Prediction

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Logistics

Homework 2 due Thursday before class

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 - e.g., we have a population of college w/ a major and the number of years in college
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- group and pivot are table operations that allow us to classify individuals according to multiple variable (or to 'cross-classify' them)

Quick demo to refresh our memory of group and pivot

Joining Tables by Columns

 When you have related data in multiple tables, you can 'join' by shared column

```
Table_1.join('Table_1_column_label', Table_2,
'Table_2_column_label')
```

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 - Sequential/temporal patterns
 What sequences of events occur frequently?

Example Patterns

Associative pattern

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Sequential patterns

After coming out of the bedroom in the morning, Bob turns off the bedroom lights, then goes to the kitchen where he makes coffee, and then leaves the house.

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o Probabilistic sequence models

If Bob turns on the TV in the evening then he will 80% of the time go to the kitchen to make popcorn.

What to Predict

- Behavior of Individuals
 - Location
 - Tasks / goals
 - Actions
- Behavior of the Environment
 - Device behavior (e.g. heating, AC)
 - o Interactions

Example: Location Prediction

- Where will Bob go next?
- Location_{t+1} = f(x)
- Input data x:
 - oLocation_t, Location_{t-1}, ...
 - oTime, date, day of the week
 - oSensor data

Example: Location Prediction

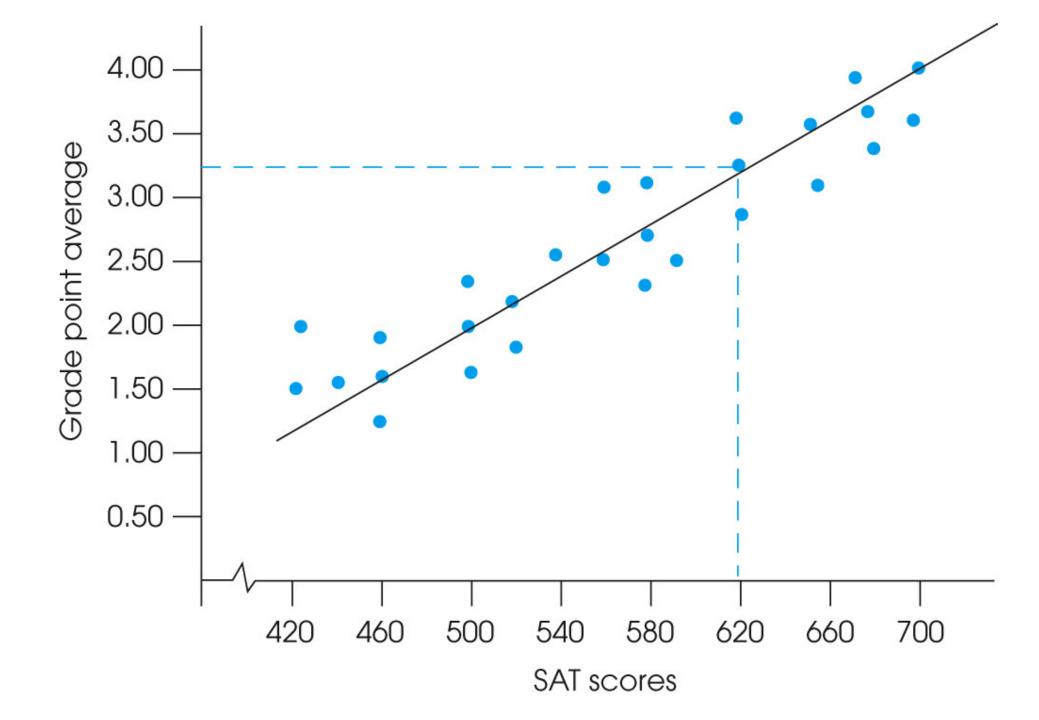
Time	Date	Day	Location _t	Location _{t+1}
6:30	02/25	Monday	Bedroom	Bathroom
7:00	02/25	Monday	Bathroom	Kitchen
7:30	02/25	Monday	Kitchen	Garage
17:30	02/25	Monday	Garage	Kitchen
18:00	02/25	Monday	Kitchen	Bedroom
18:10	02/25	Monday	Bedroom	Living room
22:00	02/25	Monday	Living room	Bathroom
22:10	02/25	Monday	Bathroom	Bedroom
6:30	02/26	Tuesday	Bedroom	Bathroom

Example: Location Prediction

- Learned pattern
 - o If Day = Monday...Friday
 - & Time > 0600
 - & Time < 0700
 - & Location_t = Bedroom
 - Then Location $_{t+1}$ = Bathroom

Prediction Techniques

- Classification-Based Approaches
 - Nearest Neighbor
 - Neural Networks
 - Bayesian Classifiers
 - o Decision Trees
- Sequential Behavior Modeling
 - Hidden Markov Models
 - o Temporal Belief Networks

