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Cost Function Or compute Cost (X, y, theta).

$$m = \text{length}(y)$$

$$J = 0$$

Same for Multi

$$h = X * \text{theta}$$

$$\text{squaredError} = (h - y) \cdot A^2$$

$$\underline{J} = \frac{1}{2 * m} * \text{sum}(\text{squaredError})$$

Gradient Descent (X, y, theta, alpha, num\_iters)

$$m = \text{length}(y)$$

$$J\text{-history} = \text{zeros}(\text{num\_iters}, 1)$$

for iter = 1 : num\_iters

$$t_1 = \text{theta}(1) - \text{alpha} * (1/m) * \text{sum}((X - \text{theta}) - y) \cdot X$$

$$t_2 = \text{theta}(2) - \text{alpha} * (1/m) * \text{sum}((X - \text{theta}) - y) \cdot X(:, 2)$$

$$\text{theta}(1) = t_1$$

$$\text{theta}(2) = t_2$$

## Gradient Descent Multivariable:- ( same same single

```
thetas = zeros(size(X, 2), 1);
```

```
// store computed theta value for a simultaneous update
```

```
for i = 1 : size(X, 2),
```

```
    t = theta(i) - alpha * (1/m) * sum((X * theta) - y) * X(:, i);
```

```
    thetas(i) = t
```

```
end
```

```
theta = thetas
```

## Feature Normalization

```
function [x_norm, mu, sigma] = FNormalize(X)
```

```
X_norm = X;
```

```
mu = zeros(1, size(X, 2));
```

```
sigma = zeros(1, size(X, 2));
```

```
mu = mean(X);
```

```
sigma = std(X);
```

```
for i = 1 : size(X, 2);
```

```
    norm = (X(:, i) - mu(i)) / sigma(i);
```

```
    X_norm(:, i) = norm
```

```
end
```