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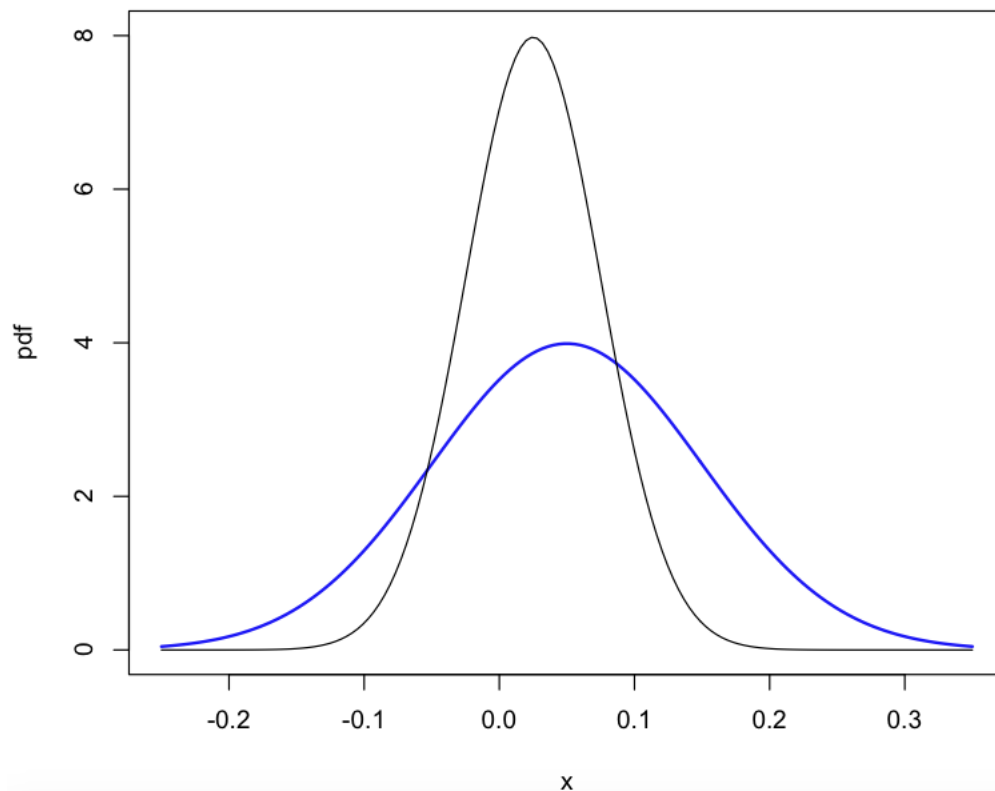
**CFRM 462: Introduction to Computational Finance and Econometrics Homework 2**

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1.
  - a)  $1 - \text{pnorm}(0.1, \text{mean} = 0.05, \text{sd} = 0.1) = 0.3085375$
  - b)  $\text{pnorm}(-0.1, \text{mean}=0.05, \text{sd} = 0.1) = 0.0668072$
  - c)  $\text{pnorm}(0.15, \text{mean}=0.05, \text{sd} = 0.1) - \text{pnorm}(-0.05, \text{mean}=0.05, \text{sd} = 0.1) = 0.6826895$
  - d)  $\text{qnorm}(.01, \text{mean}=0.05, \text{sd} = 0.1) = -0.1826348$
  - e)  $\text{qnorm}(.05, \text{mean}=0.05, \text{sd} = 0.1) = -0.1144854$
  - f)  $\text{qnorm}(.95, \text{mean}=0.05, \text{sd} = 0.1) = 0.2144854$
  - g)  $\text{qnorm}(.99, \text{mean}=0.05, \text{sd} = 0.1) = 0.2826348$
2. 

```
x.vals = seq(-0.25,0.35,length = 150)*0.1 + 0.05  
plot(x.vals,dnorm(x.vals,mean = 0.05,sd = 0.1),type = "l",lwd = 2, col = "blue", xlab =  
"x", ylab = "pdf",main = "Combined PDF",ylim = c(0,8))  
lines(x.vals,dnorm(x.vals,mean = 0.025,sd = 0.05))
```

**Combined PDF**



Microsoft has more risk than Starbucks, which means a higher potential return. However, Starbucks has a less risk which means your expected return will be lower.

```

3. > mu.R = 0.04
  > sd.R = 0.09
  >
  > w0 = 10000
  >
  > q.01.R = mu.R + sd.R *qnorm(0.01)
  > q.05.R = mu.R + sd.R *qnorm(0.05)
  >
  > var.01 = abs(q.01.R * w0)
  > var.05 = abs(q.05.R * w0)
  > var.01
[1] 1693.713
  > var.05
[1] 1080.368

