```
In [179]: import pandas as pd
import numpy as np
import itertools

In [158]: def a():
          data = pd.read_csv('forest_fires.csv').values
          processed_data = np.hstack((data[:,:2], data[:, 4:]))
          np.random.shuffle(processed_data)
          X, Y = processed_data[:, :-1], processed_data[:, -1]
          return np.array(X, dtype=np.float64)[:-2], np.array(Y, dtype=np.float64)[:-2]

X, Y = a()
    print(X.shape)
    print(Y.shape)

(515, 10)
    (515,)
```

```
In [176]: def gen_k fold(x, y, k):
              x_splits = np.vsplit(x, k)
              y_splits = np.split(y, k)
              return x_splits, y_splits
          def b():
              x, y = a()
              x splits, y splits = gen_k fold(x, y, 5)
              total_error = 0
              for i in range(5):
                   x_train = np.vstack([x for j, x in enumerate(x_splits) if j !=
          i])
                   x_test = np.vstack(x_splits[i])
                   y_train = []
                   for j in range(len(y_splits)):
                       if j != i:
                           y_train.extend(y_splits[j])
                   y_train = np.array(y_train)
                   y_test = np.array(y_splits[i])
                   print(x_train.shape)
                   print(x_test.shape)
                   print(y_train.shape)
                   print(y_test.shape)
                   coeff = np.linalg.solve(x_train.T @ x_train, x_train.T @
          y_train)
                   pred = x_test @ coeff
                   error = (1 / pred.shape[0]) * np.linalg.norm(pred - y_test)
                   total_error += error
              return total error / 5
          b()
          (412, 10)
          (103, 10)
          (412,)
          (103,)
          (412, 10)
          (103, 10)
          (412,)
          (103,)
          (412, 10)
          (103, 10)
          (412,)
          (103,)
          (412, 10)
          (103, 10)
          (412,)
          (103,)
          (412, 10)
          (103, 10)
          (412,)
          (103,)
```

Out[176]: 5.3928862887794144

```
In [177]: def c():
               x, y = a()
               x = np.array([row[-4:] for row in x])
               x splits, y splits = gen k fold(x, y, 5)
               total_error = 0
               for i in range(5):
                   x_train = np.vstack([x for j, x in enumerate(x_splits) if j !=
          i])
                   x_test = np.vstack(x_splits[i])
                   y_train = []
                   for j in range(len(y_splits)):
                       if j != i:
                           y_train.extend(y_splits[j])
                   y_train = np.array(y_train)
                   y_test = np.array(y_splits[i])
                   print(x_train.shape)
                   print(x_test.shape)
                   print(y_train.shape)
                   print(y_test.shape)
                   coeff = np.linalg.solve(x_train.T @ x_train, x_train.T @
          y_train)
                   pred = x_test @ coeff
                   error = (1 / pred.shape[0]) * np.linalg.norm(pred - y_test)
                   total_error += error
               return total_error / 5
          C()
           (412, 4)
           (103, 4)
          (412,)
          (103,)
          (412, 4)
          (103, 4)
          (412,)
          (103,)
          (412, 4)
          (103, 4)
          (412,)
          (103,)
          (412, 4)
          (103, 4)
          (412,)
          (103,)
          (412, 4)
          (103, 4)
           (412,)
          (103,)
Out[177]: 4.7619997547141155
```

```
In [184]: | partition = int(0.8 * X.shape[0])
          train_x = X[:partition]
          train y = Y[:partition]
          test_x = X[partition:]
          test_y = Y[partition:]
          def d():
              lambda values, degree values = [0.01, .1, 1, 10, 100], list(range(1,
           6))
              def create_a(x, degree):
                  def create_features(features, degree):
                      result = np.array([1])
                      for pdegree in range(1, degree + 1):
                           presult = list(itertools.combinations_with_replacement(f
          eatures, pdegree))
                           result = np.append(result, np.prod(presult, axis=1))
                      return result
                  return np.array([create_features(i, degree) for i in x])
              def regress(X_train, y_train, X_test, y_test, lambda_v):
                  ridge = X train.T @ X train + lambda v * np.eye(len(X train[0]))
                  weights = np.linalg.solve(ridge, X_train.T @ y_train)
                  pred = X_test @ weights
                  return np.linalg.norm(pred - y_test) / pred.shape[0]
              min_degree, min_lambda, min_error = 9999999999, 999999999, 999999
          99999
              for degree v in degree values:
                  train X, test X = create a(train x, degree v), create a(test x,
          degree_v)
                  for lambda v in lambda values:
                      error = regress(train X, train y, test X, test y, lambda v)
                      if error < min error:</pre>
                          min_degree, min_lambda, min_error = degree_v, lambda_v,
          error
              return min degree, min lambda, min error
          d()
```

Out[184]: (1, 100, 2.7919669888039018)

```
In [186]: def e():
              const_x_train, const_x_test = train_x[:,6:], test_x[:,6:]
              min_feature_count, min_error = 9999999999, 9999999999
              for feature_count in range(1, 7):
                  add_x_train =
          np.array(list(itertools.combinations(train_x[:,:6].T, feature_count)))
                  add_x_test =
          np.array(list(itertools.combinations(test_x[:,:6].T, feature_count)))
                  for i in range(0, len(add_x_train)):
                      Train_X = np.hstack((add_x_train[i].T, const_x_train))
                      Test_X = np.hstack((add_x_test[i].T, const_x_test))
                      Train_X, Test_X = create_a(Train_X, min_degree), create_a(Te
          st_X, min_degree)
                      error = regress(Train_X, train_y, Test_X, test_y,
          min_lambda)
                      if error < min_error:</pre>
                          min_feature count, min_error = feature count, error
              return min feature count, min error
          e()
```

Out[186]: (1, 2.6173825994574962)

In []: