```
Linear Regression with one variable
         f(x) = wx + b
          import numpy as np
          import matplotlib.pyplot as plt
In [3]:
          #x_train is size in 1000 sq feet
          #y_train is target price in 1000 USD
          x_{train} = np.array([1.0, 2.0])
          y_{train} = np.array([300.0, 500.0])
          #printing input variable and target price as f string
          print(f"x_train = {x_train}")
          print(f"y_train = {y_train}")
          x_{train} = [1. 2.]
          y_{train} = [300. 500.]
In [4]:
         # m = number of training examples
          # Numpy .shape returns a python tuple with an entry for each dimension
          print(f"x_train.shape: {x_train.shape}")
          x_train.shape: (2,)
          \# x_train.shape[0] is the length of the array and number of examples
          m = x_{train.shape[0]}
          print(f"Number of training example is: {m}")
          Number of training example is: 2
 In [6]:
          #Also len() function can be used
          m = len(x_train)
          print(f"Number of training example is: {m}")
          Number of training example is: 2
 In [7]:
          # i is ith training example
          #Python is zeroth index
          # So (x^0, y^0) is 1.0, 300.0
          # (x^1, y^1) is 2.0, 500.0
          #To access a value in a Numpy array, one indexes the array with the desired offset.
          #For example the syntax toaccess location zero ofx_train is x_train[0]
          i=0
          x_i = x_{train[i]}
          y_i = y_train[i]
          print(f''(x^{i}), y^{(i)}) = (\{x_i\}, \{y_i\})'')
          (x^0), y^0(0) = (1.0, 300.0)
 In [8]:
          i=1
          x_i = x_{train[i]}
          y_i = y_train[i]
          print(f''(x^{i}), y^{(i)}) = (\{x_i\}, \{y_i\})'')
          (x^1), y^1) = (2.0, 500.0)
In [9]:
          #Plotting the data frame
          #The function arguments marker and c show the points as red crosses
          # c is color r is 'red'
          plt.scatter(x_train, y_train, marker='x', c='r')
          # Set the title
          plt.title("Housing Prices")
          # Set the y-axis label
          plt.ylabel('Price (in 1000s of dollars)')
          # Set the x-axis label
          plt.xlabel('Size (1000 sqft)')
          plt.show()
                                Housing Prices
            500
            475
          950 dollars)
          000 400
375
          Pic 320
            325
            300
                        1.2
                                        1.6
                                                1.8
                                                         2.0
                1.0
                                1.4
                                Size (1000 sqft)
In [11]:
          \# f(x) = wx + b represents straight lines
          # We start with w= 100 and b= 100
          w= 100
          b= 100
          print(f"w: {w}")
          print(f"b: {b}")
         w: 100
          b: 100
          # Computing f(x) for two data points
          # For x(0), f_wb = w * x[0] + b
          # For x(1), f_wb = w * x[1] + b
          # For a large number of data points, this can get unwieldy and repetitive.
          # So instead, you can calculate thefunction output in aforloop as shown in thecompute_model_output function below
          \# Note: The argument description (ndarray (m, )) describes a Numpy n-dimensional arrayof shape (m, ).
          #(scalar)describes an argument without dimensions, just a magnitude.
          # Note : np.zero(n) will return a one-dimensional numpy array with entries
In [18]:
          def compute_model_output(x,w,b):
          #"""
          #Computes the prediction of a linear model
          #x (ndarray (m,)): Data, m examples
          #w, b (scalar) : model parameters
          #y (ndarray (m,)): target values
              m = x.shape[0]
              f_{wb} = np.zeros(m)
              for i in range(m):
                  f_wb[i] = w^* x[i] + b
              return f_wb
          # We will call the compute_model_output function and plot the output
In [19]:
          tmp_f_wb = compute_model_output (x_train, w, b,)
          # Plot our model prediction
          plt.plot(x_train, tmp_f_wb, c='b', label= 'Our Prediction')
          # Plot the data points
          plt.scatter(x_train, y_train, marker='x', c='r', label='Actual Values')
          # Set the title
          plt.title("Housing Prices")
          # Set the y-axis label
          plt.ylabel('Price (in 1000s of dollars)')
          # Set the x-axis label
          plt.xlabel('Size (1000 sqft)')
          plt.legend()
          plt.show()
                               Housing Prices
            500

    Our Prediction

                 × Actual Values
         n 1000s of dollars)
007
0450
050
050
          € 300
        Price 250
            200
                        1.2
                                                         2.0
                1.0
                                1.4
                                        1.6
                                                1.8
                                Size (1000 sqft)
          # We can see setting w = 100, b = 100 does not result in a line that fits our data
In [20]:
          W = 200
          b = 100
          x_i = 1.2
          cost_1200sqft = w * x_i + b
          print(f"${cost_1200sqft:.0f}thousand dollars")
          $340thousand dollars
 In [ ]:
          #Linear regression builds a model which establishes a relationship between features and targets
          #In the example above, the feature was house size and the target was house price
          #for simple linear regression, the model has two parameters and whose values are 'fit' usingtraining data
          #once a model's parameters have been determined, the model can be used to make predictions onnovel data.
```