

Load data

```
import cdt
eisen = cdt.ExpressionProfile("supp2data.cdt")
```

```
X = array(eisen.num)
X.shape
```

(2467, 79)

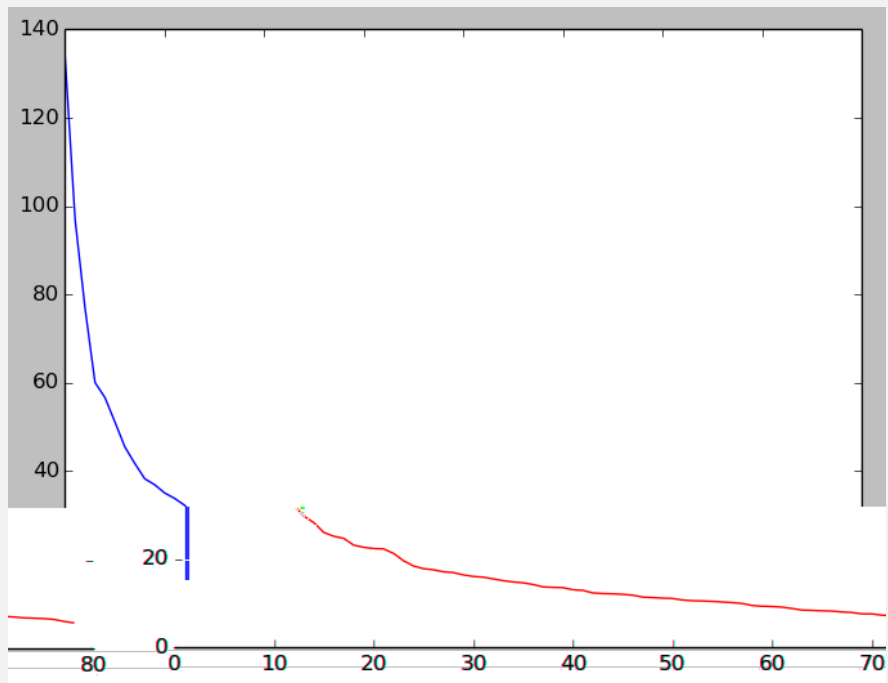
Gene oriented SVD without scaling

```
%time (u,s,v) = svd(X, full_matrices = False)
```

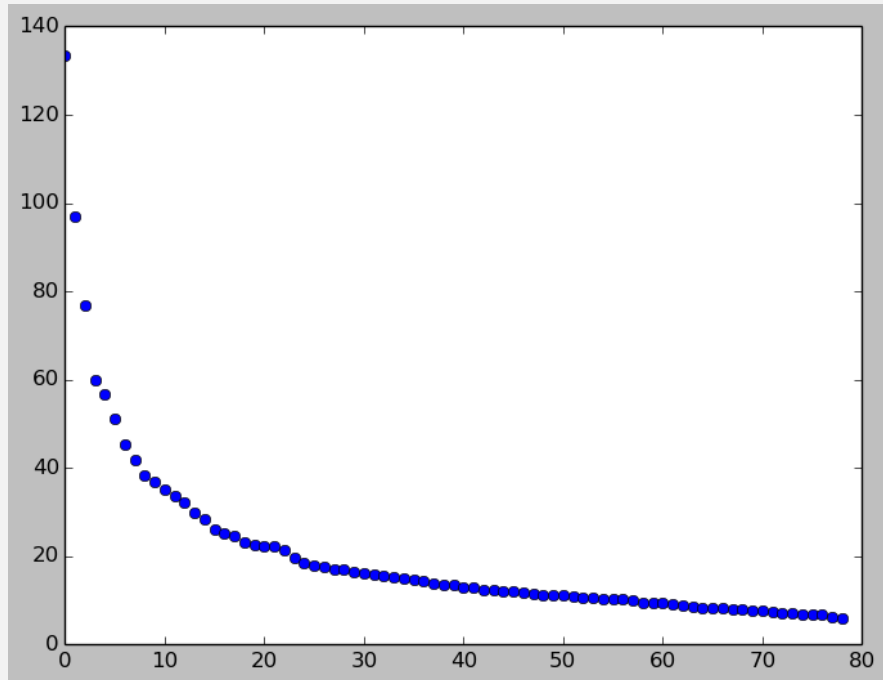
CPU times: user 44 ms, sys: 0 ns, total: 44 ms
Wall time: 44 ms

