



#### Introduction to the Course

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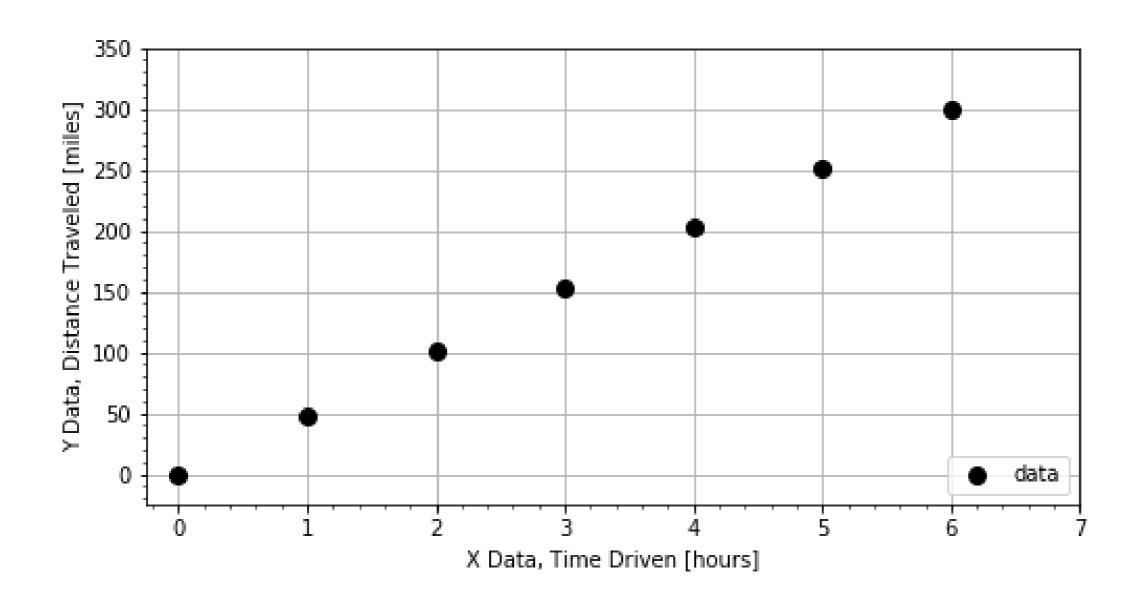
## Introduction to Chapter 1

#### Chapter Roadmap:

- Motivating Examples
- Data Visualization
- Descriptive Statistics



## Example Trip Data





#### Models as Descriptions

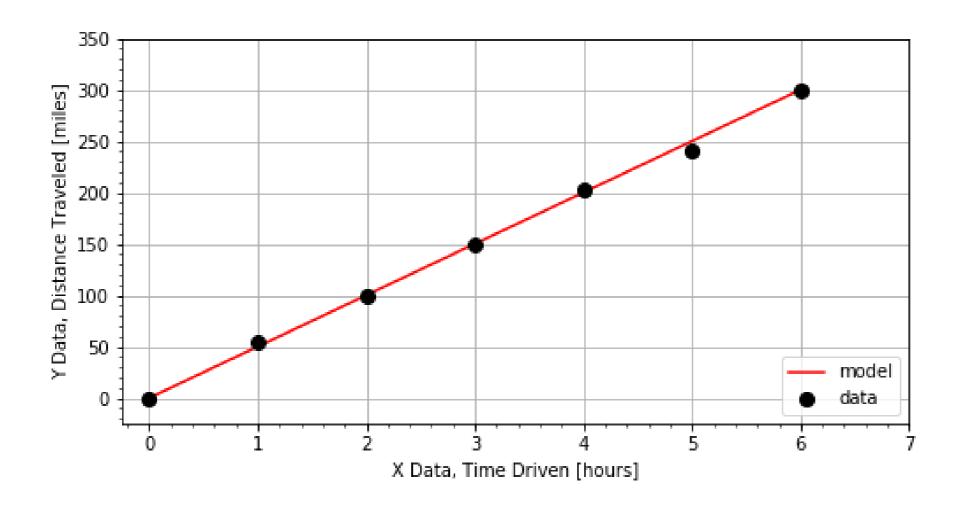
```
# range of y data
y_range = np.max(y) - np.min(y) = 300 - 0 = 300 # miles

# range of x data
x_range = np.max(x) - np.min(x) = 6 - 0 = 6 # hours

# estimating the speed
mph = y_range / x_range = 300 / 6 = 50
```



