Introduction to Machine Learning

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Lecture 17: Support Vector Machines, Clustering, K-Means

> Many figures courtesy Kevin Murphy's textbook, Machine Learning: A Probabilistic Perspective



Losses for Binary Classification



Maximum Margin Hyperplanes



If multiple linear classifiers perfectly separate training data, which should I choose?

Support Vectors & Slack Variables



Support vectors (green) for data separable by radial basis function kernels, and non-linear margin boundaries



Linear decision boundary in feature space, where data violating margin have nonzero "slack variables"

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How Many Support Vectors?



Multiclass Support Vector Machines

Complicated by the fact that binary SVM classifiers are **not** calibrated probabilistic models



On to Unsupervised Learning

	Supervised Learning	Unsupervised Learning
Discrete	classification or categorization	clustering
Continuous	regression	dimensionality reduction

- Goal: Infer label/response y given only features x
- **Classical:** Find latent variables y good for *compression* of x
- Probabilistic learning: Estimate parameters of joint distribution p(x,y) which maximize marginal probability p(x)

Clustering can be Ambiguous





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For compressing new data, more codewords is always better. Cross-validation fails for unsupervised learning!