

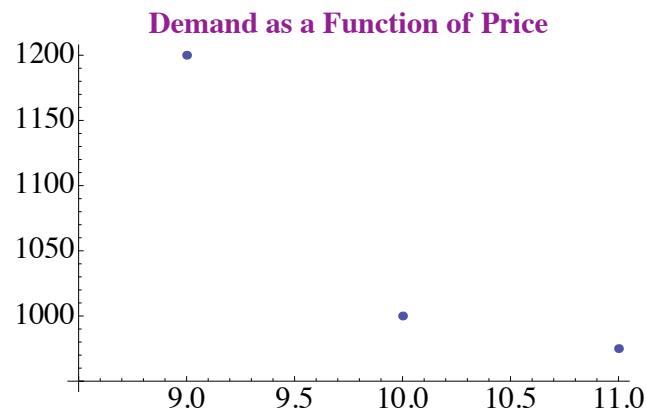
## Price – Demand Curve (p. 111–114)

*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 $d$	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  $c$ .

→ 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  $S = \sum_{(p_i, d_i)} \left[ d_i - \left( \frac{c}{p_i} \right) \right]^2$ .

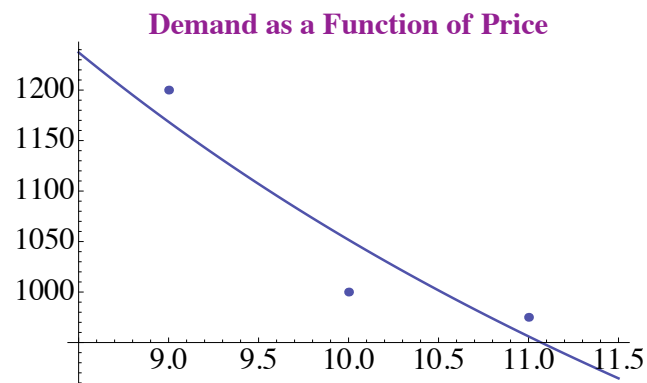
Specifying datapoints gives

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

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  $c \approx 10517$ .



# 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 \_\_\_\_\_.

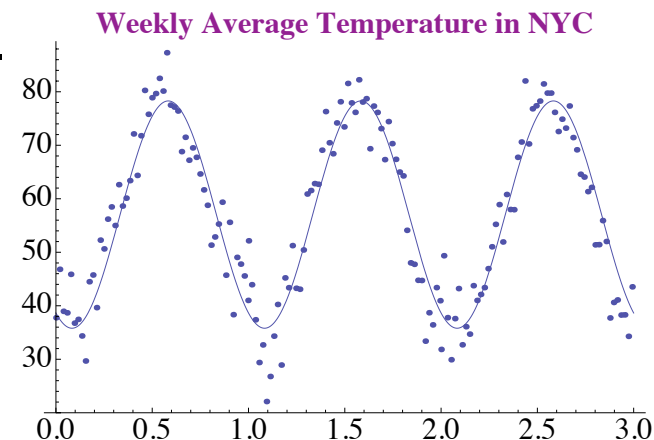
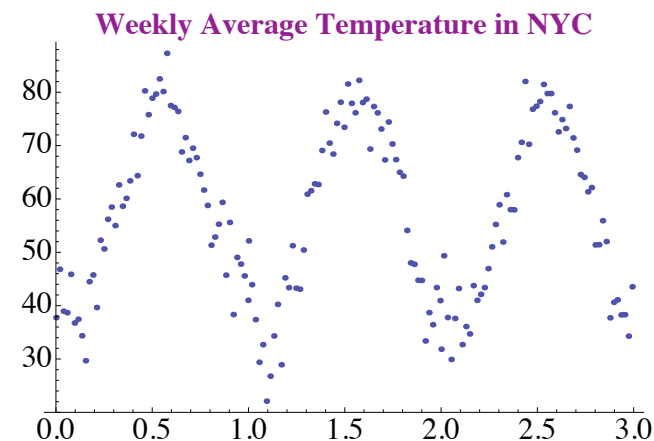
We need to determine the constants in:

$$Temp(t) = A + B \sin(C(t - D)).$$

Let's simplify our model to only determine amplitude  $B$  and vertical shift  $A$ . We must make assumptions about \_\_\_\_\_  $C$  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:  $Temp(t) = 13.9 + 11.8 \sin[t - \frac{4}{12}]$



