



# Linear Regression

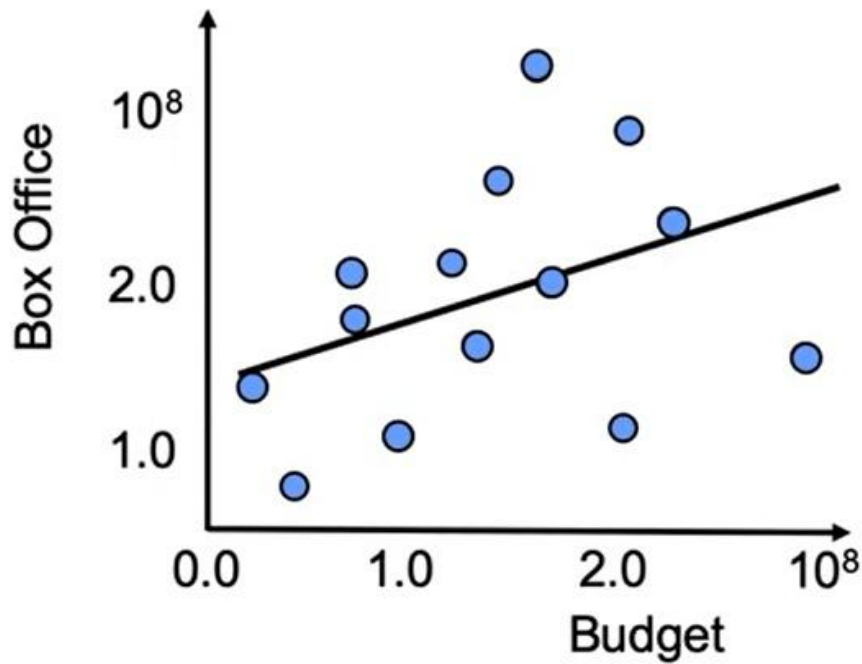
# Learning Goals

In this section, we will cover:

- Linear Regression
- Modeling Best Practice
- Measuring Errors



# Introduction to Linear Regression



$$y_{\beta}(x) = \beta_0 + \beta_1 x$$

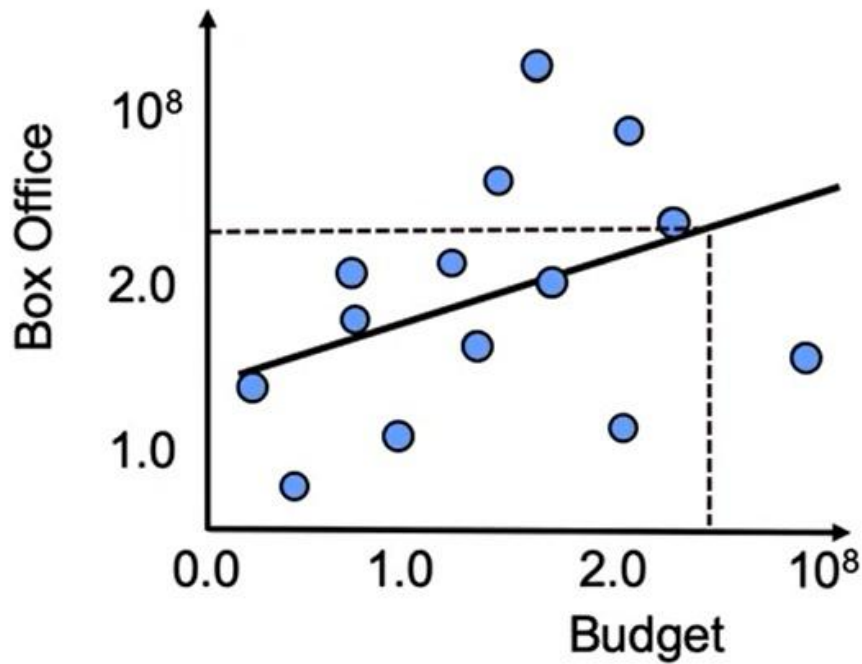
box office  
revenue

coefficient 0

coefficient 1

movie  
budget

# Introduction to Linear Regression



$$y_{\beta}(x) = \beta_0 + \beta_1 x$$

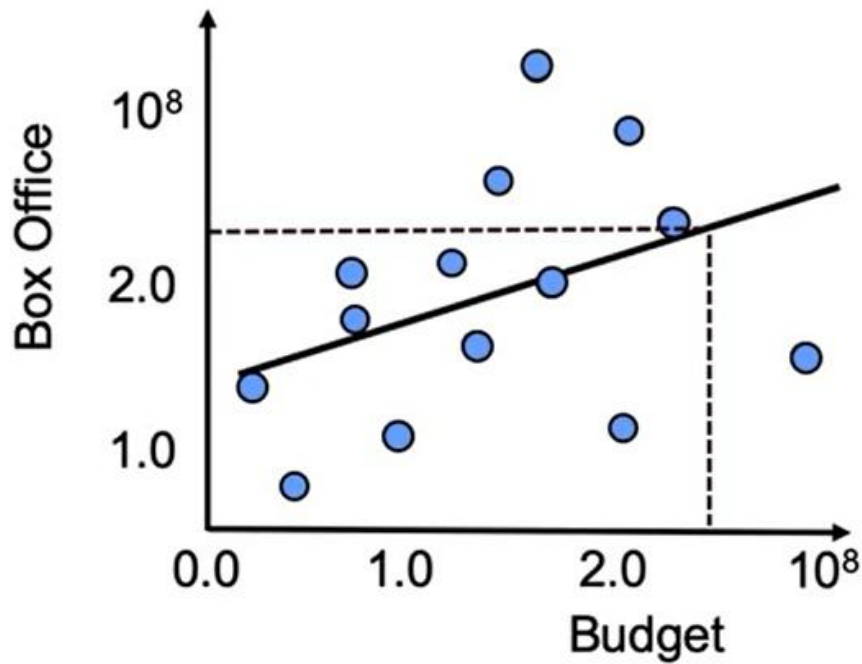
box office  
revenue

80 million

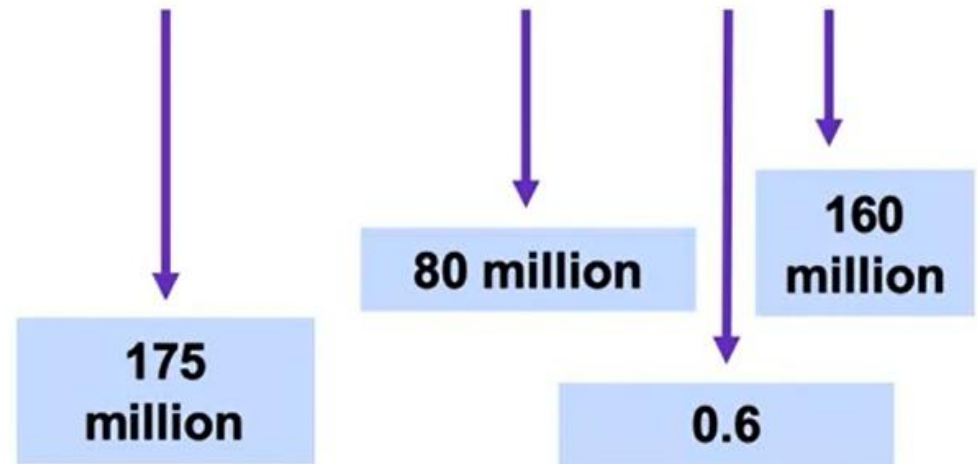
movie  
budget

0.6

# Introduction to Linear Regression

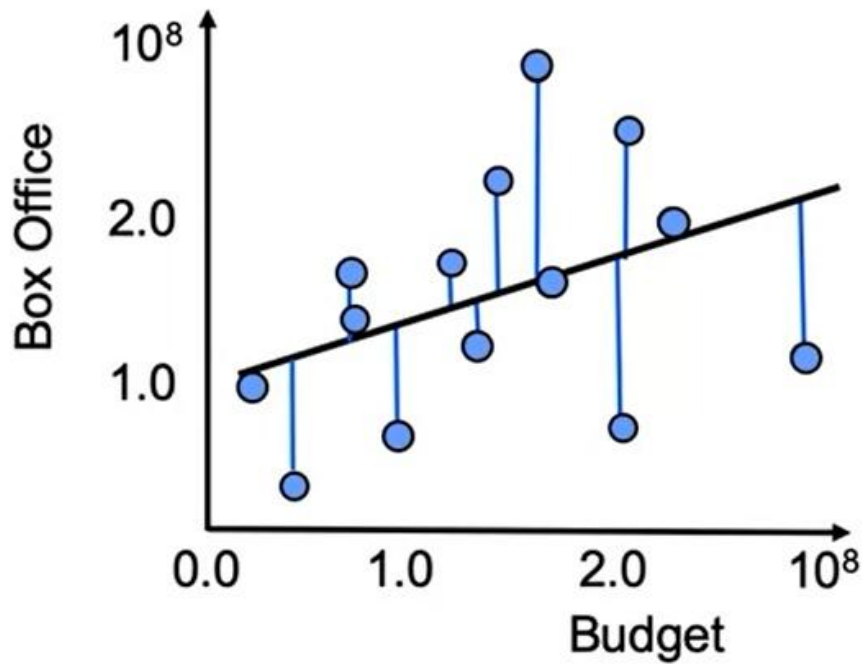


$$y_{\beta}(x) = \beta_0 + \beta_1 x$$





# Calculating the Residuals



$$y_{\beta} \left( x_{obs}^{(i)} \right) - y_{obs}^{(i)}$$

predicted  
value

observed  
value



# Minimizing the Error Function

$$\min_{\beta_0, \beta_1} \frac{1}{m} \sum_{i=1}^m \left( \left( \beta_0 + \beta_1 x_{obs}^{(i)} \right) - y_{obs}^{(i)} \right)^2$$



# Minimizing the Error Function

$$J(\beta_0, \beta_1) = \frac{1}{2m} \sum_{i=1}^m \left( \left( \beta_0 + \beta_1 x_{obs}^{(i)} \right) - y_{obs}^{(i)} \right)^2$$





# Modeling Best Practice

Use cost function to fit model

Develop multiple models

Compare results and choose best one



# Other Measures of Error

**Sum of Squared Error (SSE):** 
$$\sum_{i=1}^m \left( y_{\beta}(x^{(i)}) - y_{obs}^{(i)} \right)^2$$

**Total Sum of Squares (TSS):** 
$$\sum_{i=1}^m \left( \overline{y_{obs}} - y_{obs}^{(i)} \right)^2$$

**Coefficient of Determination ( $R^2$ ):** 
$$1 - \frac{SSE}{TSS}$$



# Linear Regression: The Syntax

Import the class containing the regression method

```
from sklearn.linear_model import LinearRegression
```

Create an instance of the class

```
LR = LinearRegression()
```

Fit the instance on the data and then predict the expected value

```
LR = LR.fit(X_train, y_train)
```

```
y_predict = LR.predict(X_test)
```

