

Regression

- Statistical method
- analyze, understand



relationship between 2+ variables

- Which factors are important
- " " " ignored
- how they influence each other

≥ 1 independent $\rightarrow 1$ dependent

— "negness"

— Prediction, Forecasting

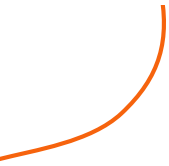
- Financial
- Marketing
- Manufacturing

— medicine

College Result	Admission
—	93
—	70
—	65
—	50
—	20

6

6



Clark College

Chance

—
—
—
—
—

Yes }
No }

Types |

- Linear

- Logistic

- Polynomial

Linear $\rightarrow X_n y$

$$y = a_1 x_1 + a_2 x_2 + \dots + a_n x_n + b$$

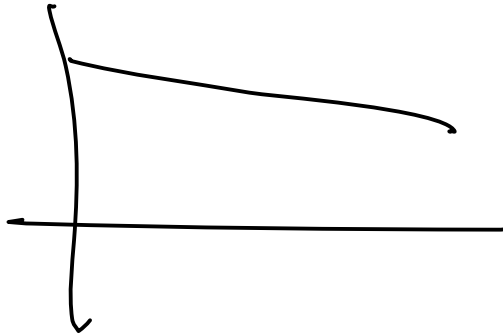
\downarrow
target variable

$x_1 \dots x_n = \text{features}$

$a_1 \dots a_n = \text{coefficients}$
 $b = \text{constant}$ } parameters

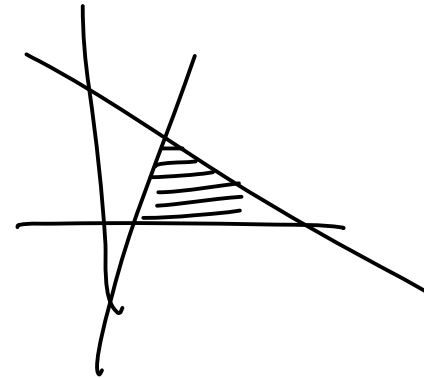
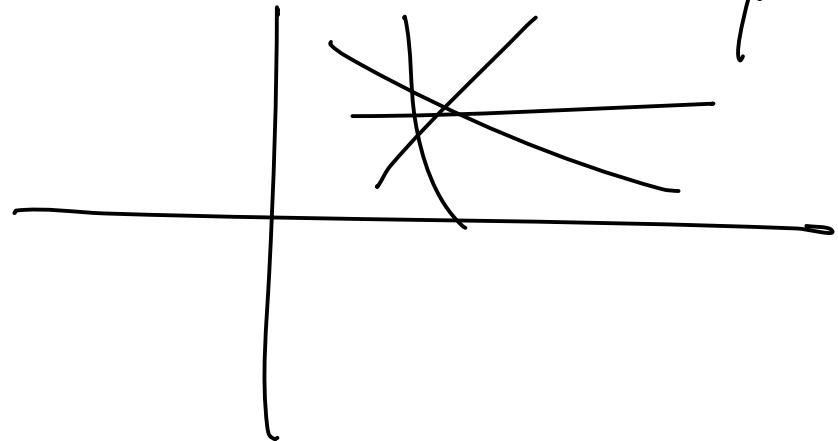
Simple

$X \rightarrow y$



Multiple

$X_1 \rightarrow y$
 $X_2 \rightarrow y$
 $X_3 \rightarrow y$
 $X_3 \rightarrow y$



Main factor

- Variance betⁿ variables

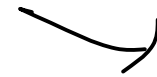
Assumption

- Linear Relationship
- Normality - Homoscedasticity
- Multicollinearity $\rightarrow R^2 / \text{var}$

$H_0 \rightarrow$ Equal Variance \nearrow Homoscedasticity

$H_a \rightarrow$ Diff

''



Heteroscedasticity

a

Multicollinearity

Tension	BP	Sugar $\pi(y)$
1	80	
4	140	

Logistic Regression → Logit

↳ classification

— dependent → categorical

Binary

Multinomial

Multinomial \rightarrow Ordered
 \searrow Nominal
(unordered)

Process

— Consider classes of dependent variables

— Probability of happening

↳ now

↳ different weight

↳ relationship Taniguchi

π

Sigmoid Function $\rightarrow 0, 1$

$$P(y=1) = \sigma(z)$$

$$= \frac{1}{1 + e^{-z}}$$

$$P(y=0) = 1 - \frac{1}{1 + e^{-z}}$$

Assumptions

- depend. var \rightarrow categorical

- No multicoline - _____

- Binary \rightarrow 1, 0

- Multiclass \rightarrow - - - - -

Linear Discriminant Analysis (LDA)

- classes known
- Future predict

- LR unstable

LDA

- Data small

- more than 2

→ Bayes's Theorem



Probability



$$P(y|x) = \frac{P(\dots)}{P(\dots)}$$

Regularized Linear Model

- Used when overfitting
- large # variables
- Multicollinearity \rightarrow

L1 Regularization → Lasso

- Objective function



minimize

penalty term

+ sum of absolute value of coefficients

Least Absolute Shrinkage

Selection Operator

(Lasso)

L2 Regularization \rightarrow Ridge

- minimize

- Penalty term $\rightarrow \lambda$

+ Sum of sq. of coeft

$$\text{Residual} = (\text{Actual} - \text{Predicted})$$

Loss function =

ordinary least squares

$$L = \alpha \cdot \sum (\text{error})^2$$

method

Elastic Net

$$L_1 + L_2$$

Mistakes

— Problemset X

— Binary class \rightarrow Logistic