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The WILLIAM STATES LEE COLLEGE *of* ENGINEERING

Introduction to ML

Lecture 9: Generative-vs-Discriminative

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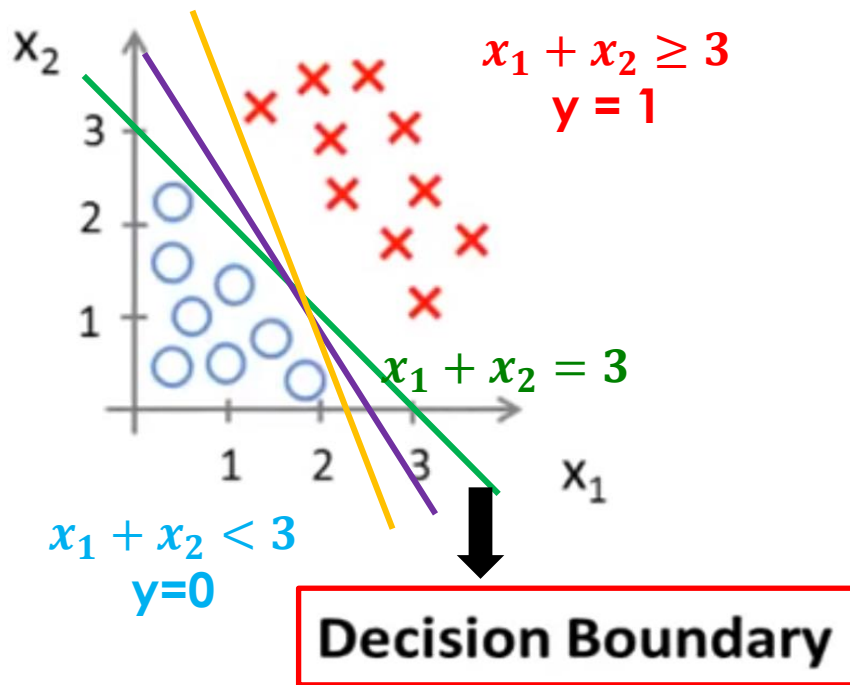
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Generative and discriminative models: An analogy

- The task is to determine the language that someone is speaking
- Generative approach:
 - is to learn each language and determine as to which language the speech belongs to
- Discriminative approach:
 - is determine the linguistic differences without learning any language— a much easier task!

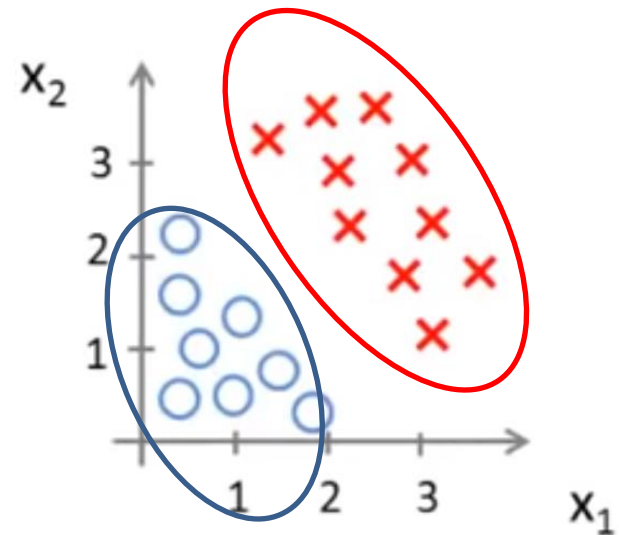
Generative and discriminative models

Discriminative model

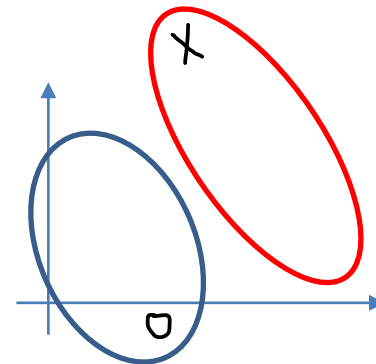


Logistic regression for classification

Generative model



Look at individual class, and build
a model for that class
e.g. model distribution



Generative and discriminative models

Discriminative model

Learn $p(y|x)$

“the probability of y given x ”

Class
e.g. malignant
tumor or benign

Features
(observation)

