# Logistic Regression

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#### References

http://www.stat.cmu.edu/~cshalizi/uADA/12/lectures/ch12.pdf

## Previously ...

- Predicting continuous outcomes conditioning Y on some values of X conditional expectations
- Not enough sometimes

## Why wasn't it enough?

- Discrete outcomes
- Assigning class labels to data classification
- Basic model "yes" or "no"
- Model to take noise into account stochastic model probabilities
- Conditional distribution of Y given input X
- P(Y|X) measure "precision" of prediction

### Bottom line -



#### Secret sauce

$$\log \frac{p(x)}{1 - p(x)} = \beta_0 + x \cdot \beta$$

#### How does this make sense?

- Linear dependence but 0,1 as o/p
- Sigmoid function [0,1]

$$S(t)=rac{1}{1+e^{-t}}.$$

## Decision boundary

- Find B's
- Plug into equations
- Eq >= 0 if Y = 1
- Figure represented is decision boundary
- Derivation of line, plane, non linear decision boundaries

# Logistic regression from scratch

# Logistic regression using sklearn

# Measuring performance

- Accuracy
- ROC receiver operating characteristic curves

## **ROC** curves

