

# Simple Linear Regression

# Linear Regression

- Linear Regression is a prediction/estimation method
- It requires you to estimate the parameters from the training data

# Linear Regression

- Estimate the relationship between a **dependent variable** (usually called Y) and an **independent variable** (usually called X)

$$Y = B_0 + B_1 \cdot X$$

Y – Output variable to be predicted

X – input variable given

B<sub>0</sub> and B<sub>1</sub> – parameters (or coefficients) to be estimated from training data.

# Linear Regression

- **Training** : Determine parameters  $B_0$  and  $B_1$  from the training data
- **Testing** : Use the estimated parameters  $B_0$  and  $B_1$  in the regression equation to predict the value of  $Y$  for given test data  $X$ .

# Linear Regression

- Estimated regression line based on sample data is given as:

$$\hat{Y} = B_0 + B_1 X$$

- The method of least squares chooses the values for  $B_0$ , and  $B_1$  to minimize the sum of squared errors:

$$SSE = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 = \sum_{i=1}^n (Y - B_0 - B_1 X)^2$$

# Linear Regression

In order to get the parameters for best fit lines, we need to:

1. Partially differentiate the SSE w.r.t  $B_0$  and equate to 0. Solve this equation for value of  $B_0$
2. Similarly, partially differentiate the SSE w.r.t  $B_1$  and equate to 0. Solve this equation for the value of  $B_1$

# Determine B0 and B1

$$B1 = \frac{\sum_{i=1}^n ((x_i - \text{mean}(x)) \times (y_i - \text{mean}(y)))}{\sum_{i=1}^n (x_i - \text{mean}(x))^2}$$

$$B0 = \text{mean}(y) - B1 \times \text{mean}(x)$$

# Example

- Example

The weekly advertising expenditure (x) and weekly sales (y) are presented in the following table.

y	x
1250	41
1380	54
1425	63
1425	54
1450	48
1300	46
1400	62
1510	61
1575	64
1650	71



# Example

- From previous table we have:

$$\begin{array}{lll} n=10 & \sum x = 564 & \sum x^2 = 32604 \\ & \sum y = 14365 & \sum xy = 818755 \end{array}$$

- The least squares estimates of the regression coefficients are:

$$b_1 = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} = \frac{10(818755) - (564)(14365)}{10(32604) - (564)^2} = 10.8$$

$$b_0 = 1436.5 - 10.8(56.4) = 828$$

# Example

- The estimated regression function is:

$$\hat{y} = 828 + 10.8x$$

$$\text{Sales} = 828 + 10.8 \text{ Expenditure}$$

- This means that if the weekly advertising expenditure is increased by \$1 we would expect the weekly sales to increase by \$10.8.

