

Today:

1. Computing those Bivariate Contour Ellipses
2. More about principal components:

1. Computing those Bivariate Contour Ellipses

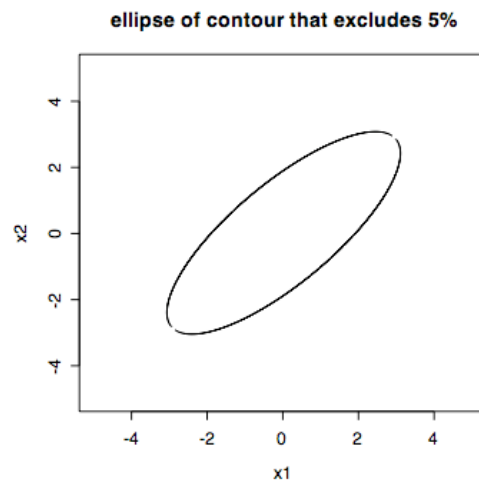
One way: Start with a (discretized circle)

Expand using eigenvalues of sample covariance matrix

Rotate according to angle of first eigenvalue

contour

```
function (x=mvrnorm(n=100,mu=c(0,0),Sigma=matrix(c(1,0.5,0.5,1),ncol=2)))
{
  eigen=eigen(cov(x))
  z1=(-1000:1000)/400 #fewer points if gap OK
  z2=(5.99-z1^2)^.5 #circle
  z1=z1*eigen$values[1] # angle is for 1st PC
  z2=z2*eigen$values[2]
  angle=pi/2+atan(eigen$vectors[,1][2]/eigen$vectors[,1][1])
  x1=mean(x[,1])+z1*cos(angle)+z2*sin(angle) # see text p 35
  x2=mean(x[,2])-z1*sin(angle)+z2*cos(angle)
  plot(x1,x2,ylim=c(mean(x[,2])-5,mean(x[,2])+5),xlim=c(mean(x[,1])-
5,mean(x[,1])+5),type="l",main="ellipse of contour that excludes 5%")
  z2=-z2 # to get rest of ellipse
  x1=mean(x[,1])+z1*cos(angle)+z2*sin(angle)
  x2=mean(x[,2])-z1*sin(angle)+z2*cos(angle)
  lines(x1,x2,type="l")
}
```



## 2. More about principal components:

Data Matrix  $x$  ( $n \times p$ )

eigenvalues of  $\text{cor}(x)$  (or  $\text{cov}(x)$ )  $\lambda_1 > \lambda_2 > \dots > \lambda_p$  (actually all estimates)

$i$ th eigenvector of  $x$ ,  $e_i$  corresponding to  $\lambda_i$  is  $p \times 1$  (column vector)

Suppose  $j$ th row of  $x$  is  $x_j'$  which is  $1 \times p$  (row vector)

Compute  $i$ th Principal Component using  $x_j' e_i$  (a scalar)

Compute vector of principal components for row  $x_j'$  by  $x_j' e$

where  $e$  is the  $p \times p$  matrix of column vectors that are eigenvectors.

$x e$  is the  $n \times p$  matrix of PCs – each row is a  $p$ -tuple of PCs describing one case.

Note:  $x_j' e_i$  and  $e_i' x_j$  are the same scalar quantity. (book uses  $e' x$ )

An example:

Weekly rates of return for 5 US stocks: Example 8.5 p 447

```
> T8.5.df=read.table("T8.4.txt")
> cor=cor(T8.5.df)
> eigen=eigen(cor)
> eigen
$values
[1] 2.8564869 0.8091185 0.5400440 0.4513468 0.3430038

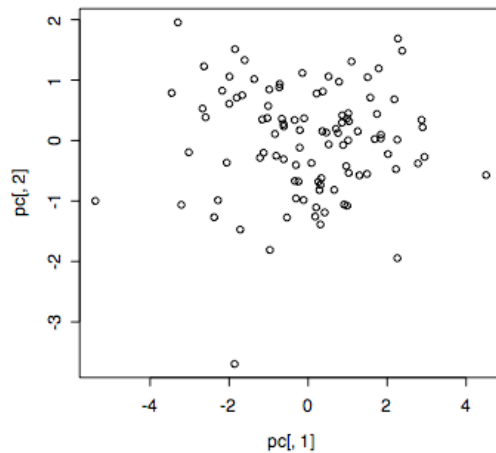
$vectors
      [,1] [,2] [,3] [,4]
[1,] -0.4635405 0.2408499 0.6133570 -0.3813727
[2,] -0.4570764 0.5090997 -0.1778996 -0.2113068
[3,] -0.4699804 0.2605774 -0.3370355 0.6640985
[4,] -0.4216770 -0.5252647 -0.5390181 -0.4728036
[5,] -0.4213291 -0.5822416 0.4336029 0.3812273
      [,5]
[1,] -0.4532876
[2,] 0.6749814
[3,] -0.3957247
[4,] -0.1794482
[5,] 0.3874672

> z=scale(T8.5.df)
```

```

> dim(z)
[1] 100 5
> 2.856/5
[1] 0.5712
> .809/5
[1] 0.1618
> (2.856+.809)/5
[1] 0.733
> mode(z)
[1] "numeric"
> pc=z%*%as.matrix(eigen$vectors)
> plot(pc[,1],pc[,2])

```



Can the dimensions be described? pc1 is like mean. pc2 is chem – oil.

Actually sequence may be of interest here. How can we label points?

sequence, size, colour?

```

week=as.character(rep(1:10,each=10))
> week
[1] "1" "1" "1" "1" "1" "1" "1" "1" "1" "1"
[11] "2" "2" "2" "2" "2" "2" "2" "2" "2" "2"
[21] "3" "3" "3" "3" "3" "3" "3" "3" "3" "3"
[31] "4" "4" "4" "4" "4" "4" "4" "4" "4" "4"
[41] "5" "5" "5" "5" "5" "5" "5" "5" "5" "5"
[51] "6" "6" "6" "6" "6" "6" "6" "6" "6" "6"
[61] "7" "7" "7" "7" "7" "7" "7" "7" "7" "7"
[71] "8" "8" "8" "8" "8" "8" "8" "8" "8" "8"
[81] "9" "9" "9" "9" "9" "9" "9" "9" "9" "9"
[91] "10" "10" "10" "10" "10" "10" "10" "10" "10" "10"

```

```
> plot(pc[,1],pc[,2],pch=week)
```