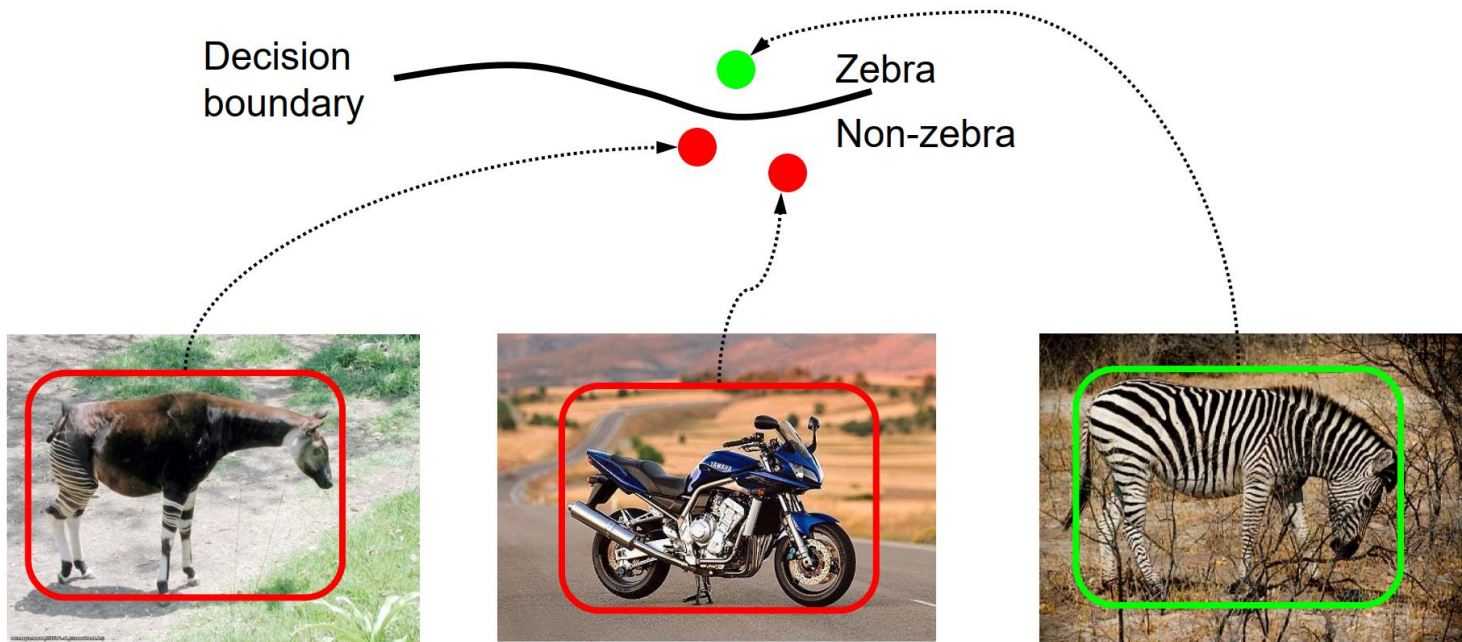
Generative model

Learn $p(x|y)$

What is probability of
having the feature
conditioned on class y

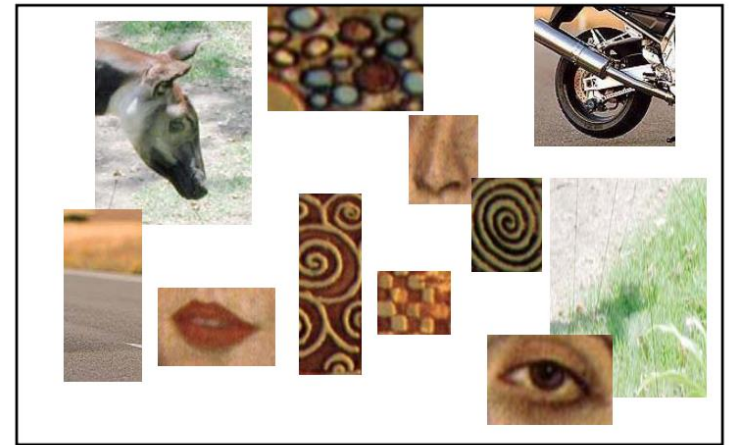
Discriminative methods

- Direct modeling of $p(\text{zebra} \mid \text{image})$







Generative methods

Model $p(image|zebra)$ and $p(image|no\ zebra)$



Generative vs Discriminative Models

	Generative Models (ex. Naïve Bayes)	Discriminative Models (ex. Logistic Regression)
Classification Accuracy		
Missing Features		

Generative and discriminative models

PDF: Probability Density Function

- **Generative Methods**

- Model class-conditional pdfs and prior probabilities
- “Generative” since sampling can generate synthetic data points
- Popular models
 - Gaussians, Naïve Bayes, Mixtures of multinomials
 - Mixtures of Gaussians, Mixtures of experts, Hidden Markov Models (HMM)
 - Sigmoidal belief networks, Bayesian networks, Markov random fields

- **Discriminative Methods**

- Directly estimate posterior probabilities
- No attempt to model underlying probability distributions
- Focus computational resources on given task– better performance
- Popular models
 - Logistic regression, SVMs
 - Traditional neural networks, Nearest neighbor
 - Conditional Random Fields (CRF)