```
In [ ]: import sys
        import sklearn
        import numpy as np
        import os
        import pandas as pd
        # Checking versions
        # Python ≥ 3.5 required else it will raise assertionError
        # Scikit-Learn ≥0.20 is required
        assert sys.version info >= (3, 5)
        assert sklearn. version >= "0.20"
        # to make this notebook's output stable across runs
        np.random.seed(42)
        # To plot pretty figures
        # The "%matplotlib inline" magic command ensures that any Matplotlib p
        lots will be displayed directly in the output cells of the notebook.
        %matplotlib inline
        import matplotlib as mpl
        import matplotlib.pyplot as plt
        mpl.rc('axes', labelsize=14)
        mpl.rc('xtick', labelsize=12)
        mpl.rc('ytick', labelsize=12)
        # Where to save the figures
        # This code defines a function for saving Matplotlib figures to a spec
        ified directory.
        PROJECT ROOT DIR = "."
        CHAPTER ID = "training linear models"
        IMAGES PATH = os.path.join(PROJECT ROOT DIR, "images", CHAPTER ID)
        os.makedirs(IMAGES PATH, exist ok=True)
        def save fig(fig id, tight layout=True, fig extension="png", resolutio
        n=300):
            path = os.path.join(IMAGES PATH, fig id + "." + fig extension)
            print("Saving figure", fig id)
            if tight layout:
                plt.tight layout()
            plt.savefig(path, format=fig extension, dpi=resolution)
```

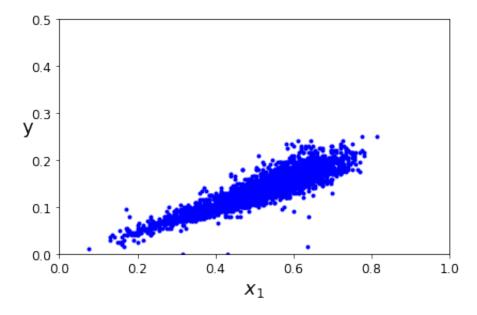
```
In [ ]: # plotting x and y with blue dot
plt.plot(X, y, "b.")

# the label is "$x_1$", which is typeset in LaTeX font
plt.xlabel("$x_1$", fontsize=18)

# rotation representing how much the label is rotated degrees
plt.ylabel("y", rotation=0, fontsize=18)

# assigning x-axis and y-axis limit
plt.axis([0, 1, 0, 0.5])
save_fig("generated_data_plot")
plt.show()
```

## Saving figure generated\_data\_plot



```
In [ ]: rows = len(data.axes[0])
         cols = len(data.axes[1])
         # print(rows)
         # print(cols)
         X_b = np.c_{np.ones((3320, 1)), X]} # add x0 = 1 to each instance
         theta best = np.linalg.inv(X b.T.dot(X b)).dot(X b.T).dot(y)
In [ ]: theta best
Out[]: array([[-0.0108267],
                [ 0.28716253]])
In [ ]: | X new = np.array([[0], [2]])
         X_{new_b} = np.c_{[np.ones((2, 1)), X_{new}]} \# add x0 = 1 to each instance
         y predict = X new b.dot(theta best)
         y predict
Out[]: array([[-0.0108267],
                [ 0.56349837]])
In [ ]: plt.plot(X_new, y_predict, "r-")
         plt.plot(X, y, "b.")
         plt.axis([0, 1, 0, 0.5])
         plt.show()
         0.5
          0.4
          0.3
          0.2
          0.1
          0.0
                    0.2
                            0.4
                                     0.6
                                             0.8
           0.0
                                                      1.0
```

```
In [ ]: plt.plot(X_new, y_predict, "r-", linewidth=2, label="Predictions")
    plt.plot(X, y, "b.")
    plt.xlabel("$x_1$", fontsize=18)
    plt.ylabel("$y$", rotation=0, fontsize=18)
    plt.legend(loc="upper left", fontsize=14)
    plt.axis([0, 1, 0, 0.5])
    save_fig("linear_model_predictions_plot")
    plt.show()
```

Saving figure linear model predictions plot