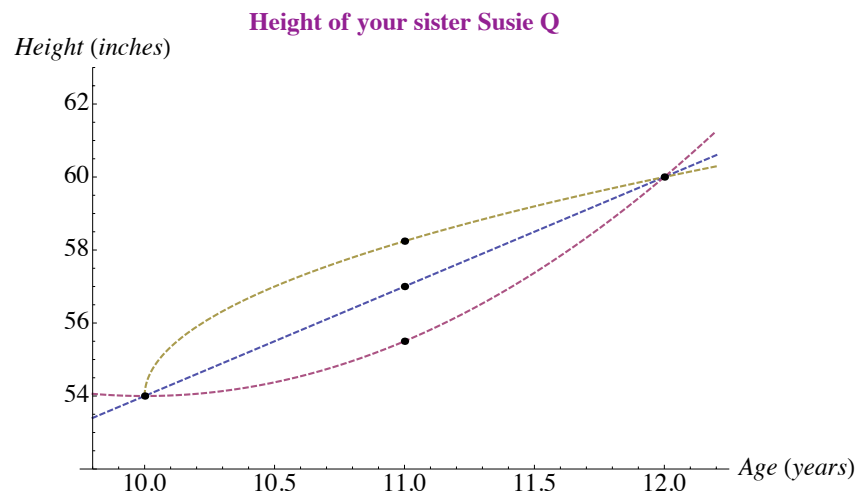
# Interpolation vs. Extrapolation

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 relative to the known  $x$ -values.

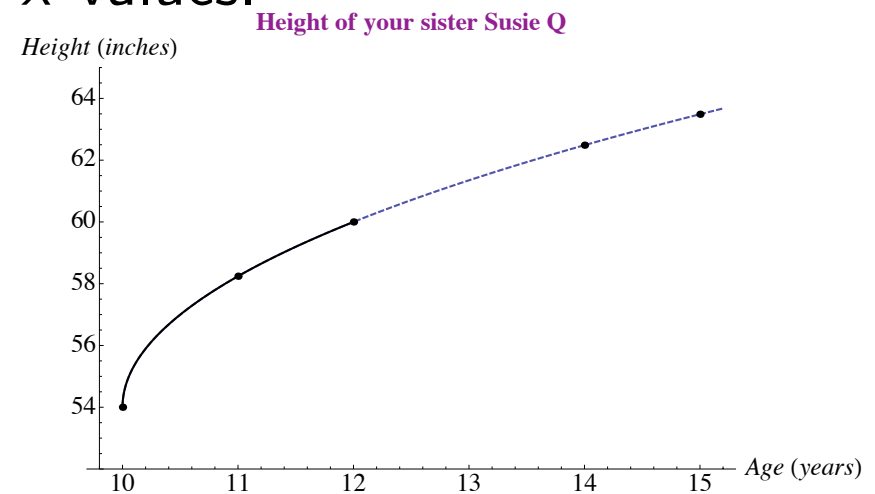
## Interpolation

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



## Extrapolation

Inserting one or more  $x$ -values outside of the range of known  $x$ -values.



# 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,

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

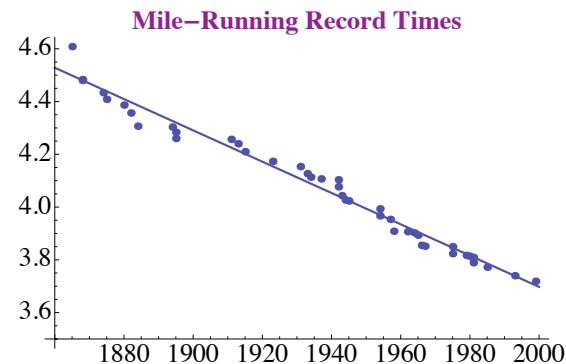
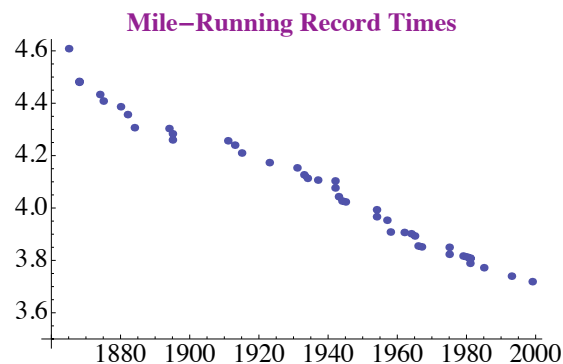
- ▶ In both interpolation and extrapolation, when you have a function  $f$  that is a good fit to the data, simply plug in  $y = f(x)$ .
- ▶ Confidence in approximated values depend on confidence in your data and your model.
- ▶ Confidence in extrapolated data higher when closer to the range of known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