## Visualizing a Model





#### **Model Predictions**

```
# model as python expression
miles = 50*hours

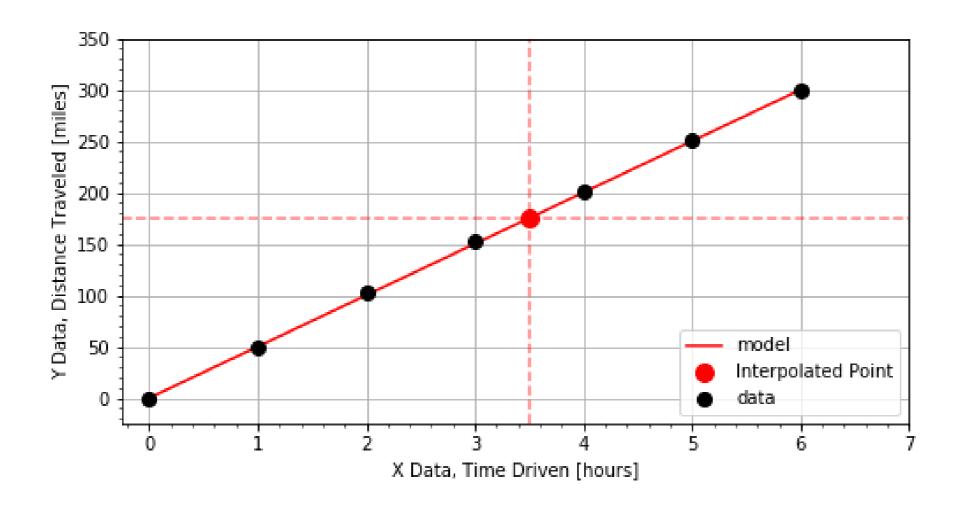
# model predicts distance is 300 miles at 6 hours
time = 6
distance = 50 * time = 50 * 6 = 300

def model(time):
    return 50*time

predicted_distance = model(time=10)
```

