Visualisation: Assignment 1

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Dead Line: 03 Oct 2022

Instruction:

1. Work on the `Assignment1.Rmd' file. Compile the file as pdf. Submit only the pdf file in moodle.

2. Make sure you write your name and roll number at the top of the `Assignment1.Rmd' file.

Total Marks: 10 points

Problem 1 (3 points)

Problem Statement: Write an R function which will test Central Limit Theorem.

- Assume the underlying population distribution follow Poisson distribution with rate parameter λ
- We want to estimate the unknown λ with the sample mean

$$\hat{\lambda} = rac{1}{n} \sum_{i=1}^n X_i$$

- The exact sampling distribution of $\hat{\lambda}$ is unknown
- But CLT tells us that as sample size n increases the sampling distribution of $\hat{\lambda}$ can be approximated by Gaussian distribution.

Input in the function:

• n: sample size

• λ : rate parameter

N: simulation size

Output from the function:

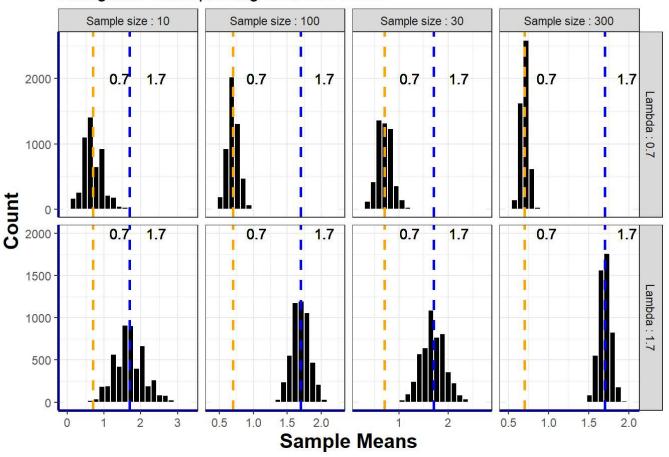
- Histogram of the sampling distribution using ggplot
- QQ-plot using ggplot

Test cases:

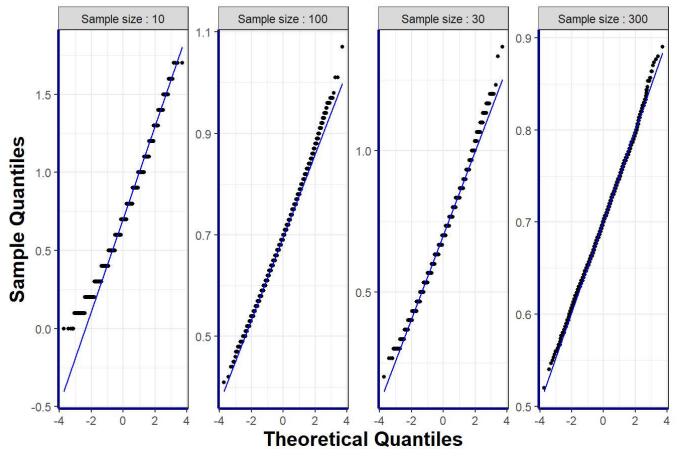
- case 1 a: $\lambda = 0.7$, n=10, N=5000
- case 1 b: $\lambda = 0.7$, n=30, N=5000
- case 1 c: $\lambda=0.7$, n=100, N=5000
- case 1 c: $\lambda = 0.7$, n=300, N=5000
- case 2 a: $\lambda = 1.7$, n=10, N=5000
- case 2 b: $\lambda = 1.7$, n=30, N=5000
- case 2 c: $\lambda = 1.7$, n=100, N=5000

```
cumulative_df = data.frame(means = numeric(0),
                samp_size = character(0), lambda = character(0))
problem_1 = function(lambda, sample_size, sim_size){
  samplings_mean_vec = replicate(sim_size,mean(rpois(sample_size,lambda)))
  samplings_mean_df = data.frame(samplings_mean_vec,
        rep(paste("Sample size :", sample_size,sep = " "),sim_size),
        rep(paste("Lambda :",lambda,sep = " "),sim_size))
 return(samplings mean df)
}
lambdas = c(0.7, 1.7)
test cases = c(010,030,100,300)
simulation size = 5000
for (lambda in lambdas){
 for (test in test cases){
    cumulative_df = rbind(cumulative_df, problem_1(lambda,test,simulation_size))
  }
colnames(cumulative_df) = c('means', 'samp_size', 'lambda')
# Histigrams
print(ggplot(cumulative df)+aes(x=means)+
 geom_histogram(color="white", fill="black",bins=22)+
  geom_vline(xintercept = 0.7, linetype="dashed", color="orange",size=1)+
 geom_vline(xintercept = 1.7, linetype="dashed", color="blue",size=1)+
 #facet_wrap(~samp_size, ncol=4)+
 facet_grid(lambda ~ samp_size, scales = "free")+
 theme(axis.line = element_line(colour = "darkblue", size = 1,
                                 linetype = "solid"),
        axis.title.x = element text(face="bold",size=15),
        axis.title.y = element_text(face="bold",size=15))+
 geom text(aes(0.6,2000,label = 0.7, hjust=-1))+
  geom text(aes(1.6,2000,label = 1.7, hjust=-1))+
  labs(title = "Histograms corresponding to all test-cases",x = "Sample Means",
       y="Count"))
```