Plot singular values

```
fig = figure()
plot(s)
display(fig)
```



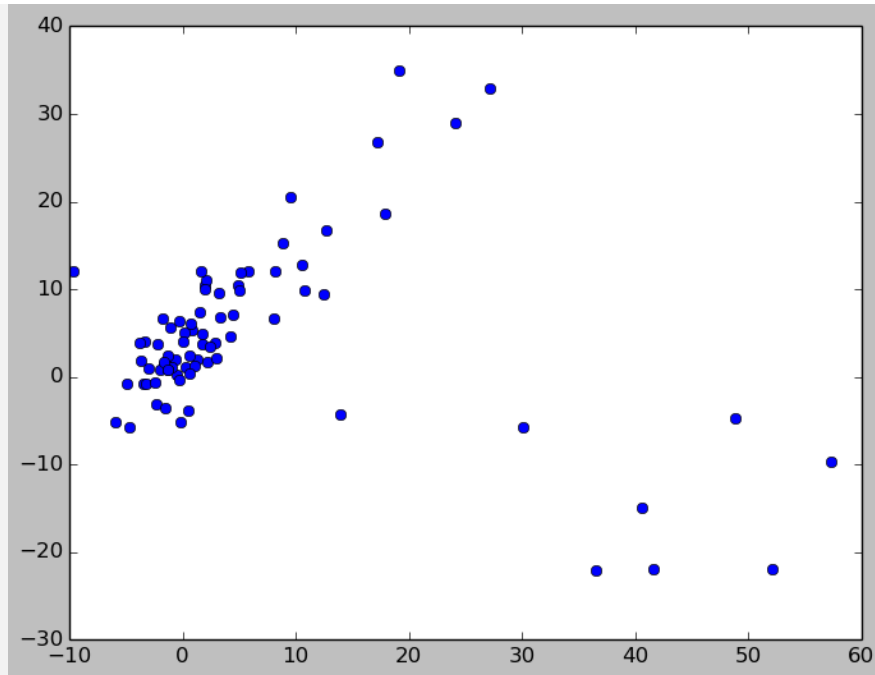
```
fig = figure()
plot(s, "bo")
display(fig)
```



Project arrays into gene space

```
p1 = dot(u[:, 0], X)  
p2 = dot(u[:, 1], X)
```

```
fig = figure()  
plot(p1, p2, "bo")  
display(fig)
```



Lets color by data set

```
print eisen.expCond
```

```
['alpha 0', 'alpha 7', 'alpha 14', 'alpha 21', 'alpha 28', 'alpha 35', 'alpha 42', 'alpha 49', 'alpha 56']
```

```
sorted(set(i.split()[0] for i in eisen.expCond))
```

```
['Elu', 'alpha', 'cdc15', 'col d', 'di au', 'dtt', 'heat', 'spo', 'spo-', 'spo5']
```

