

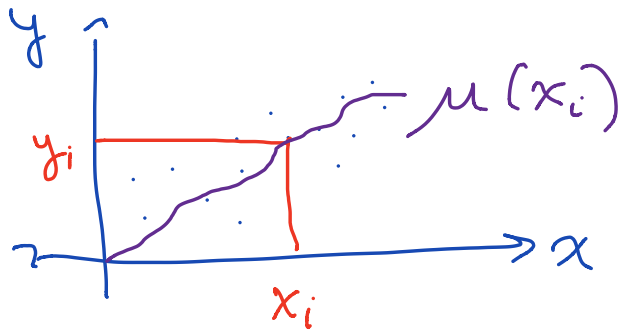
## Overview

Office Hrs: 10-12 a.m.

Ex: Linear regression

$$\{(y_i, \vec{x}_i) : 1 \leq i \leq n\}$$

$\underbrace{\quad}_{\text{response}} \quad \underbrace{\quad}_{\text{covariates}}$



$$\mathbb{E}(y_i | x_i) = \mu(x_i)$$

general regression

For a linear regression model  
we assume that this mean  
function is linear

$$\mathbb{E}[y_i] = \beta_0 + \beta_1 x_i$$

$$= \sum_{i=1}^n x_i y_i - x^T y$$

Well studied solution

OLS Estimator:  $\hat{\beta} = (X^T X)^{-1} X^T y$

We can show that  $\hat{\beta}$  is good  
(MSE, Unbiased, MLE, Consistent, ...)

But...

1. How do we compute  $\hat{\beta}$

- We don't need to calculate  $X^T X$  or even the inverse
- Just solve  $(X^T X) \hat{\beta} = X^T y$
- If  $X$  has a special structure we can use it in computations.
- What if we have missing data?

2. What if  $p \gg n$ ?

- Regularization
- Prior knowledge
- We can also select a subset of predictors to be included
- Dist. properties of  $\vec{\beta}$   
     ↳ Bootstrap, MCMC

## Topics

1. Intro, prob / Stat
2. Numerical Computing / LA
3. Basic Optimization & EM
4. Sampling & Monte Carlo

Stat  
Comp.

5. Graphical Models

6. MCMC

Comp.  
Stat.

7. Bootstrap

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## 0. Data Partitioning & Density