Massage (Kernel methods) is the most popular and robust methods (up to 5 years ago)

# "Massage" Classifiers

\* "Massage" classifiers
\* Use more than one
\* Use a hierarchy





## General Training Procedures

- Data collection, shuffling, batching
- \* Repeat  $\Box$  Training with known f(x)-> y  $\Box$  Validation with non-overlapping f(x)-> y □ Error drop, keep training results □ Error not drop, discard training results Final accuracy figures (on non-overlapping testing data set)



## Unsupervised Clustering