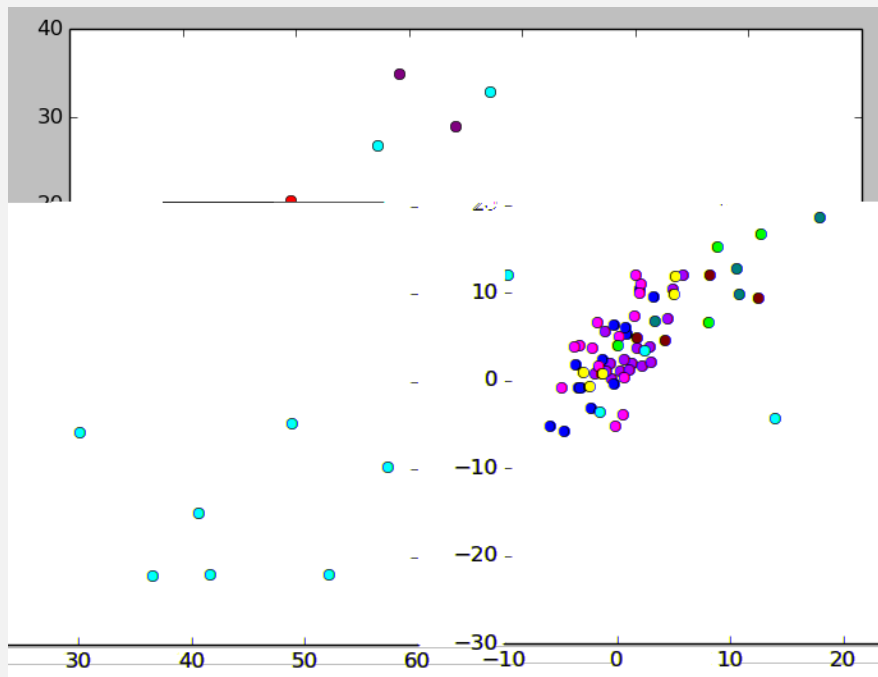
```
colors = []
for i in eisen.expCond:
    if(i.startswith("Elu")):
        colors.append("red")
    elif(i.startswith("alpha")):
        colors.append("orange")
    elif(i.startswith("cdc15")):
        colors.append("yellow")
    elif(i.startswith("col d")):
        colors.append("green")
    elif(i.startswith("di au")):
        colors.append("cyan")
    elif(i.startswith("dtt")):
        colors.append("blue")
    elif(i.startswith("heat")):
        colors.append("purple")
    elif(i.startswith("spo")):
        colors.append("magenta")
```

```
else:
    raise ValueError
```

```
len(colors), len(p1)
```

```
(79, 79)
```

```
fig = figure()
for (i, j, c) in zip(p1, p2, colors):
    plot([i], [j], color = c, marker = "o")
display(fig)
```



So, the second singular vector is separating out a lot of the sporulation data – let's color and label this series

```
ei sen. expCond. index("spo 0"), ei sen. expCond. index("heat 0")
```

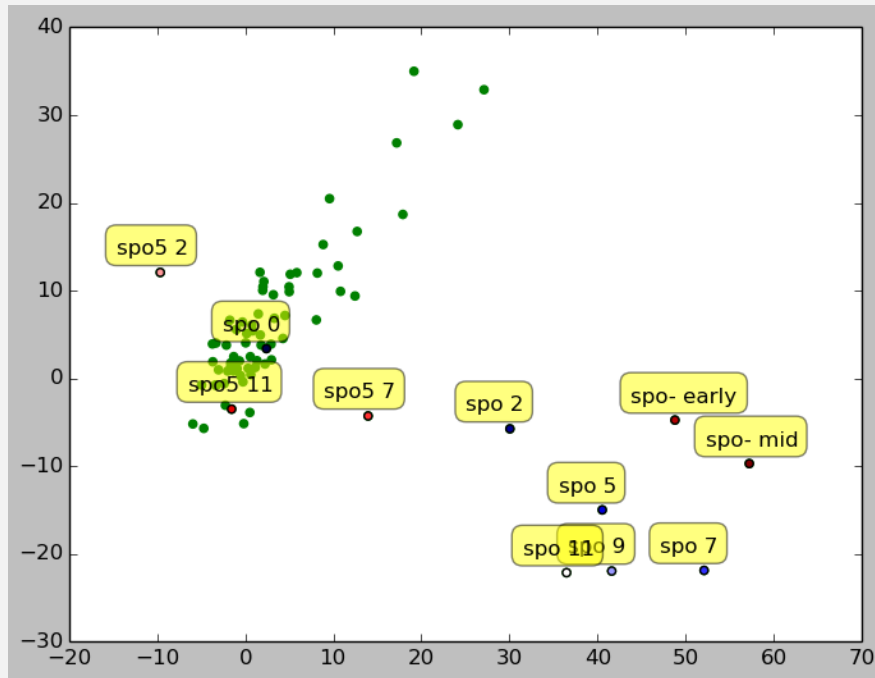
```
(47, 58)
```

```
58-47
```

```
11
```

```
(fig, ax) = subplots(1, 1)
ax.scatter(p1, p2, color = "green", marker = "o")
ax.scatter(p1[47:58], p2[47:58], c = arange(47, 58), cmap = get_cmap("seismic"))
for (x, y, label) in zip(p1[47:58], p2[47:58], ei sen. expCond[47:58]):
```