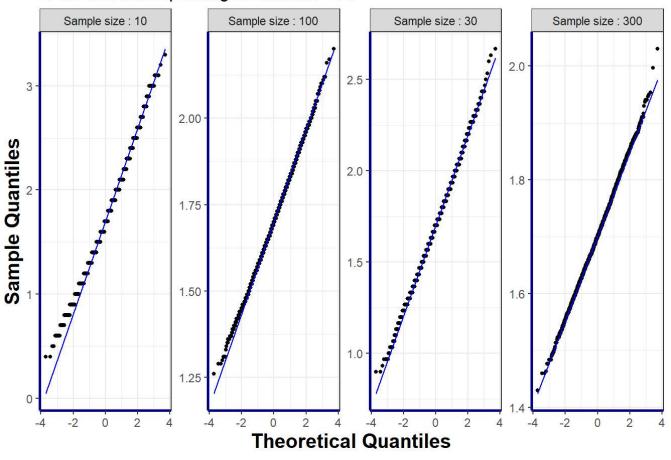
Histograms corresponding to all test-cases



Q-Q Plots corresponding to Lambda = 0.7



Q-Q Plots corresponding to Lambda = 1.7



Problem 2: (2 points)

Consider the Johnson dataset. The datset contains the Quarterly earnings (dollars) per Johnson & Johnson share 1960–80.

a. Draw the time series plot of Quarterly earnings in regular scale and log-scale using the ggplot (1 point)

```
head(JohnsonJohnson)
```

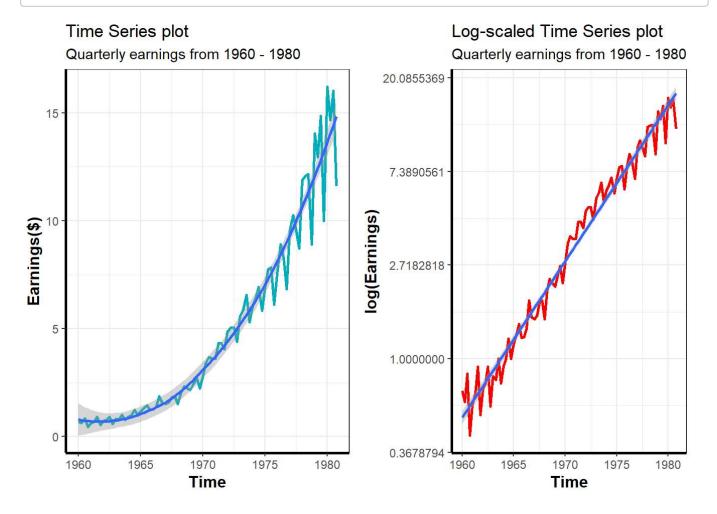
```
## [1] 0.71 0.63 0.85 0.44 0.61 0.69
```

```
problem_2 = function(){
 jj earnings df = data.frame(time = time(JohnsonJohnson),
                              earnings = JohnsonJohnson)
  line = ggplot(data = jj_earnings_df, aes(x = time, y = earnings)) +
    geom_line(color = "#00AFBB", size = 1)+
    geom smooth()+
    theme(axis.line = element_line(colour = "black", size = 1,
                                   linetype = "solid"),
          axis.title.x = element_text(face="bold",size=12),
          axis.title.y = element_text(face="bold",size=12))+
    labs(title = "Time Series plot",
         subtitle = "Quarterly earnings from 1960 - 1980",
         x = "Time",
         y = "Earnings(\$)")
 log = ggplot(data = jj_earnings_df, aes(x = time, y = earnings)) +
    geom_line(color = "red", size = 1) + scale_y_continuous(trans = "log")+
    geom smooth(method = "lm")+
    theme(axis.line = element_line(colour = "black", size = 1,
                                   linetype = "solid"),
          axis.title.x = element_text(face="bold",size=12),
          axis.title.y = element_text(face="bold",size=12))+
    labs(title = "Log-scaled Time Series plot",
         subtitle = "Quarterly earnings from 1960 - 1980",
         x = "Time",
         y = "log(Earnings)")
 figure = ggarrange(line, log, ncol = 2)
  return(figure)
print(problem 2())
```

```
## Don't know how to automatically pick scale for object of type ts. Defaulting to continuou
s.
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s.
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