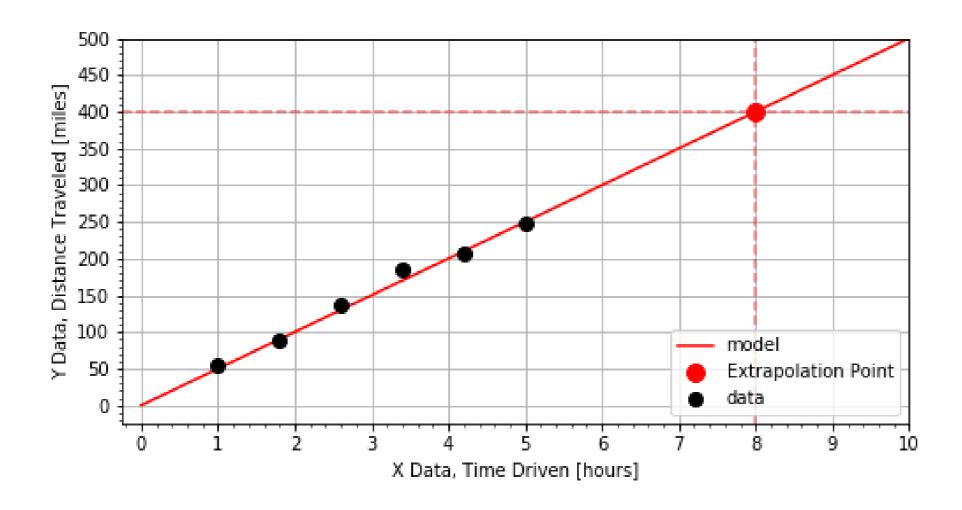


# Interpolation





## Extrapolation







# Let's practice!





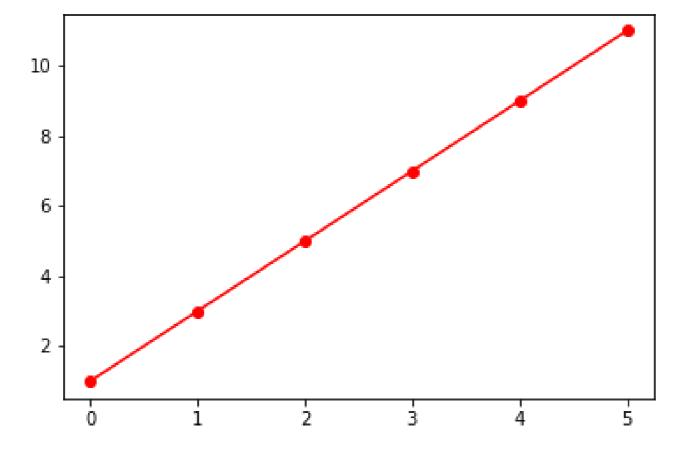
# Visualizing Linear Relationships

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#### **Quick Plots**

```
import matplotlib.pyplot as plt
plt.plot(x, y, 'r-o')
plt.show()
```





#### Object Interface

```
# Import the pyplot module
import matplotlib.pyplot as plt
# Create figure and axis objects
fig, axis = plt.subplots()
# prepare initial style options
options = dict(marker='o', color='blue')
# call the plot method on the axis object
line = axis.plot(x, y, **options)
# modify the axis object with set methods
  = axis.set ylabel('Times')
  = axis.set xlabel('Distances')
# display figure
plt.show()
```



#### Visualizing Linear Data

- two points:
  - $\blacksquare$  (x1, y1) = (0,0)
  - $\blacksquare$  (x2, y2) = (2,3)
- change in x and y:

$$dy = (y2 - y1) = 3 - 0$$

$$- dx = (x2 - x1) = 2 - 0$$

- slope = rise-over-run
  - slope = dy/dx = 3/2
- intercept:
  - when x=0, y=?







# Let's practice!





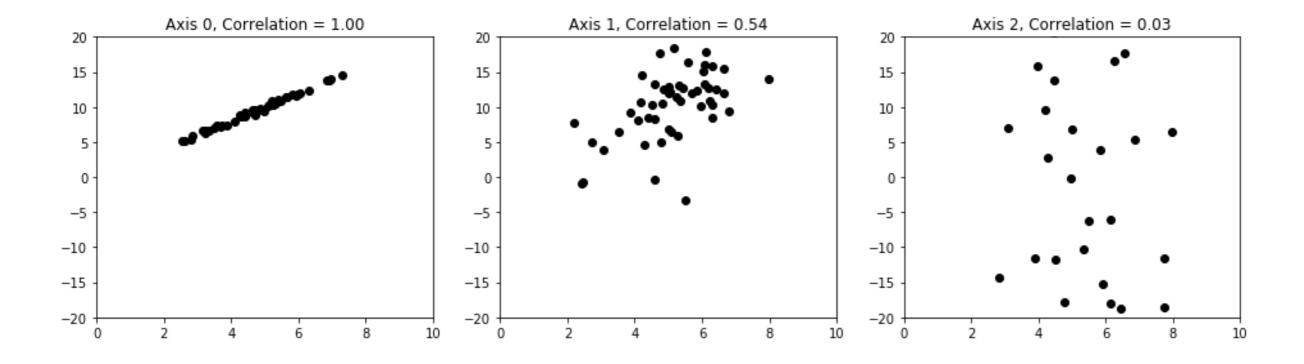
# Quantifying Linear Relationships

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#### Pre-Visualization





## Review of Single Variable Statistics

```
# Mean
mean = sum(x)/len(x)

# Deviation, sometimes called "centering"
dx = x - np.mean(x)

# Variance
variance = np.mean(dx*dx)

# Standard Deviation
stdev = np.sqrt(variance)
```



#### Covariance

```
# deviations of two variables
dx = x - np.mean(x)
dy = y - np.mean(y)

# co-vary means to vary together
deviation_products = dx*dy

# covariance as the mean
covariance = np.mean(dx*dy)
```