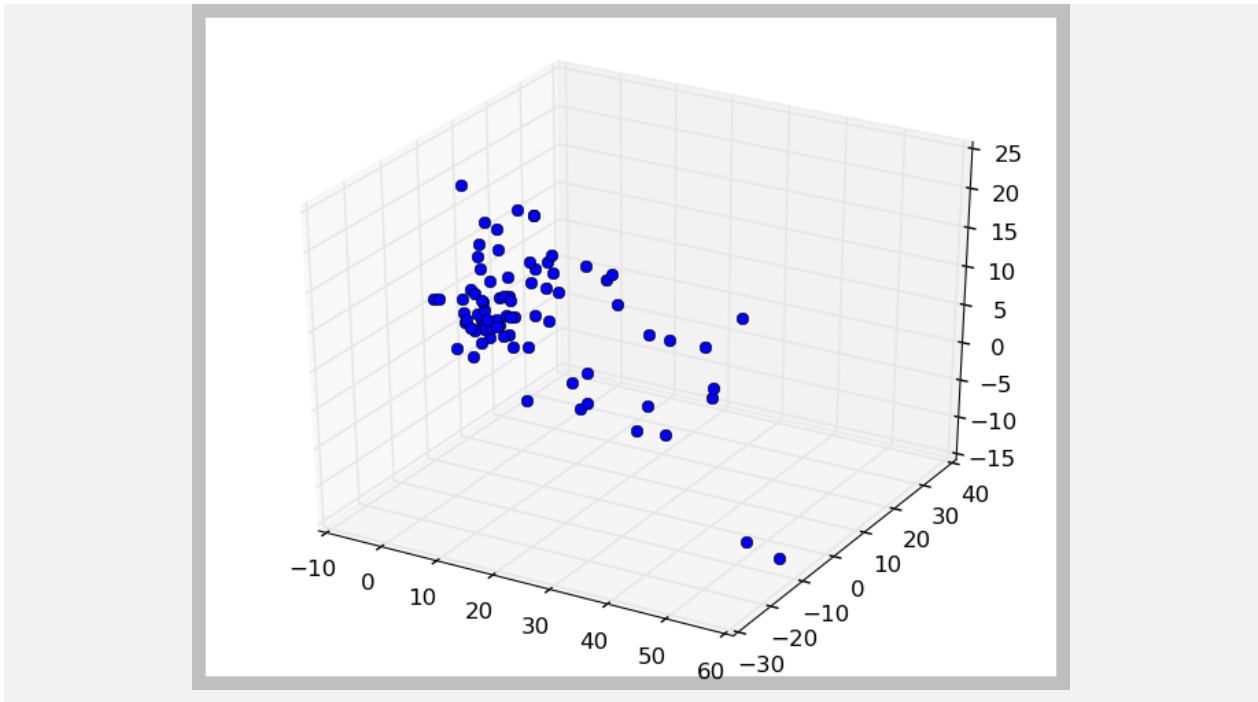
```
ax.annotate(label, xy = (x, y), xytext = (x-5, y+2),
            bbox = {"fc": 'yellow', "alpha": .5, "boxstyle": "round, pad=0.5"})
display(fig)
```



Based on the key, the “spo5” columns are relative to $t=5h$ rather than $t=0$ (effectively subtracting the spo5 vector) and the “spo-” columns are from an *ndt80* knockout relative to $t=2h$ or $t=5h$.

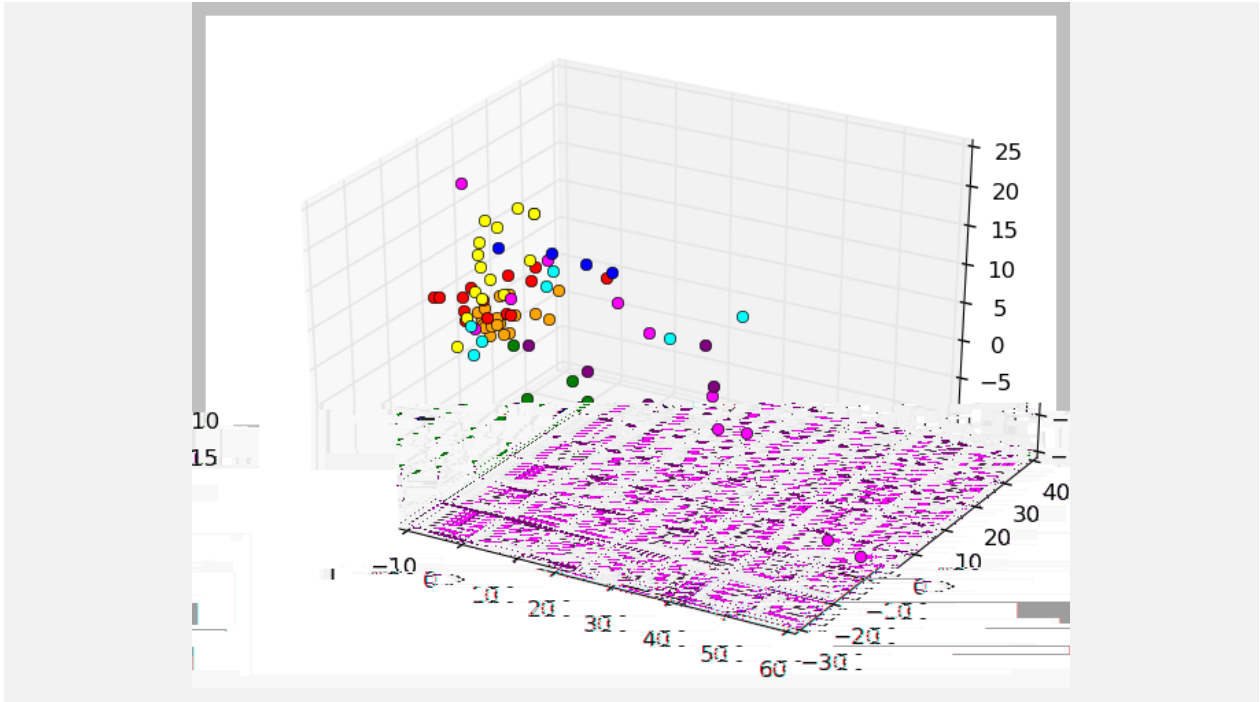
Here’s the equivalent plot for the arrays projected onto the first three components (try rotating the plot in IPython)

```
p3 = dot(u[:, 2], X)
from mpl_toolkits.mplot3d import Axes3D
fig = figure()
ax = fig.add_subplot(111, projection='3d')
ax.plot(p1, p2, p3, "bo")
display(fig)
```



Or, with our previous coloring:

```
fig = figure()
ax = fig.add_subplot(111, projection='3d')
for (i, j, k, c) in zip(p1, p2, p3, colors):
    ax.plot([i], [j], [k], color = c, marker = "o")
display(fig)
```



Another view, highlighting the separation of the heat-shock data:

`display(fig)`

