1 1 2 1 3 1 4 1 var def	330000 7420 4 2 3 yes no no no yes 2 yes furnished 225000 8960 4 4 4 yes no no no yes 3 no furnished 225000 9960 3 2 2 yes no yes no yes no no yes 3 yes furnished 2215000 7500 4 2 2 yes no yes no yes no yes a yes furnished 2215000 7420 4 1 2 yes yes yes no yes no yes 2 no furnished 21161 = ['mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'prefarea'] **Ising[varlist] = housing[varlist].apply(binary_map) **Ising[varlist] = housing[varlist] apply(binary_map) **Ising[varlist] = housing[varlist] apply(binary_map) **Ising[varlist] apply(binary_map)
1 1 2 1 3 1 4 1 fro df_	330000 7420
num df_ df_ df_ df_ df_	n_vars = ['area', 'bedrooms', 'bathrooms', 'stories', 'parking','price'] Newtrain = df_train[num_vars] Newtest = df_test[num_vars] Norm = df_Newtrain Standard = df_Newtrain Newtrain.head() area bedrooms bathrooms stories parking price
392 231 271 250 imp war	4500
df_df_df_	Norm [num_vars] = scaler.fit_transform(df_Norm [num_vars]) Norm.head(20) area bedrooms bathrooms stories parking price 0.193548 0.50 0.00 0.333333 0.000000 0.120606 0.156495 0.50 0.00 0.333333 0.000000 0.151515 0.180471 0.50 0.00 0.000000 0.000000 0.254545 0.005013 1.00 0.00 0.666667 0.000000 0.224242 0.121622 0.50 0.00 0.666667 0.000000 0.239394
461 124 154 451 59 493	0.040976 0.50 0.0 0.000000 0.001485 0.226969 0.25 0.0 0.00000 0.15152 0.340671 0.50 0.5 1.00000 0.33333 0.36363 0.131793 0.50 0.5 0.00000 0.327273 0.357018 0.25 0.0 0.00000 0.121212 0.302528 0.50 0.5 1.00000 0.33333 0.472727 0.154316 0.50 0.0 0.00000 0.09909 0.142691 0.25 0.0 0.00000 0.112121
540 406 289 190 55	0.182650 0.50 0.0 0.333333 0.093939 0.084568 0.25 0.0 0.000000 0.66667 0.00601 0.253124 0.25 0.0 0.000000 0.333333 0.148485 0.291630 0.25 0.0 0.00000 0.66667 0.212121 0.418774 0.75 0.0 0.333333 0.484848 0.612685 0.50 0.0 0.00000 0.333333 0.303030
war fro sca df_ df_ 454	rings.filterwarnings('ignore') misklearn.preprocessing import MinMaxScaler, StandardScaler cler = StandardScaler() Standard[num_vars] = scaler.fit_transform(df_Standard[num_vars]) Standard.head(20) area bedrooms bathrooms stories parking price -0.286366 0.073764 -0.581230 0.207401 -0.822960 -0.868394 -0.544762 0.073764 -0.581230 -0.937813 -0.822960 -0.041744
271250541461124	-1.6011452.884176-0.5812300.207401-0.822960-0.228768-0.7879580.073764-0.5812301.352614-0.822960-0.135256-1.3503490.073764-0.581230-0.937813-0.822960-1.603589-0.053303-1.331442-0.581230-0.937813-0.822960-0.902058
289 190 55	-0.641027-1.331442-0.581230-0.937813-0.822960-0.920761-0.3623650.073764-0.5812300.2074010.321375-1.032976-1.046354-1.331442-0.581230-0.9378131.465710-1.5753480.129094-1.331442-0.581230-0.9378130.321375-0.6963310.397623-1.331442-0.581230-0.9378131.465710-0.3035781.2842761.478970-0.5812300.2074011.4657100.145281
X_r y_r x_t y_t x_s y_s	<pre>in = df_Norm.values[:,[0,1,2,3,4]] in = df_Norm.values[:,5] in = df_Newtest.values[:,[0,1,2,3,4]] in = df_Newtest.values[:,5] in = df_Newtest.values[:,5] in = df_Newtest.values[:,5] in = np.ones(X_n.shape[1]) in = np.ones(X_n.shape[1]) in range(0, X_n.shape[1]): in range(0, X_n.shape[1</pre>
sto	<pre>mean[i] = np.mean(X_n.transpose()[i]) std[i] = np.std(X_n.transpose()[i]) for j in range(0, X_n.shape[0]):</pre>
sto for	<pre>in = np.ones(X_t.shape[1]) i = np.ones(X_t.shape[1]) i in range(0, X_t.shape[1]): mean[i] = np.mean(X_t.transpose()[i]) std[i] = np.std(X_t.transpose()[i]) for j in range(0, X_t.shape[0]): X_t[j][i] = (X_t[j][i] - mean[i])/std[i] **compute_cost(X, n, theta): h = np.ones((X.shape[0],1)) thata = thata rankape(0, 21)</pre>
def	<pre>theta = theta.reshape(1,n+1) for i in range(0,X.shape[0]): h[i] = float(np.matmul(theta, X[i])) h = h.reshape(X.shape[0]) return h gradient_descent(X, y, theta, alpha, iterations, n, h): cost = np.ones(iterations) for i in range(0,iterations): theta[0] = theta[0] - (alpha/X.shape[0]) * sum(h - y) for j in range(1,n+1): theta[i] = theta[j] - (alpha/X.shape[0]) * sum((h-y) * X.transpose()[j]) h = compute_cost(X, n, theta)</pre>
def	<pre>h = compute_cost(X, n, theta) cost[i] = (1/X.shape[0]) * 0.5 * sum(np.square(h - y)) theta = theta.reshape(1,n+1) return theta, cost f linear_regression(X, y, alpha, iterations): n = X.shape[1] one_column = np.ones((X.shape[0],1)) X = np.concatenate((one_column, X), axis = 1) theta = np.zeros(n+1) h = compute_cost(X, n, theta) theta, cost = gradient_descent(X, y, theta, alpha, iterations, n, h) return theta, cost</pre>
the pri cos n_i Fina	rations = 500; reta, cost = linear_regression(X_n, y_n, 0.1, iterations)
pri cos n_i Fina 2 the pri cos n_i	ta2, cost2 = linear_regression(X_s, y_s, 0.1, iterations) int('Final value of theta with standardization =', theta2) it2 = list(cost2) iterations2 = [x for x in range(1,(iterations + 1))] al value of theta with standardization = [[1.24535844e-16 3.83653304e-01 1.04343457e-01 2.98541735e-01 34542828e-01 1.49757135e-01]] ita_t, cost_t = linear_regression(X_t, y_t, 0.1, iterations) int('Final value of theta of the test set =', theta_t) it_t = list(cost_t) iterations_t = [x for x in range(1,(iterations + 1))]
plt plt plt plt plt plt	al value of theta of the test set = [[4009323.46427773 844638.61768703 225437.77741561 911745.77297157 s.plot(n_iterations, cost, label='Training normalization') s.legend() s.rcParams["figure.figsize"]=(10,6) s.grid() s.xlabel('Iterations') s.ylabel('Convergence of gradient descent') c(0.5, 1.0, 'Convergence of gradient descent') Convergence of gradient descent
0.1 0.1 0.1	Convergence of gradient descent Training normalization 375 350 350 360
o.:	275 250 250 260 275 275 275 275 275 275 275 275 275 276 277 277 277 277 277 277 277 277 277
plt plt plt plt	.:grid() .:xlabel('Iterations') .:ylabel('Cost') .:title('Convergence of gradient descent') Convergence of gradient descent Convergence of gradient descent Training standardization
0.1 to 0.1 0.1	350
plt plt plt plt plt plt	<pre>c.plot(n_iterations_t, cost_t, label='Test set') c.legend() c.rcParams["figure.figsize"]=(10,6) c.sgrid() c.xlabel('Iterations') c.ylabel('Cost') c.title('Convergence of gradient descent')</pre> c(0.5, 1.0, 'Convergence of gradient descent')
0.1 0.1 0.0	Convergence of gradient descent Test set
df_ df_ df_ 454 392 231 271	Newtest = df_test[num_vars] Norm = df_Newtrain Standard = df_Newtrain Newtrain.head() area bedrooms bathrooms mainroad guestroom basement hotwaterheating airconditioning parking prefarea price
imp war fro sca df_ df_	port warnings rnings.filterwarnings('ignore') rm sklearn.preprocessing import MinMaxScaler, StandardScaler tler = MinMaxScaler() Norm[num_vars] = scaler.fit_transform(df_Norm[num_vars]) Norm.head(20) area bedrooms bathrooms mainroad guestroom basement hotwaterheating airconditioning parking prefarea price 0.193548 0.50 0.0 1.0 0.0 0.0 0.0 0.0 1.0 0.000000 0.0 0.
231 271 250 541 461 124 154	0.156495 0.50 0.0 1.0 0.0 0.0 0.0 0.000000 0.0 0.151515 0.180471 0.50 0.0 1.0 0.0
493 465 490 540 406 289 190	0.302528 0.50 0.5 1.0 1.0 0.0 0.0 1.0 0.3333333 0.0 0.472727 0.154316 0.50 0.0 1.0 0.0
imp war fro sca df_ df_	0.612685
392 231 271 250 541 461 124	-0.544762 0.073764 -0.581230 0.393123 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 -0.564215 -0.677628 -0.377564 0.073764 -0.581230 0.393123 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 1.772373 -0.041744 -1.601145 2.884176 -0.581230 -2.543735 -0.457738 1.405903 -0.216109 -0.702935 -0.822960 -0.564215 -0.228768 -0.787958 0.073764 -0.581230 0.393123 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 -0.564215 -0.135256 -1.350349 0.073764 -0.581230 -2.543735 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 -0.564215 -1.603589 -0.053303 -1.331442 -0.581230 0.393123 -0.457738 1.405903 -0.216109 -0.702935 -0.822960 -0.564215 -1.603589 -0.053303 -0.2581230 0.0457738 1.405903 -0.216109 -0.7
465	0.473622 0.073764 1.488383 0.393123 2.184657 -0.216109 1.422607 0.321375 -0.564215 1.304836 -0.559962 0.073764 -0.581230 0.393123 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 -0.564215 -1.051678 -0.362365 0.073764 -0.581230 0.393123 -0.457738 -0.711287 -0.216109 -0.702935 -0.822960 -0.564215 -0.920761 -0.362365 0.073764 -0.581230 -2.543735 -0.457738 -0.711287 4.627285 -0.702935 0.321375 -0.564215 -1.032976 -1.046354 -1.331442 -0.581230 0.393123 -0.457738 1.405903 -0.216109 -0.702935 1.465710 -0.564215 -1.032976 -1.046354 -1.331442 -0.581230 0.393123 -0.457738 1.405903 -0.216109 -0.702935 1.465710 -0.564215 -1.032976 -1.046354 -1.046354 -0.581230 0.393123 -0.457738 1.405903 -0.216109
55 171 X_r y_r X_t y_t X_s	
mea sto	<pre>in = np.ones(X_n.shape[1]) i = np.ones(X_n.shape[1]) i in range(0, X_n.shape[1]): mean[i] = np.mean(X_n.transpose()[i]) std[i] = np.std(X_n.transpose()[i]) for j in range(0, X_n.shape[0]):</pre>
sto for	<pre>for j in range(0, X_s.shape[0]): X_s[j][i] = (X_s[j][i] - mean[i])/std[i] in = np.ones(X_t.shape[1]) i = np.ones(X_t.shape[1]): iin range(0, X_t.shape[1]): mean[i] = np.mean(X_t.transpose()[i]) std[i] = np.std(X_t.transpose()[i]) for j in range(0, X_t.shape[0]):</pre>
pri cos n_i Fina 1 2 the pri cos n_i	ta, cost = linear_regression(X_n, y_n, 0.1, iterations) int('Final value of theta with normalization =', theta) it = list(cost) iterations = [x for x in range(1,(iterations + 1))] al value of theta with normalization = [[1.24885520e-16 2.64108625e-01 1.27168543e-01 2.88652311e-01 int('Interations') = [x for x in range(1, (iterations))] al value of theta with normalization = [[1.24885520e-16 2.64108625e-01 1.27168543e-01 2.88652311e-01 int('Interations') = [x for x in range(1, (iterations))] al value of theta with normalization = [1.24885520e-16 2.64108625e-01 1.27168543e-01 2.88652311e-01 int('Interations') = [x for x in range(1, (iterations))] al value of theta with normalization = [1.24885520e-16 2.64108625e-01 1.27168543e-01 2.88652311e-01 int('Interations') = [x for x in range(1, (iterations))]
the pri cos n_i Fina	al value of theta with standardization = [[1.24885520e-16 2.64108625e-01 1.27168543e-01 2.88652311e-01 2.21315783e-01 9.99334571e-02 3.90689188e-02 1.43965719e-01 7.0513007e-01 9.08846334e-02 1.67876986e-01]] eta_t, cost_t = linear_regression(X_t, y_t, 0.1, iterations)
plt plt plt plt plt plt	<pre>c.legend() c.rcParams["figure.figsize"]=(10,6) c.grid() c.xlabel('Iterations') c.ylabel('Cost') c.title('Convergence of gradient descent') Convergence of gradient descent Convergence of gradient descent Taing set normalization Traing set normalization Traing set normalization Traing set normalization</pre>
0.1 0.1 0.1 0.1	350 325 330 330 325 325 325 325 325 325 325 325 325 325
plt plt plt plt plt	<pre>c.plot(n_iterations2, cost2, label='Traing set standardization') c.legend() c.rcParams["figure.figsize"]=(10,6) c.grid() c.xlabel('Iterations') c.ylabel('Cost') c.title('Convergence of gradient descent')</pre>
0.0 0.0 0.0	Convergence of gradient descent Convergence of gradient descent Taing set standardization 335 330
0.1 0.1 0.1	275 250 225 200 200 300 400 500 lterations
plt plt plt plt plt plt	<pre>plot(n_iterations_t, cost_t, label='Test set')legend()rcParams["figure.figsize"]=(10,6)grid()xlabel('Iterations')ylabel('Cost')title('Convergence of gradient descent') c(0.5, 1.0, 'Convergence of gradient descent') lel2</pre>
8 6 4	
2	0 100 200 300 400 500 lterations