```

#### 4. Continuously Compounded Monthly and Annual Var

```

e.01.R = exp(mu.R + sd.R * qnorm(0.01)) - 1
e.05.R = exp(mu.R + sd.R * qnorm(0.05)) - 1

```

```

e.Var.01 = abs(e.01.R * w0)
e.Var.05 = abs(e.05.R * w0)

```

```

> e.Var.01
[1] 1558.046
> e.Var.05
[1] 1024.055

```

b)

```

> abs(e.01.R * 12 * w0)
[1] 18696.55
> abs(e.05.R * 12 * w0)
[1] 12288.66

```

#### 5. x.vals = seq(-5,5, length = 100) pos = seq(0,5,length = 100)

```

plot(x.vals,dt(x.vals,1),type = "l", lwd = 2, col = 1, ylim = c(0,0.6), xlab = "X", ylab = "P")
lines(x.vals,dt(x.vals,2),type = "l", lwd = 1, col = 2)
lines(x.vals,dt(x.vals,5),type = "l", lwd = 1, col = 3)
lines(x.vals,dt(x.vals,10),type = "l", lwd = 1, col = 4)

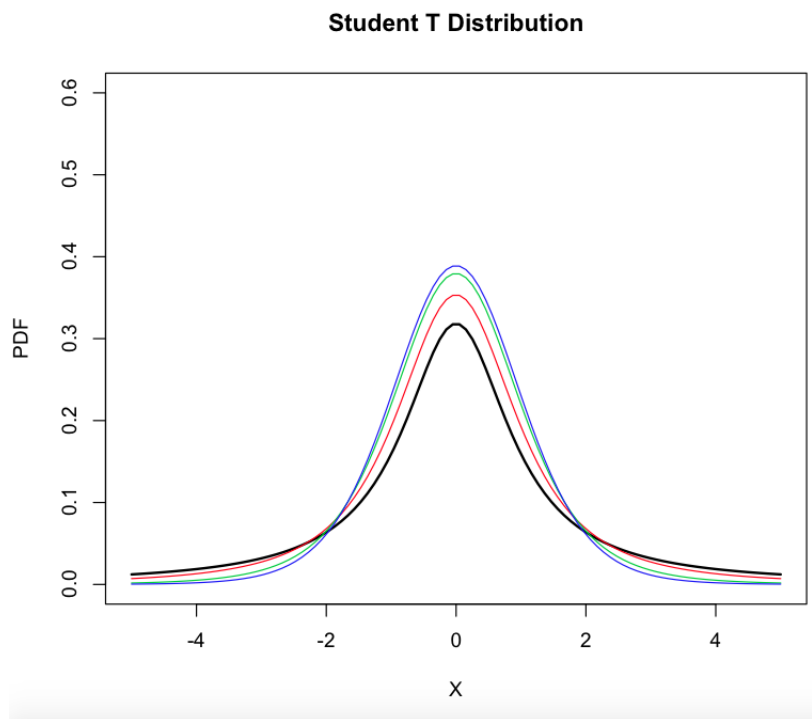
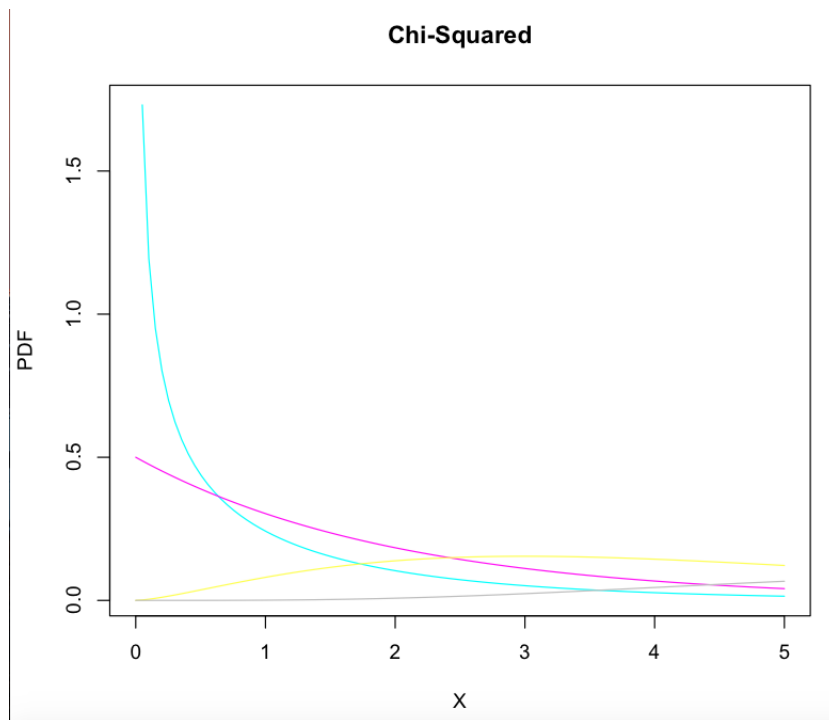
```

#Chi Squared

```

plot(pos,dchisq(pos,1),type = "l", lwd = 1, col = 5, xlab = "X", ylab = "P")
lines(pos,dchisq(pos,2),type = "l", lwd = 1, col = 6)
lines(pos,dchisq(pos,5),type = "l", lwd = 1, col = 7)
lines(pos,dchisq(pos,10),type = "l", lwd = 1, col = 8)

```



```

6. e.x = 1*.3 + 2*.3 + 3 * 0.4
   var.x = 1*.3 + 4*.3 + 9 * .4 - E.x^2
   sd.x = sqrt(var.x)

   e.y = 1 * 0.2 + 2*0.3 + 3*0.5
   var.y = 1* 0.2 + 4 * 0.3 + 9 * 0.5 - e.y^2
   sd.y = sqrt(var.y)

```

```
e.xy = (1*1*0.1) + (1*2*0.2) + (2*1*0.1) + (2*3*0.2) + (3*2*0.1) + (9*0.3)
```

```
cov.xy = e.xy - e.y*e.x
```

```
cor.xy = cov.xy/(sd.x*sd.y)
```

```
> answers <- c(e.x,var.x,sd.x,e.y,var.y,sd.y)
```

a) 2.1000000 0.6900000 0.8306624 2.3000000 0.6100000 0.7810250

b) 0.37, 0.5703117

c) No they are not independent.

$$P(y = 1|x = 1) = \frac{P(x = 1, y = 1)}{P(x = 1)} = \frac{.1}{.3} \neq P(y = 1) = 0.2$$

7. #Problem 7

```
b.amzn = 38.23
```

```
b.costco = 41.11
```

```
s.amzn = 41.29
```

```
s.costco = 41.74
```

```
#Simple Monthly Return
```

```
r.amzn = (s.amzn - b.amzn)/b.amzn
```

```
r.costco = (s.costco - b.costco)/b.costco
```

```
> r.amzn
```

```
[1] 0.08004185
```

```
> r.costco
```

```
[1] 0.01532474
```

```
#CC Return
```

```
cc.amzn = log(s.amzn/b.amzn)
```

```
cc.costco = log(s.costco/b.costco)
```

```
> cc.amzn
```

```
[1] 0.07699979
```

```
> cc.costco
```

```
[1] 0.0152085
```

```
#dividend
```

```
div = 0.1
```

```
rd.costco = (s.costco + div)/(b.costco) - 1
```

```
rd.yield = div/b.costco
```

```
> rd.costco
```

```
[1] 0.01775724
```

```

> rd.yield
[1] 0.002432498

#annualized
annual.amzn = (1+r.amzn)^12 - 1
annual.costco = (1+r.costco)^12 - 1

> annual.amzn
[1] 1.519341
> annual.costco
[1] 0.2002166

annual.amzn.cc = 12 * cc.amzn
annual.costco.cc = 12 * cc.costco

> annual.amzn.cc
[1] 0.9239975
> annual.costco.cc
[1] 0.182502

x_a = 0.8
x_c = 0.2

r.combined = x_a * r.amzn + x_c * r.costco
cc.combined = log(1+r.combined)

> r.combined
[1] 0.06709843
> cc.combined
[1] 0.06494322

> R_e
[1] -0.1333333
> R_uk
[1] 0.125
> R_us
[1] -0.025

```

$$R_e = \frac{e_1 - e_0}{e_0}$$

$$R_uk = \frac{uk_1 - uk_0}{uk_0}$$

$$R_{us} = (1 + R_e)(1 + R_uk) - 1$$

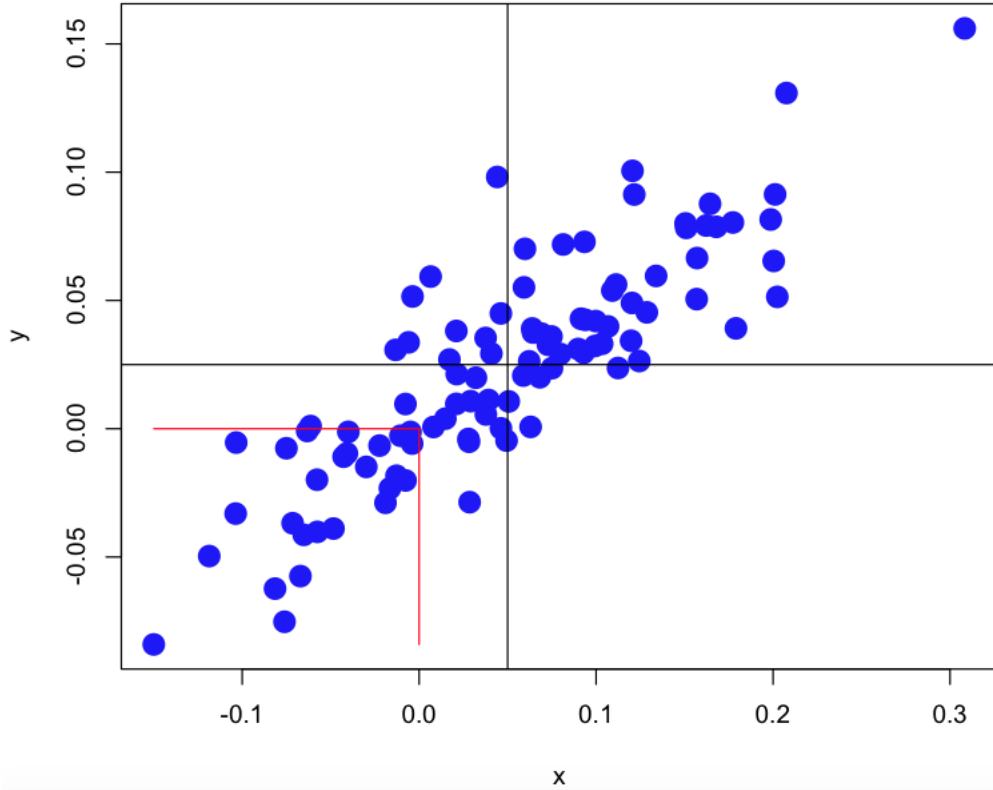
8.

$$cov(R_t, R_{t-1}) = E[R_t R_{t-1}] - E[R_t]E[R_{t-1}]$$

$$\begin{aligned}
0 &= E[R_t R_{t-1}] - E[R_t]E[R_{t-1}] \\
E[R_t R_{t-1}] &= E[R_t]E[R_{t-1}] \\
R_t &\sim N(\mu, \sigma^2) \therefore E[R_t] = \mu \\
&\implies E[R_t R_{t-1}] = \mu^2 \\
E[R_t(2)] &= E[(1 + R_t)(1 + (R_{t-1}) - 1)] \\
&= (1 + E[R_t])(1 + E[R_{t-1}]) - 1 \\
&\implies (1 + \mu)^2 - 1
\end{aligned}$$

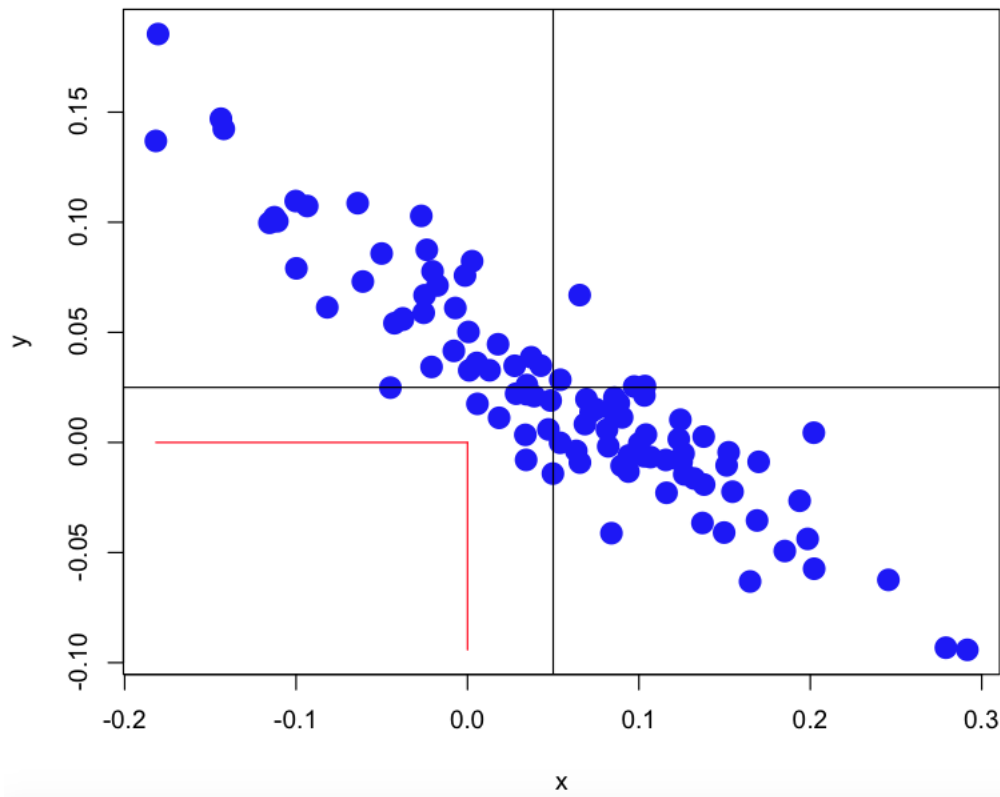
No  $R_t(2)$  is not normally distributed since it is the product of two normal distribution.

**Bivariate normal: rho=0.9**



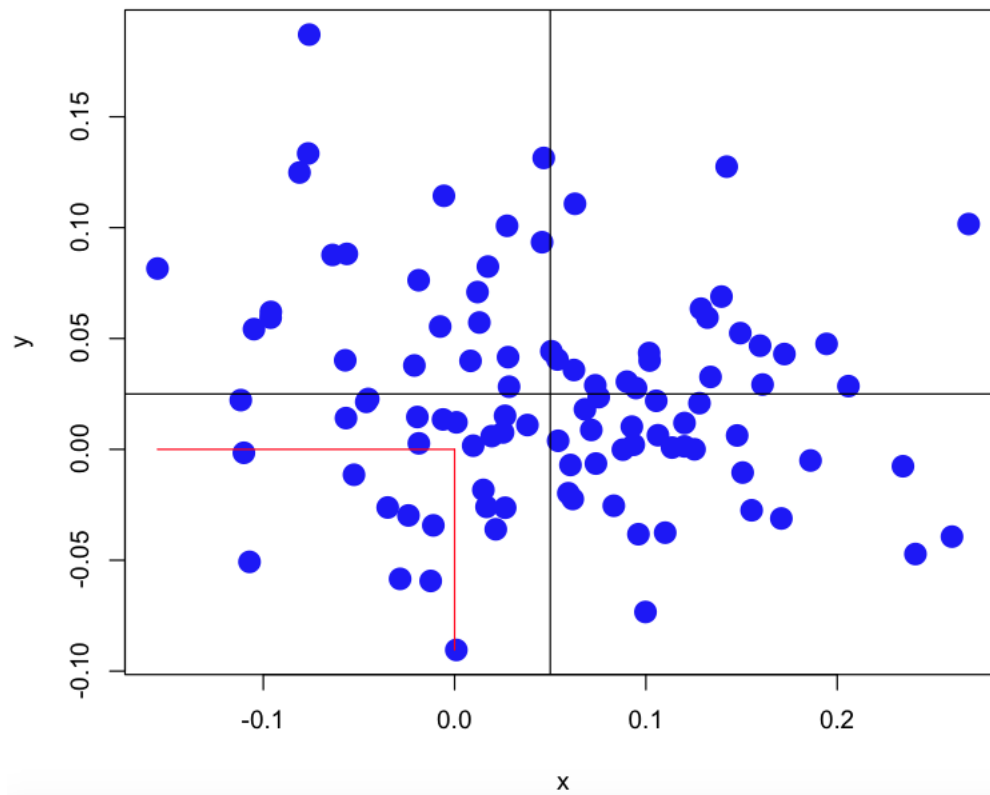
The graph has a positive linear relationship and the joint probability is 0.245

**Bivariate normal: rho=-0.9**



The graph has a negative linear relationship and the joint probability is 0.000803

**Bivariate normal:  $\rho=0$**



The graph has no correlation and the joint probability is 0.0952