

statweb.stanford.edu/~owen/courses/305

Prerequisites

- Matrix algebra: eigenvalues, rank, Orthogonal matrices...
- Probability: normal, t, χ^2 , CLT, covariance
- Statistics: p-value, conf. intervals, hypothesis testing, regression
- Computation: R, python, Matlab, C
- Experience: fitting models, applying methods

Statistics is almost but not quite math

Statistics is almost but not quite computing

Modeling is tricky:

- hard to choose a model
- wrong assumptions can lead to right answers
- can't quite prove things about the world

Linear Models

have X predict $y \in R$

X arbitrary

data $(X_i, y_i)_{i=1, \dots, n}$

Least Square Error criteria:

Best predict of y :

for $X = x, \mu(x) = E(y|X)$

$$E([Y - m(x)]^2 | X = x) = E([Y - \mu + \mu - m(x)]^2 | X = x) = V(Y | X = x) + (\mu - m(x))^2 \geq V(Y)$$

However, if you do not believe in the mean square error criteria, the answer is different:

For absolute deviation loss,

$$E(|Y - m(X)| | X = x)$$

take $m(X) = \text{median}(Y | X = x)$

Alternate Proof (sketch)

Set $\frac{d}{dm} E((y - m)^2 | X = x) = 0$

Linear Model Examples

$$Y_i = \beta_0 + \beta_1 X_i + e_i$$

$$e \sim (0, \sigma^2)$$

(maybe normal)