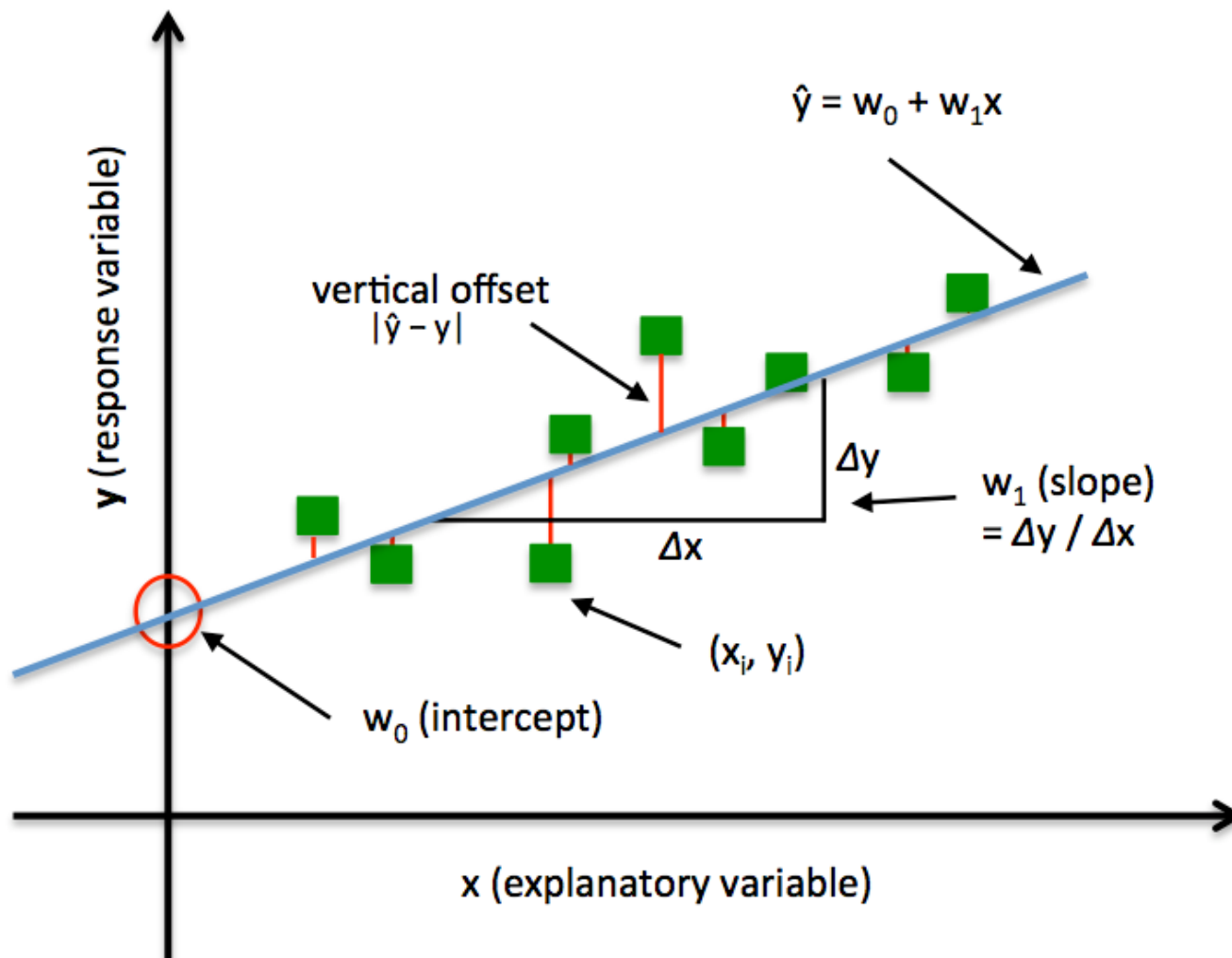
# Example

- Fitted values for the sample data are obtained by substituting the x value into the estimated regression function.
- For example if the advertising expenditure is \$50, then the estimated Sales is:

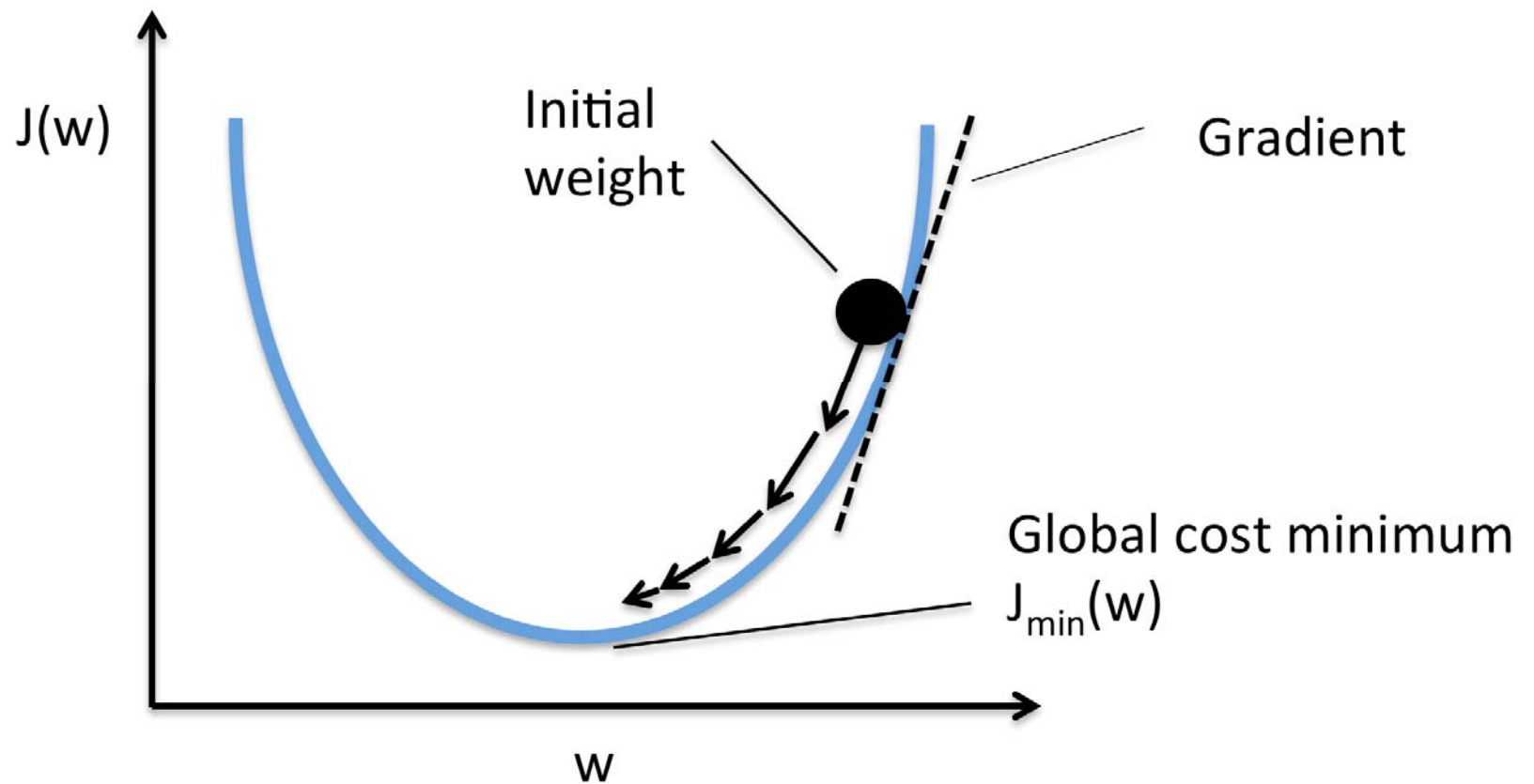
$$Sales = 828 + 10.8(50) = 1368$$

- This is called the point estimate (forecast) of the mean response (sales).