#### Correlation

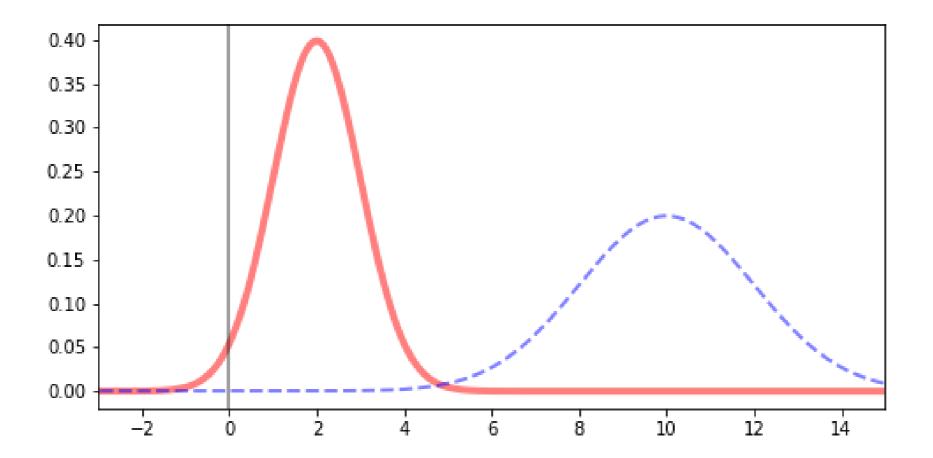
correlation = np.mean(zx\*zy)

```
# divide deviations by standard deviation
zx = dx/np.std(x)
zy = dy/np.std(y)

# mean of the normalize deviations
```

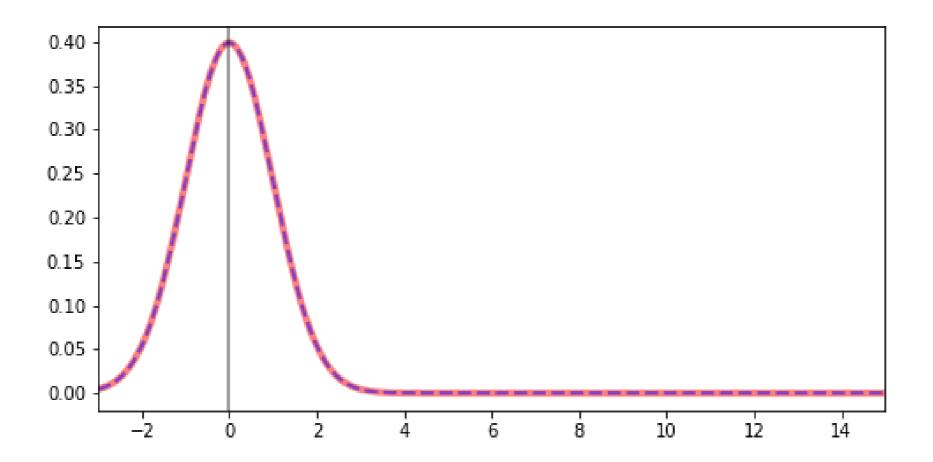


#### Normalization: Before





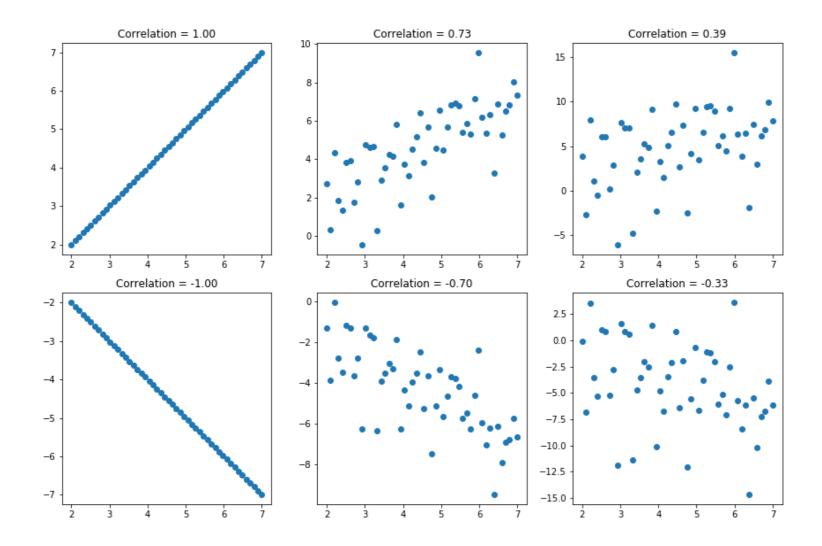
#### Normalization: After





#### Magnitude versus Direction

Correlation values: -1 to +1



• Two Parts: Magnitude (1 to 0) versus Sign (+ or -)





# Let's practice!