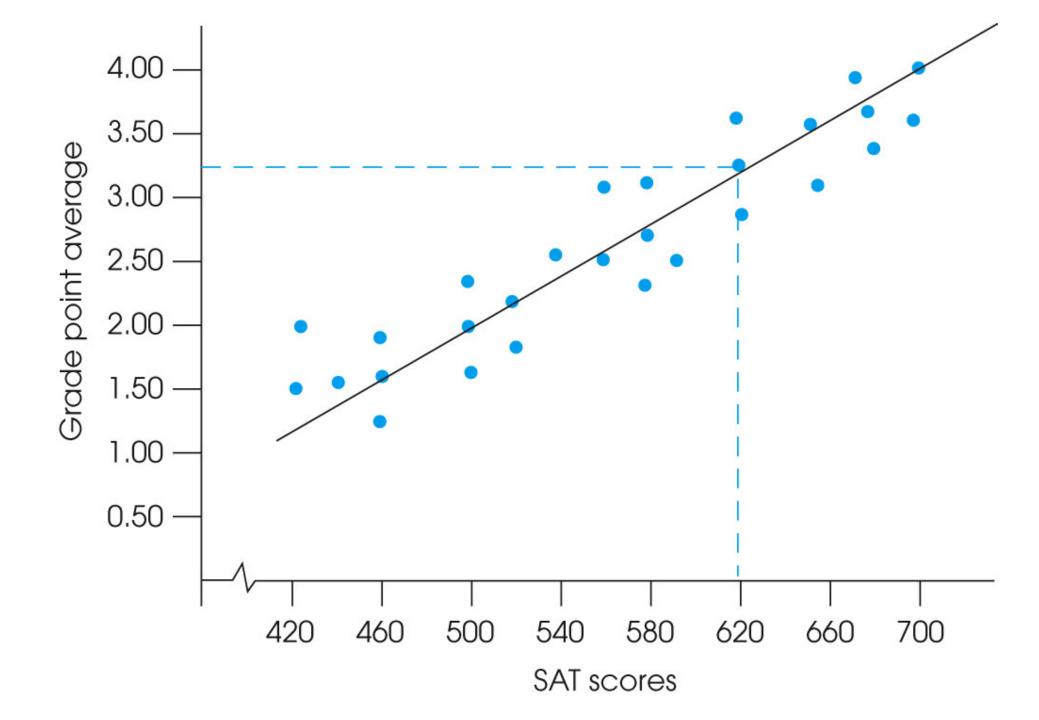


Introduction to Linear Regression

 Any straight line can be represented by an equation of the form y = mx + a, where m and a are constants.

• b: slope constant, determines the direction and degree to which the line is tilted (= $\Delta y - \Delta x$)

• a: the Y-intercept, determines the point where the line crosses the Y-axis.

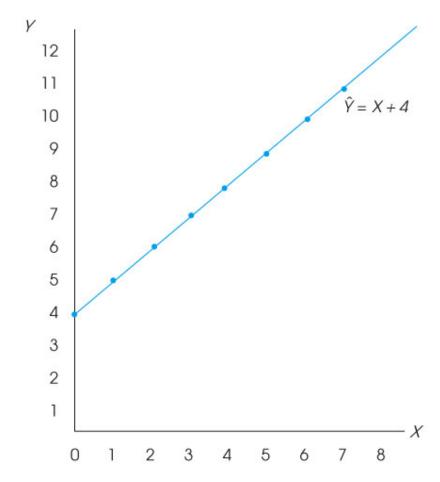


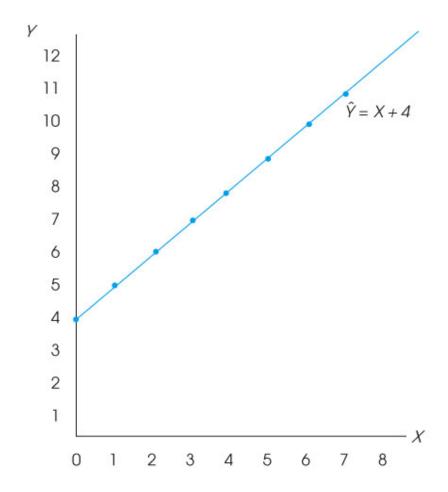
Introduction to Linear Regression

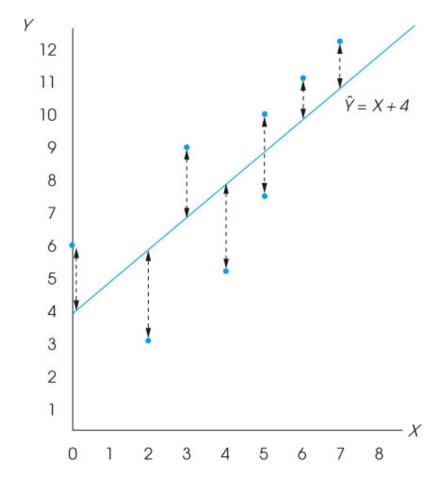
 How well a set of data points fits a straight line can be measured by calculating the distance between the data points and the line.

• The total error between the data points and the line is obtained by squaring each distance and then summing the squared values.

• Simple linear regression: find the equation of the straight line that produces the minimum sum of squared errors.







Introduction to Linear Regression

 Can use this 'line' to make predictions about the values of unseen or new data

Prediction: Sir Francis Galton

- 1822 1911 (knighted in 1909)
- A pioneer in making predictions
- Particular interest in heredity (father of eugenics)
- Charles Darwin's half-cousin