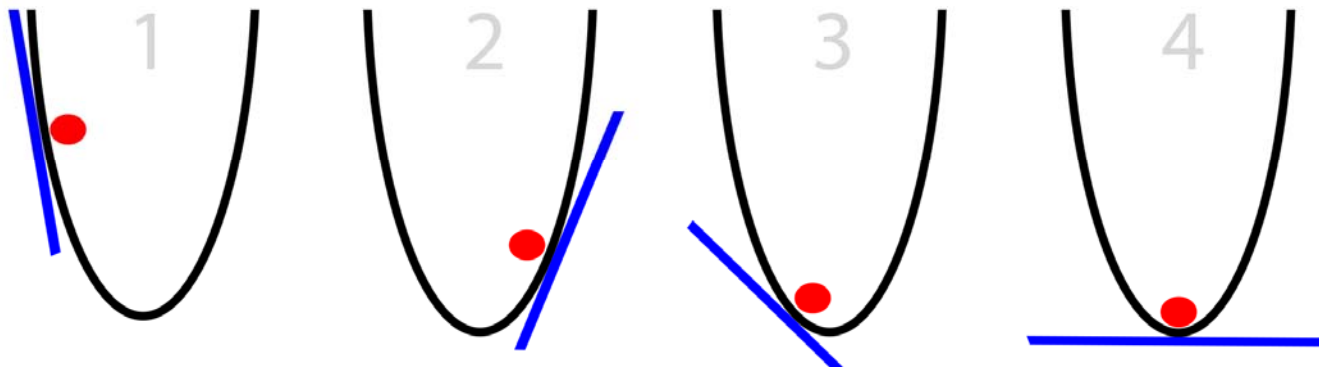
# Gradient Decent



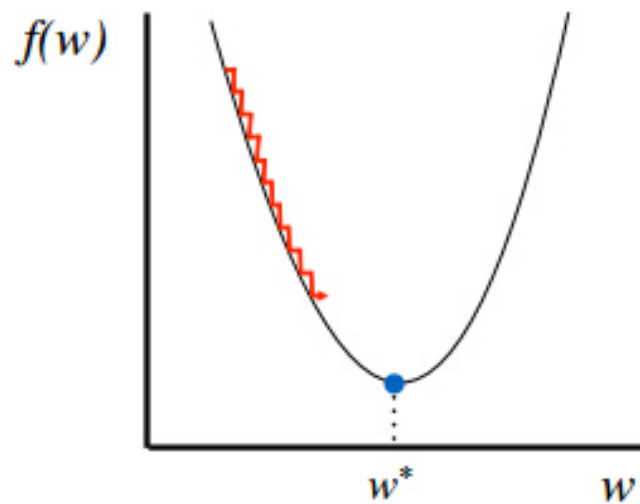
# Gradient Decent



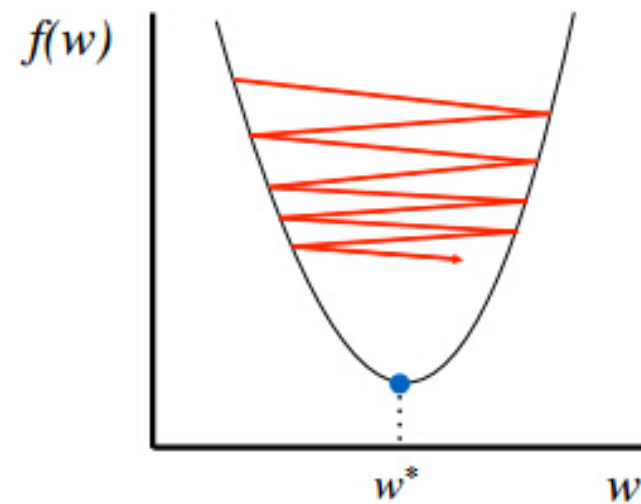
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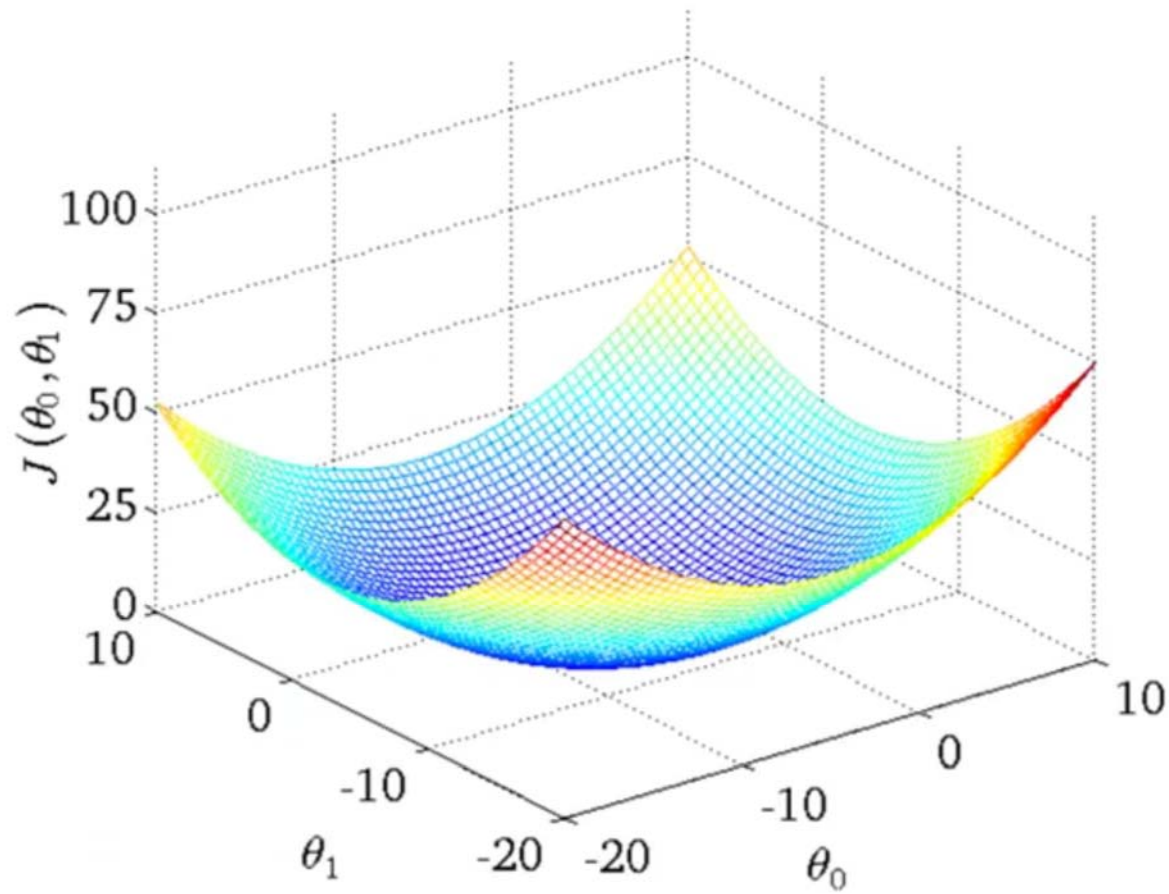


Too small: converge  
very slowly



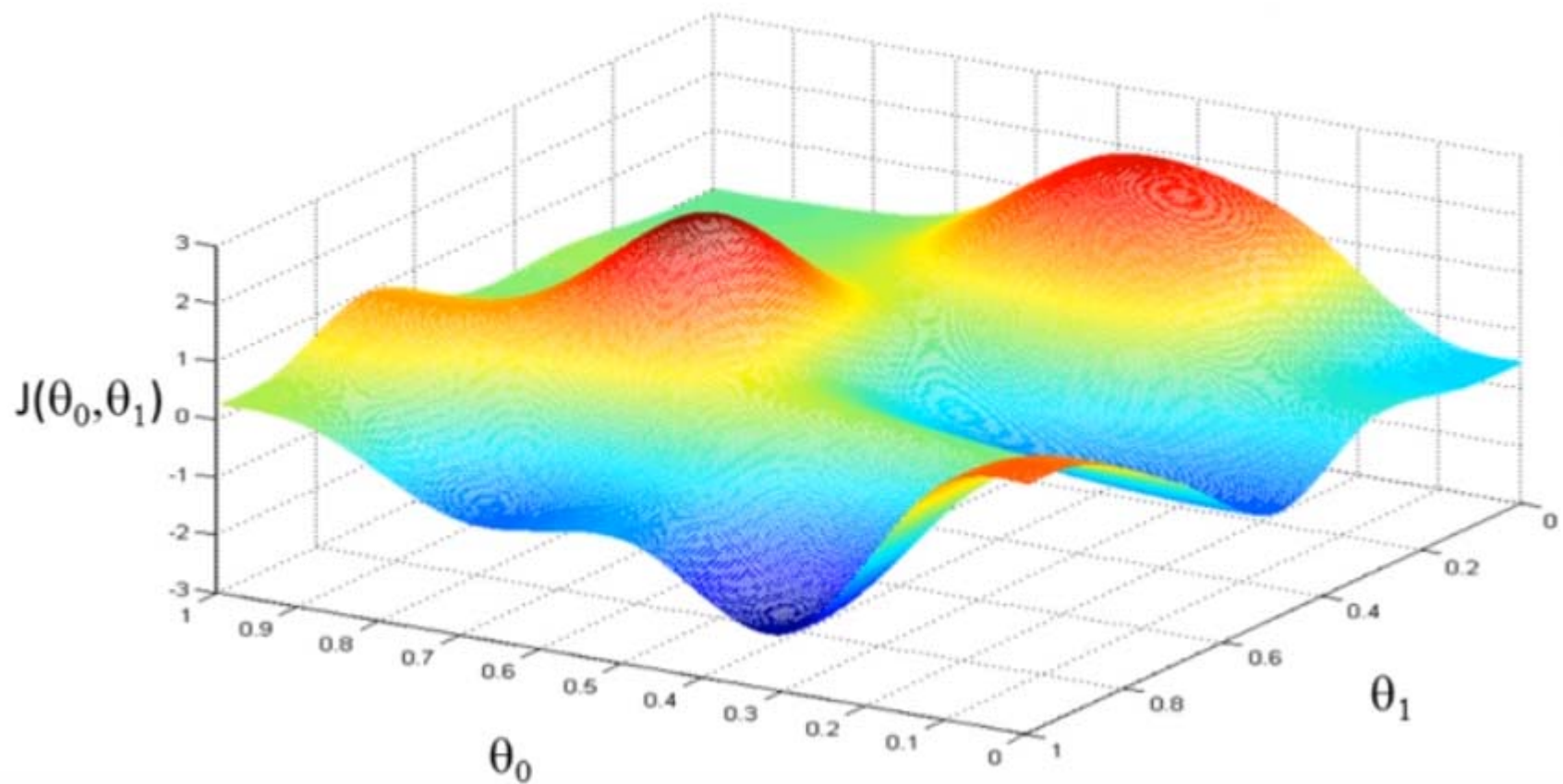
Too big: overshoot and  
even diverge

# Gradient Decent





# Gradient Decent



# Gradient Decent

