Class 1, PCA

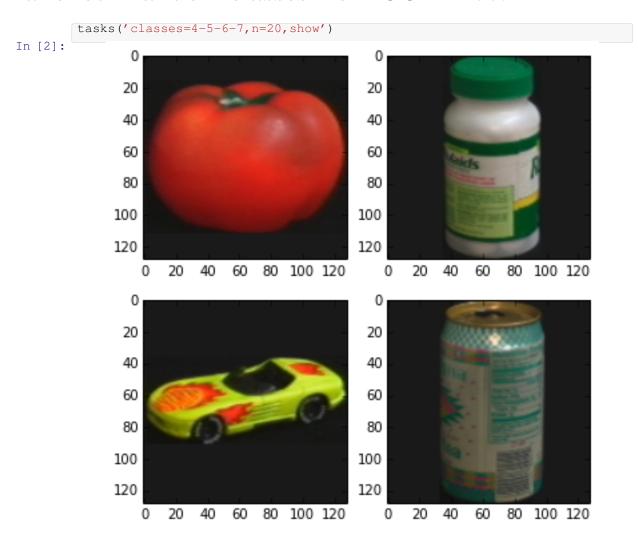
Unknown Author

November 16, 2015

%matplotlib inline
In [1]: from tasks import tasks

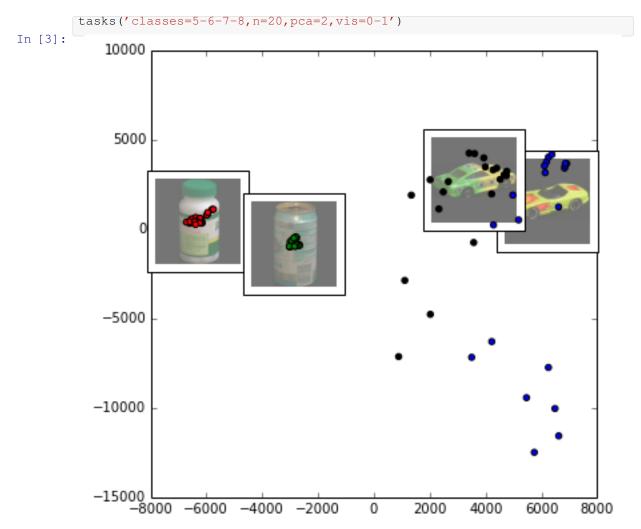
Part I

Take four random classes from COIL-100.



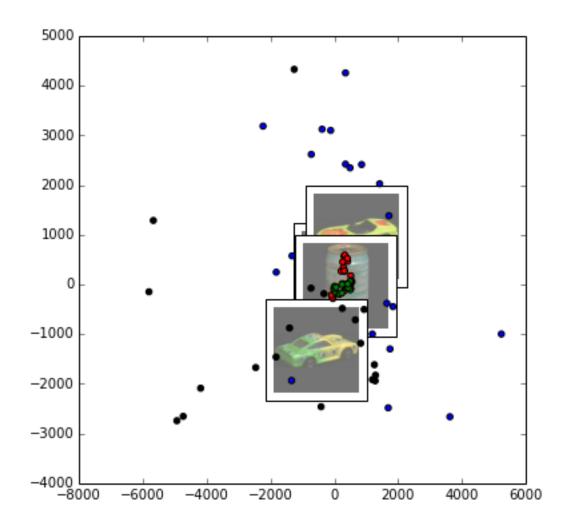
Part II

Plot first two principal components.



Part III Plot next two principal components.

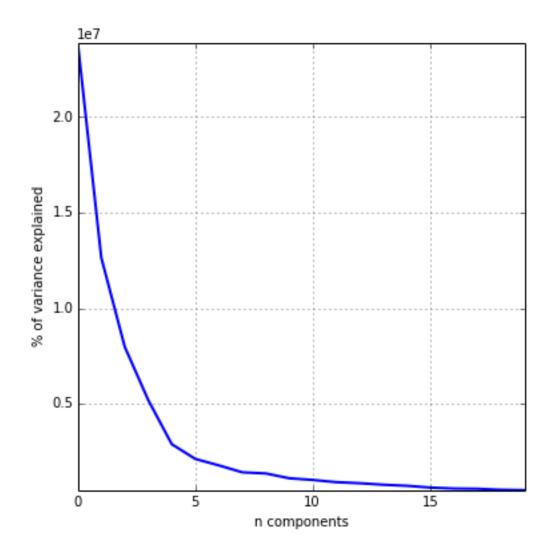
```
tasks('classes=5-6-7-8, n=20, pca=2, vis=4-5')
In [4]:
```



