# Price – Demand Curve (p. 111–114)

*Example.* A company is trying to determine how demand for a new product depends on its price and collect the following data:

price p	\$9	\$10	\$11
demand <i>d</i>	1200/mo.	1000/mo.	975/mo.

The company has reason to believe that price and demand are **inversely proportional**, that is,  $d = \frac{c}{p}$  for some constant c.

ightarrow Use the method of least squares to determine this constant *c*.



# Price – Demand Curve (p. 111–114)

**Solution.** Since 
$$f(p) = \frac{c}{p}$$
, then the sum  $S = \sum_{(p_i, d_i)} \left[ d_i - \left( \frac{c}{p_i} \right) \right]^2$ .  
Specifying datapoints gives

$$S = \left[1200 - \frac{c}{9}\right]^2 + \left[1000 - \frac{c}{10}\right]^2 + \left[975 - \frac{c}{11}\right]^2$$

Setting the derivative equal to zero gives

$$\frac{dS}{dc} = \frac{-2}{9} \left[ 1200 - \frac{c}{9} \right] + \frac{-2}{10} \left[ 1000 - \frac{c}{10} \right] + \frac{-2}{11} \left[ 975 - \frac{c}{11} \right] = 0$$

Solving for *c* gives  $c \approx 10517$ .



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## New York City Temperature (similar to p. 158)

The graph of average weekly temperature in New York City from Jan. 2006 to Dec. 2008 gives the distinct impression of a \_\_\_\_\_.

We need to determine the constants in:  $Temp(t) = A + B \sin(C(t - D)).$ 



Let's simplify our model to only determine amplitude B and vertical shift A. We must make assumptions about \_\_\_\_\_ and \_\_\_\_\_ D. We can assume that C =

For *D*, find when the sine passes through zero. Since January is cold and July is hot, the zero should occur in April; guess  $D \approx \frac{4}{12}$ .

Fitting to 
$$Temp(t) = A + B \sin[t - \frac{4}{12}]$$
  
gives:  $Temp(t) = 13.9 + 11.8 \sin[t - \frac{4}{12}]$ 



# Interpolation vs. Extrapolation

Suppose you have collected a set of *known* data points  $(x_i, y_i)$ , and you would like to estimate the *y*-value for an *unknown x*-value.

The name for such an estimation depends on the placement of the *x*-value <u>relative to the known *x*-values</u>.

### Interpolation

Inserting one or more *x*-values between known *x*-values.



#### Extrapolation

# Interpolation vs. Extrapolation

► The most common method for interpolation is taking a weighted average of the two nearest data points; suppose  $x_1 < x < x_2$ , then,  $f(x) \approx y_1 + \frac{y_2 - y_1}{x_1 - x_1} (x - x_1)$ 

$$f(x) \approx y_1 + \frac{y_2 - y_1}{x_2 - x_2}(x - x_1).$$

- ► In both interpolation and extrapolation, when you have a function f that is a good fit to the data, simply plug in y = f(x).
- Confidence in approximated values depend on confidence in your data and your model.
- Confidence in extrapolated data higher when closer to the range of known x-values.

# Extrapolation: Running the Mile (p. 162)

Below is a plot of the years in which a record was broken for running a mile and the record-breaking time. The data appears to fit a line; running least-squares gives



T(t) = 15.5639 - 0.00593323t

Solving for T(t)=0 gives  $t \approx 2623$ .

Conclusion: in the year 2623, the record will be zero minutes!

- Note the lack of descriptive realism.
- Always be careful when you extrapolate!