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Task 4

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```
In [13]:
          import pandas as pd
          import numpy as np
          from itertools import repeat
 In [3]: features = pd.read_csv("features.txt", sep=' ', names=['id', 'cols'])
          subject_train = np.loadtxt("subject_train.txt")
          subject_test = np.loadtxt("subject_test.txt")
          x_train = np.loadtxt("X_train.txt")
          y_train = np.loadtxt("Y_train.txt")
          x_test = np.loadtxt("X_test.txt")
          y_test = np.loadtxt("Y_test.txt")
          1. Merges the training and the test sets to create one data set.
          x_train = np.append(x_train, subject_train.reshape(-1,1), axis=1)
 In [4]:
          x_train = np.append(x_train, y_train.reshape(-1,1), axis=1)
          x_test = np.append(x_test, subject_test.reshape(-1,1), axis=1)
          x_test = np.append(x_test, y_test.reshape(-1,1), axis=1)
          X = pd.DataFrame(np.append(x_train, x_test, axis=0), columns=features.cols.values.
                            tBodyAcc- tBodyAcc- tBodyAcc- tBodyAcc- tBodyAcc- tBodyAcc-
 Out[4]:
                  tBodyAcc-
                             mean()-Y
                                        mean()-Z
                                                      std()-X
                                                                 std()-Y
                                                                                     mad()-X
                   mean()-X
                                                                            std()-Z
                                                                                                mad()-
                   0.288585
                             -0.020294
                                        -0.132905
                                                   -0.995279
                                                              -0.983111
                                                                         -0.913526
               0
                                                                                    -0.995112
                                                                                               -0.98318
                   0.278419
                             -0.016411
                                         -0.123520
                                                   -0.998245
                                                              -0.975300
                                                                         -0.960322
                                                                                    -0.998807
                                                                                               -0.97491
              2
                   0.279653
                             -0.019467
                                        -0.113462
                                                   -0.995380
                                                              -0.967187
                                                                         -0.978944
                                                                                    -0.996520
                                                                                               -0.96366
              3
                   0.279174
                             -0.026201
                                                               -0.983403
                                                                                               -0.98275
                                         -0.123283
                                                   -0.996091
                                                                         -0.990675
                                                                                    -0.997099
               4
                   0.276629
                             -0.016570
                                        -0.115362
                                                   -0.998139
                                                               -0.980817
                                                                         -0.990482
                                                                                     -0.998321
                                                                                                -0.97967
          10294
                   0.310155
                             -0.053391
                                        -0.099109
                                                   -0.287866
                                                              -0.140589
                                                                         -0.215088
                                                                                    -0.356083
                                                                                               -0.14877
          10295
                   0.363385
                             -0.039214
                                         -0.105915
                                                   -0.305388
                                                               0.028148
                                                                         -0.196373
                                                                                               -0.03003
                                                                                    -0.373540
          10296
                   0.349966
                              0.030077
                                        -0.115788
                                                   -0.329638
                                                              -0.042143
                                                                         -0.250181
                                                                                    -0.388017
                                                                                               -0.13325
          10297
                   0.237594
                              0.018467
                                         -0.096499
                                                   -0.323114
                                                              -0.229775
                                                                         -0.207574
                                                                                    -0.392380
                                                                                               -0.27961
          10298
                   0.153627
                             -0.018437
                                        -0.137018
                                                   -0.330046
                                                                                    -0.430974
                                                              -0.195253
                                                                         -0.164339
                                                                                               -0.21829
```

2. Extracts only the measurements on the mean and standard deviation for each measurement.

10299 rows × 563 columns

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-0.019467

-0.026201

-0.016570

In [8]:		<pre>X_mean_std = X[features[features.cols.str.contains('mean') features.cols.str.col X_mean_std.head()</pre>									
Out[8]:		tBodyAcc- mean()-X	•	tBodyAcc- mean()-Z	tBodyAcc- std()-X	tBodyAcc- std()-Y	tBodyAcc- std()-Z	tGravityAcc- mean()-X	tGravityAcc mean()-\		
	0	0.288585	-0.020294	-0.132905	-0.995279	-0.983111	-0.913526	0.963396	-0.140840		
	1	0.278419	-0.016411	-0.123520	-0.998245	-0.975300	-0.960322	0.966561	-0.14155		

-0.995380

-0.996091

-0.998139

-0.967187

-0.983403

-0.980817

-0.978944

-0.990675

-0.990482

0.966878

0.967615

0.968224

-0.142010

-0.143970

-0.148750

•

5 rows × 79 columns

0.279653

0.279174

0.276629

3. Uses descriptive activity names to name the activities in the data set

-0.113462

-0.123283

-0.115362

In [10]: activities = pd.read_csv("activity_labels.txt",sep=' ', names=['activity', 'activity', 'activity']

Out[10]:		activity	activity_type
	0	1	WALKING
	1	2	WALKING_UPSTAIRS
	2	3	WALKING_DOWNSTAIRS
	3	4	SITTING
	4	5	STANDING
	5	6	LAYING

```
In [12]: X = X.merge(activities, how='inner', on='activity')
X
```

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ityMean)	angle (t Body Gyro Mean, gravity Mean)	angle (t Body Gyro Jerk Mean, gravity Mean)	angle(X,gravityMean)	an
0.030400	-0.464761	-0.018446	-0.841247	
-0.007435	-0.732626	0.703511	-0.844788	
0.177899	0.100699	0.808529	-0.848933	
-0.012892	0.640011	-0.485366	-0.848649	
0.122542	0.693578	-0.615971	-0.847865	
0.346295	0.884904	-0.698885	-0.651732	
-0.372889	-0.657421	0.322549	-0.655181	
0.088574	0.696663	0.363139	-0.655357	
-0.819188	0.929294	-0.008398	-0.659719	
-0.287951	0.876030	-0.024965	-0.660080	

5. From the data set in step 4, creates a second, independent tidy data set with the average of each variable for each activity and each subject.

14]:	X.group	oby(['ac	tivity','s	subject']).	.agg(dict(zip(features	[features.co	ols.str.cont	ains
•			tBodyAcc- mean()-X	tBodyAcc- mean()-Y	tBodyAcc- mean()-Z	tGravityAcc- mean()-X	tGravityAcc- mean()-Y	tGravityAcc- mean()-Z	tBod
	activity	subject							
	1.0	1.0	0.277331	-0.017384	-0.111148	0.935223	-0.282165	-0.068103	
		2.0	0.276427	-0.018595	-0.105500	0.913017	-0.346607	0.084727	
		3.0	0.275567	-0.017177	-0.112675	0.936507	-0.261986	-0.138108	
		4.0	0.278582	-0.014840	-0.111403	0.964000	-0.085854	0.127764	
		5.0	0.277842	-0.017285	-0.107742	0.972625	-0.100440	0.002476	
	•••	•••							
	6.0	26.0	0.271646	-0.019190	-0.105003	-0.621310	0.803015	0.614523	
		27.0	0.274102	-0.017987	-0.107700	-0.530435	0.567879	0.845392	
		28.0	0.275913	-0.016754	-0.108345	-0.490335	0.144073	0.015757	
		29.0	0.287295	-0.017197	-0.109462	-0.346790	0.807535	0.590452	
		30.0	0.281034	-0.019449	-0.103658	-0.344738	0.732661	0.681459	
	180 rows	s × 46 co	lumns						