

# linear\_regression\_single\_variable

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## 1 Linear Regression using one variable

```
In [1]: import helpers as hlp
import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
```

### 1.1 Configuration

```
In [2]: # Configuration Parameters
learning_rate = 0.00001
epochs = 10000
display_step = 100
display_step_num = int(np.floor(
    epochs / display_step)) + 1 # +1 to include the very first step
```

### 1.2 Import Data

```
In [6]: # Import data
train_data = np.loadtxt(
    "../.../data/ex1/ex1data1.txt", dtype="float64", delimiter=",")
train_x_unbiased = np.delete(train_data, -1, axis=1)
train_x = np.insert(train_x_unbiased, 0, 1, axis=1)
train_y = np.delete(train_data, np.s_[:-1], axis=1)
```

### 1.3 Visualize Data

```
In [4]: # Visualize Data
labels = {
    "figure_labels": ["Food Truck Profit Prediction"],
    "x_labels": ["Population of City in 10,000s"],
    "y_label": "Profit in $10,000s",
}

hlp.plot_unclassified_data(train_x_unbiased, train_y, **labels)
```

<generator object macro at 0x107464e60>

## 1.4 Define Model

```
In [5]: # Parameters
x = tf.placeholder(tf.float64, name="x")
y = tf.placeholder(tf.float64, name="y")
m = train_y.shape[1] # number of training examples

# Model
theta = tf.Variable(
    tf.zeros(
        (train_x.shape[1], 1), dtype=tf.float64), name="theta")
prediction = tf.matmul(x, theta)

# Cost Function
cost = tf.reduce_sum(1 / 2 * m * tf.pow(tf.subtract(prediction, y), 2))

# Optimizer
optimizer = tf.train.GradientDescentOptimizer(learning_rate).minimize(cost)
```

## 1.5 Initialization

```
In [6]: # Initialize Session
sess = tf.InteractiveSession()
tf.global_variables_initializer().run()

file_writer = tf.summary.FileWriter("logs/single/run1", sess.graph)
```

## 1.6 Run Model

```
In [7]: costs = np.zeros(display_step_num)
cost_indices = np.zeros(display_step_num)

for epoch in range(epochs):
    sess.run(optimizer, feed_dict={x: train_x, y: train_y})

    if epoch % display_step == 0:
        display_step_cur_num = int(epoch / display_step)
        costs[display_step_cur_num] = cost.eval(
            feed_dict={x: train_x,
                       y: train_y})
        cost_indices[display_step_cur_num] = epoch

    costs[display_step_num - 1] = cost.eval(feed_dict={x: train_x, y: train_y})
    cost_indices[display_step_num - 1] = epochs - 1
```

## 1.7 Plot Cost Function

```
In [8]: plt.figure("Cost Function")
plt.xlabel("Iterations")
```

```
plt.ylabel("Cost")

plt.plot(cost_indices, costs)
plt.show()
```

<generator object macro at 0x107464e60>

## 1.8 Plot Fitted Line

```
In [9]: hlp.plot_unclassified_data(train_x_unbiased, train_y, show=False, **labels)
        hlp.plot_fitted_line(train_x_unbiased, theta.eval(), padding=0.35)
```

<generator object macro at 0x107464518>

## 1.9 Results

```
In [10]: print("Theta: \n" + str(theta.eval()) + "\n")
         print("Final Cost: \n" + str(cost.eval(feed_dict={x: train_x, y: train_y})))
```

```
Theta:
[[-3.20402165]
 [ 1.12353893]]
```

```
Final Cost:
438.492044623
```

## 1.10 Close Session

```
In [11]: sess.close()
```