1.

Analyze the most common words in the clusters. Use TF-IDF to

remove irrelevant words such as "the". (1%)

The top 10 most common words in each clusters are:

(after removing stop words)

Cluster 0:

spring mvc security bean using framework use configuration web application Cluster 1:

magento product custom products page add category order checkout module Cluster 2:

scala qt apache svn spring oracle linq ajax haskell hibernate

Cluster 3:

qt mac os x file sharepoint spring ajax application oracle

Cluster 4:

svn apache bash files excel visual repository sharepoint script studio Cluster 5:

apache ajax svn wordpress magento spring hibernate matlab haskell oracle Cluster 6:

wordpress page post posts plugin category custom blog theme add Cluster 7:

haskell type function list 's error does problem types data

Cluster 8:

bash script command files shell variable string line output function Cluster 9:

drupal node custom module views content form page view menu Cluster 10:

excel vba data cell macro sheet files range formula function

Cluster 11:

oracle sql query table database use 's stored server 10g

Cluster 12:

scala java type 's does use list function class actors

Cluster 13:

ling sql query multiple group use list xml join data

Cluster 14:

hibernate mapping query criteria table using object problem join use Cluster 15:

ajax jquery page asp.net request php using javascript problem post Cluster 16:

matlab function matrix array plot image using 's data text

Cluster 17:

apache rewrite mod_rewrite server redirect error use url php files Cluster 18:

sharepoint list web custom site page services document create workflow Cluster 19:

mac os x application development osx cocoa 's terminal windows Cluster 20:

qt application window windows widget does custom creator use c++ Cluster 21:

file svn subversion repository cocoa apache using best way directory

Cluster 22:

visual studio qt svn scala matlab mac excel ajax apache Cluster 23:

using cocoa list subversion files image mod_rewrite svn scala apache Cluster 24:

svn files repository subversion copy server directory update revision working Cluster 25:

visual studio project files projects solution 's code does build Cluster 26:

drupal sharepoint magento wordpress custom qt page ajax excel list Cluster 27:

haskell scala matlab bash excel oracle linq function type hibernate Cluster 28:

file subversion cocoa text line best way xml data reading Cluster 29:

hibernate spring scala using use ajax query mapping linq does Cluster 30:

ling using sql query list multiple group xml join cocoa Cluster 31:

file oracle sharepoint scala spring qt mac excel drupal svn Cluster 32:

bash file script matlab subversion using command mac line cocoa Cluster 33:

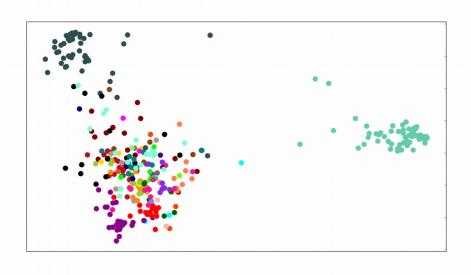
ling oracle hibernate sql excel query using drupal sharepoint scala Cluster 34:

file using mac subversion qt os cocoa x scala application Cluster 35:

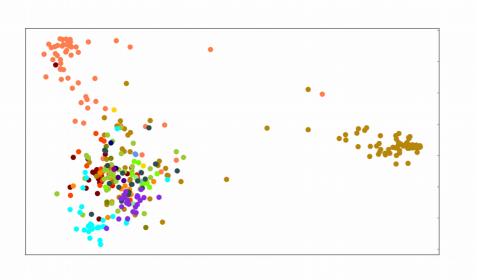
excel file matlab oracle linq data qt haskell ajax vba

2. Visualize the data by projecting onto 2-D space. Plot the results and color the data points using your cluster predictions. Comment on your plot. Now plot the results and color the data points using the true labels. Comment on this plot. (1%)

cluster prediction by cluster.py



true label



3. Compare different feature extraction methods. (2%)

method	procedure	F-measure score
method1	1.remove stopwords, stemming, and tokenizing 2.tfidf: fit_transform(titles) 3.k means	0.27558
method2	1.remove stopwords, stemming, and tokenizing 2.tfidf: fit(titles+docs), transform(titles) 3.LSA 4.k means	0.78782
method3	1.remove stopwords, stemming, and tokenizing 2.tfidf: fit(titles+docs), transform(titles) 3.PCA 4.k means	0.70987
method4	1.remove stopwords, stemming, and tokenizing 2.bag of words : fit(titles+docs),transform(titles) 3.PCA 4.k means	0.53619
method5	1.remove stopwords, stemming, and tokenizing 2.bag of words : fit(titles+docs) , transform(titles) 3.LSA 4.k means	0.65517

In general, LSA is performs better than PCA and tfidf performs better than bag of words.

4. Try different cluster numbers and compare them. You can compare the scores and also visualize the data. (1%)

number of clusters in cluster.py	F-measure score
16	0.45789
21	0.59269
26	0.72786
31	0.76308
36	0.77954