Part IV

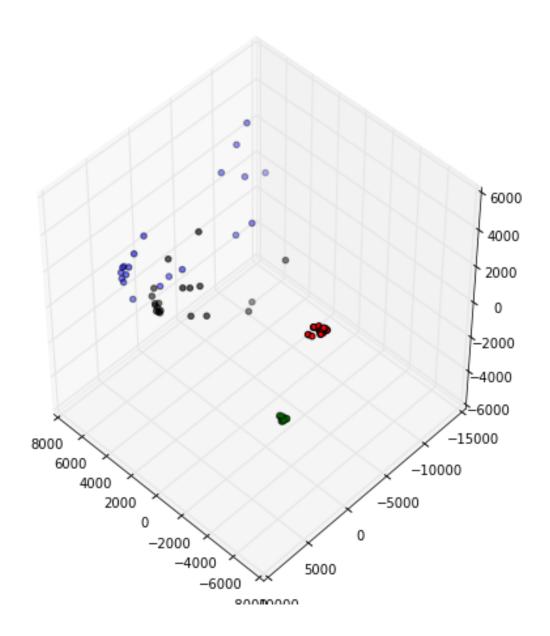
Show how much variance is explained by each principal component.

tasks('classes=5-6-7-8, n=20, pca=20, eig')
In [5]:



Part V How about three principal components?

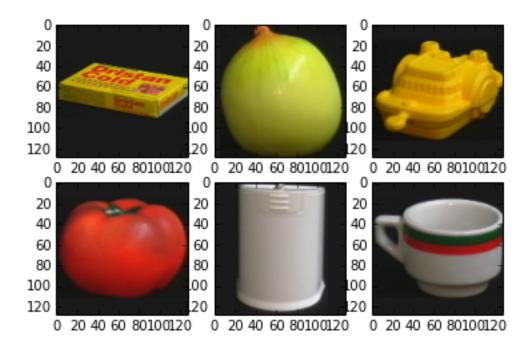
```
tasks('classes=5-6-7-8, n=20, pca=3, vis3d=0-1-2')
In [6]:
```



Part VI
Now let's try more objects.

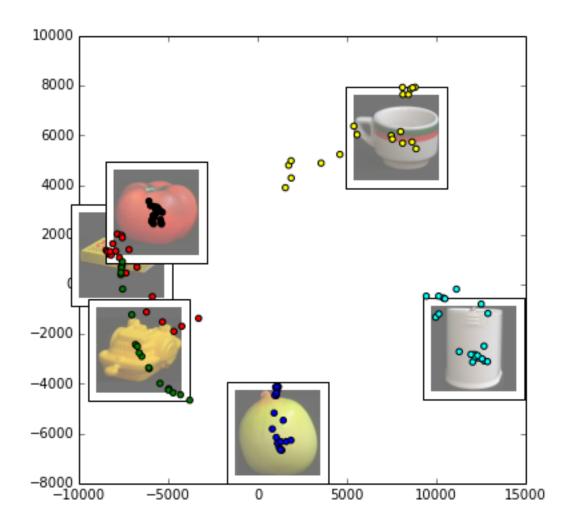
In [7]:

tasks('classes=1-2-3-4-9-10, n=20, show')



Part VII **Again, first two components.**

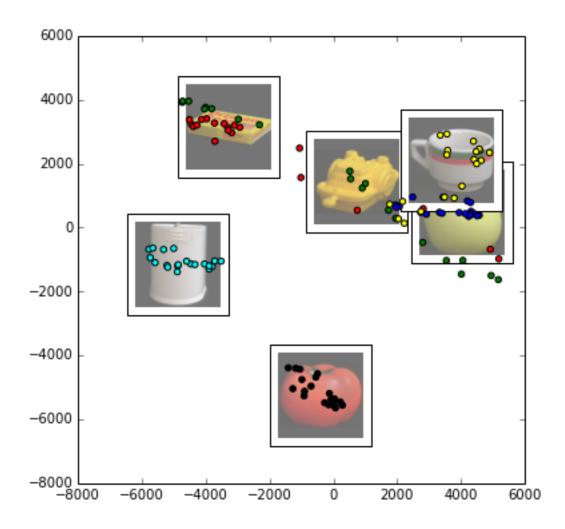
tasks('classes=1-2-3-4-9-10, n=20, pca=2, vis=0-1')
In [8]:



Part VIII

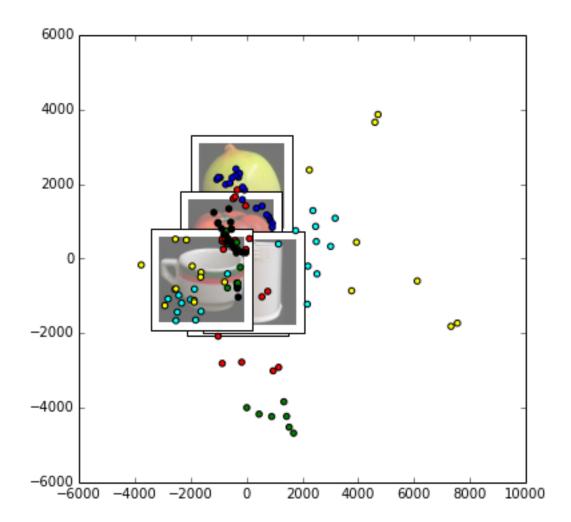
Next two components.

```
tasks('classes=1-2-3-4-9-10, n=20, pca=2, vis=2-3')
In [9]:
```



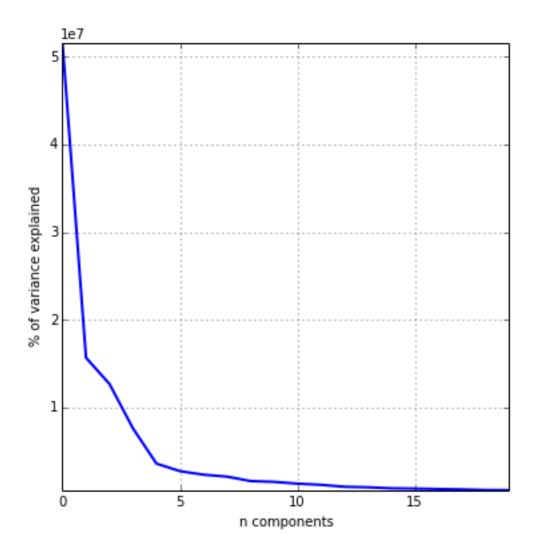
Part IX And next two...

```
tasks('classes=1-2-3-4-9-10, n=20, pca=2, vis=4-5')
In [10]:
```



Part X
What about component distribution?

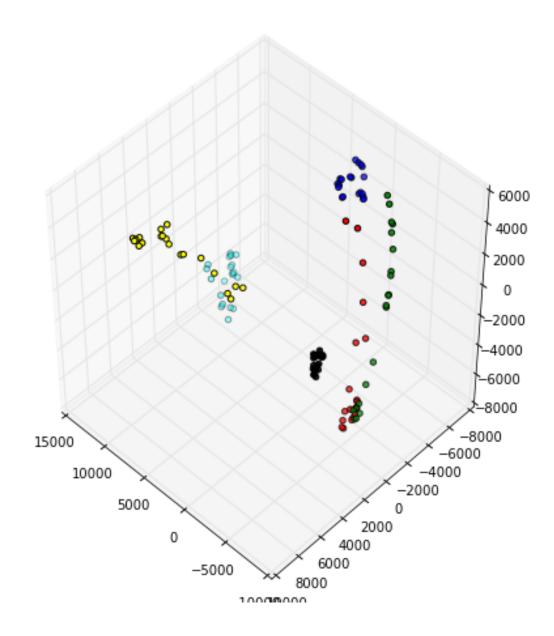
```
tasks('classes=1-2-3-4-9-10, n=20, pca=20, eig')
In [11]:
```



Part XI

First three

```
tasks('classes=1-2-3-4-9-10, n=20, pca=3, vis3d=0-1-2')
In [12]:
```



Part XII

In [13]:

Let's try to learn to distinguish between them. We'll use Naive Bayes classifier.

1 First, without PCA.

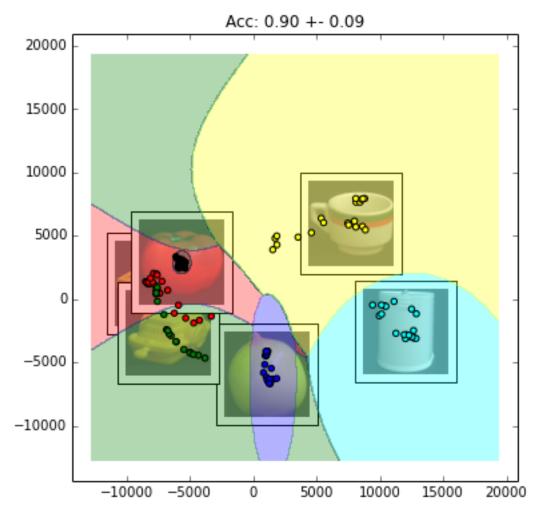
tasks('classes=1-2-3-4-9-10, n=20, nb')

2 Now with PCA with two first components.

```
tasks('classes=1-2-3-4-9-10, n=20, pca=2, nb')
In [14]: Acc: 0.90 +- 0.09
```

3 Let's visualize decision boundaries.

```
tasks('classes=1-2-3-4-9-10, n=20, vis=0-1, pca=2, nb, heat')
In [15]: Acc: 0.90 +- 0.09
```



In [15]: