



Midterm: wed, March 4
open book, notes
absolutely no communication – wireless or other

Machine Learning 10-601

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Midterm Course Correction Survey*

(5 min)

1. if you could change just one thing to improve the course, what should we do?
2. if you could pick just one thing to keep because it's helpful, what is that?

* anonymous survey

Machine Learning:

Study of algorithms that

- improve their performance P
- at some task $T \rightarrow \text{learn } f: X \rightarrow Y \text{ or } P(Y|X, \theta)$
- with experience $E \rightarrow \text{precollect data with fully obs vars. (EM). incompletely obs}$

$\rightarrow$ best fitting data (MLE for θ)
best fitting (data + priors) MAP

Some topics we've covered:

- Decision trees
 - overfitting, pruning
 - information gain, entropy
- Basic probability
 - MLE's, MAP
 - manipulating probabilities
 - Bayes rule
- Naïve Bayes
 - conditional indep.
 - Discrete, Gaussian
- Logistic regression
 - derive from Naïve Bayes
 - Conditional likelihood
 - gradient descent
- Logistic regression
 - derive from Naïve Bayes
 - Conditional likelihood
 - MLE, MAP, regularization
 - gradient descent
- Cross validation, feature selection, regularization,
- Linear regression
 - probabilistic interpretation
- Bayes nets
 - representation of joint distrib.
 - cond. indep./ D-separation
 - simple inference
 - learning from fully observed data
 - learning using EM
 - clustering using EM

learn $f: X \rightarrow Y$ or $P(Y|X)$

Given a learning task.

1. pick a ^{parameterized} representation - model class
hypothesis space

2. pick an objective fn.

det. $\epsilon \sim N(0, \sigma)$ fit the data - MLikelihood

$$Y = f(x) + \epsilon$$

MAP

$$\theta \leftarrow \underset{\theta}{\operatorname{argmax}} \prod_{\ell \text{ examples}} P(x^\ell y^\ell | \theta)$$

$$\theta \leftarrow \underset{\theta}{\operatorname{argmax}} P(\theta | \{x^\ell, y^\ell\}) \propto \prod_{\ell} P(x^\ell y^\ell | \theta) \cdot P(\theta)$$

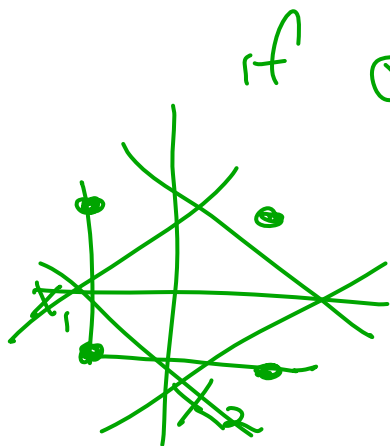
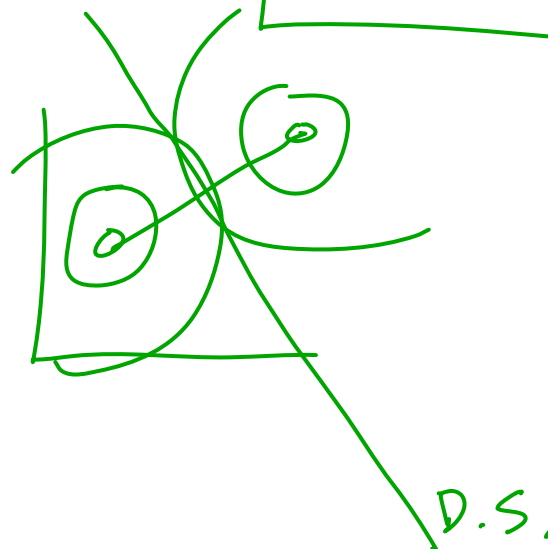
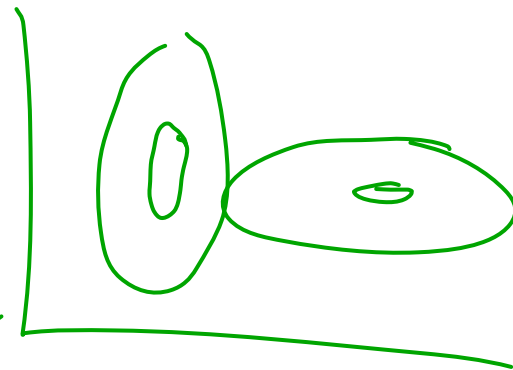
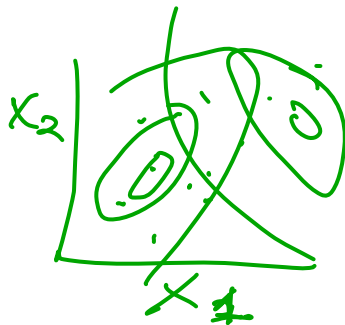
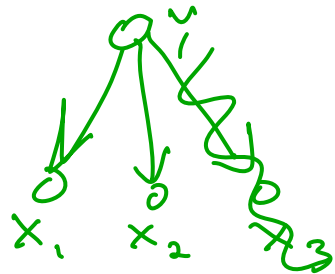
$$MLE = \underset{\theta}{\operatorname{argmax}} \prod_{\ell} P(y^\ell | x^\ell \theta)$$

Max
obs W
- obs Z

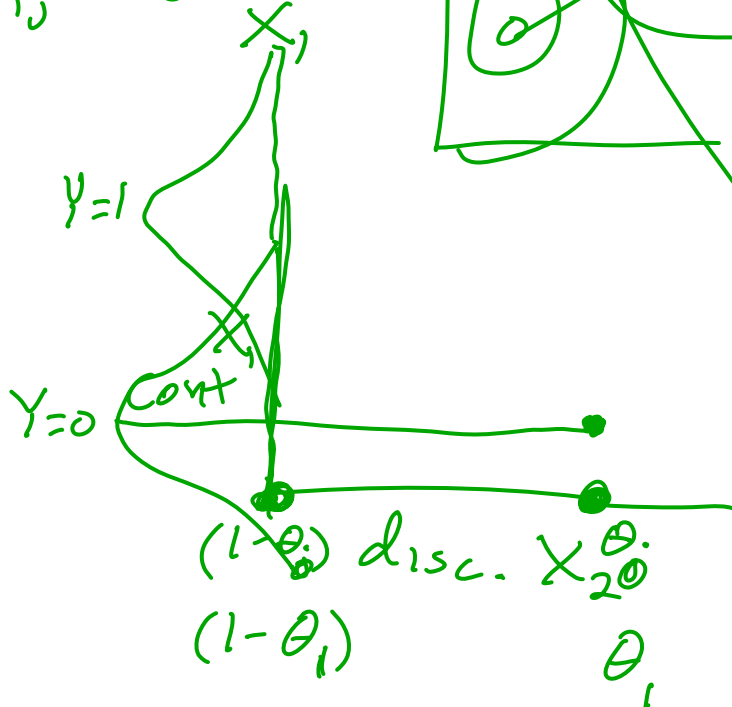
Likelihood

$$\underset{\theta}{\operatorname{argmax}} \mathbb{E}_{P(z|w)} \left[\prod_{\ell} P(w^\ell, z^\ell | \theta) \right] \leftarrow EM$$

continuous
GNB



if $\sigma_{ij} = \sigma$



$$P(x_i | y_i)$$

$$P(y|x) \propto P(y) \prod_i P(x_i | y)$$

$$\ln P(Y/X) = -m + \sum_i \ln \left(\frac{P(x_i | Y=1)}{P(x_i | Y=0)} \right)$$

log reg $P(Y/X) = \frac{1}{1 + \exp(\sum_i w_i x_i)}$

$P(Y=1|x) = \frac{\exp(w)}{1 + \exp(w)}$

$$\ln \frac{P(Y=1|x)}{P(Y=0|x)} = \sum_i w_i x_i$$

$$\ln \left(\frac{P(Y=1|x)}{P(Y \neq 1|x)} \right)$$