```
\mbox{\tt ## Don't} know how to automatically pick scale for object of type ts. Defaulting to continuous.
```

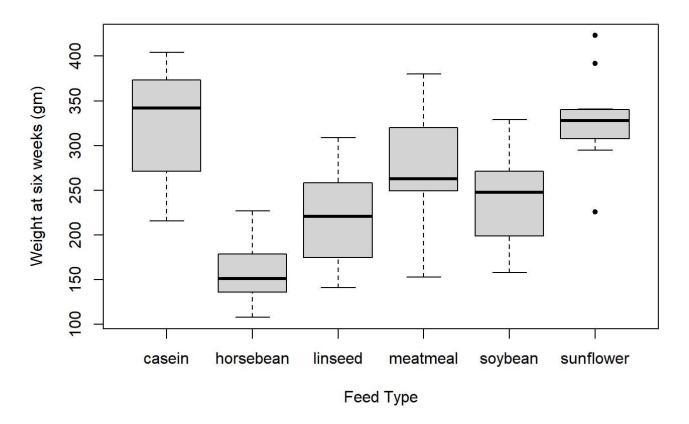


Problem 3: (2 points)

- An experiment was conducted to measure and compare the effectiveness of various feed supplements on the growth rate of chickens.
- Following R-code is a standard side-by-side boxplot showing effect of feed supplements on the growth rate of chickens.

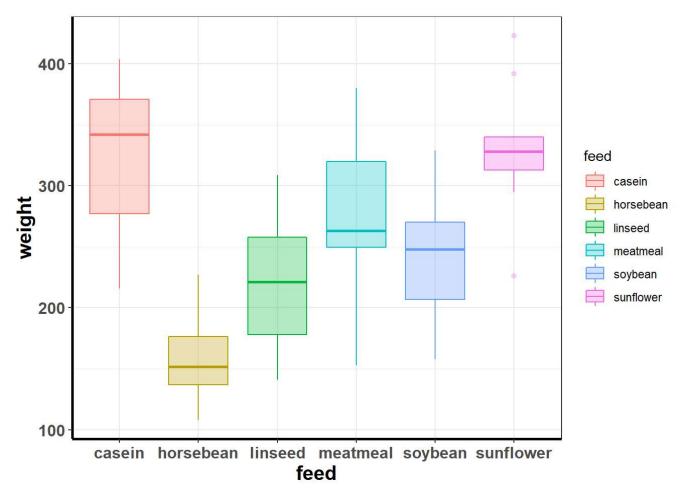
```
boxplot(weight~feed,data=chickwts,pch=20
,main = "chickwt data"
,ylab = "Weight at six weeks (gm)"
,xlab = "Feed Type")
```

chickwt data

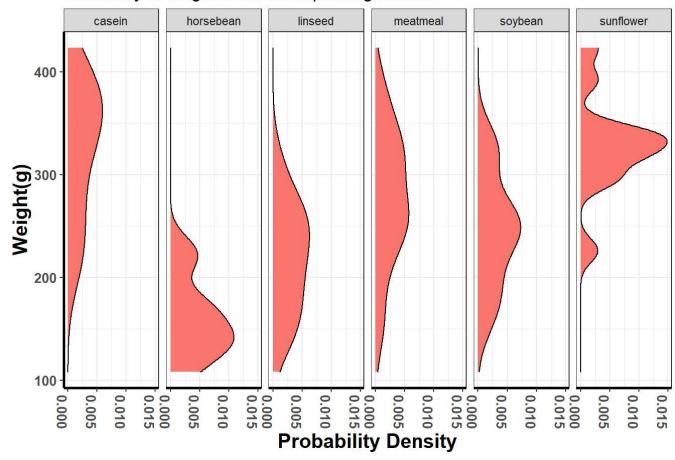


- a. Reproduce the same plot using the ggplot; while fill each boxes with different colour.
- b. In addition draw probability density plot for weights of chicken's growth by each feed seperately using the ggplot . Draw this plot seperately.

```
problem_3 = function(){
 chickwts_df = data.frame(chickwts)
  print(ggplot(chickwts_df, aes(x = feed, y = weight, fill=feed))+
          aes(color = feed)+
          geom_boxplot(alpha=0.3)+
          theme(axis.line = element_line(colour = "black", size = 1,
                                         linetype = "solid"),
                axis.title.x = element_text(face="bold",size=15),
                axis.title.y = element text(face="bold",size=15),
                axis.text.x = element_text(face = "bold",size = 12),
                axis.text.y = element_text(face = "bold",size = 12)))
 # Probability densities
  print(ggplot(chickwts_df)+
          geom_density(aes(y=weight, fill="f8766d"))+
          facet_grid(~feed)+
          theme(axis.line = element_line(colour = "black", size = 1,
                                         linetype = "solid"),
                axis.title.x = element_text(face="bold",size=15),
                axis.title.y = element_text(face="bold",size=15),
                axis.text.x = element_text(angle = -90, face = "bold", size = 10),
                axis.text.y = element_text(face = "bold",size = 10),
                legend.position="none")+
          labs(title = "Probability of weight of chicks depending on feed",
               x = "Probability Density",
               y = "Weight(g)"))
}
problem_3()
```



Probability of weight of chicks depending on feed



Problem 4: (3 points)

Consider the EuStockMarkets data available in R. Contains the daily closing prices of major European stock indices: Germany DAX (Ibis), Switzerland SMI, France CAC, and UK FTSE. The data are sampled in business time, i.e., weekends and holidays are omitted.

head(EuStockMarkets)

```
## DAX SMI CAC FTSE

## [1,] 1628.75 1678.1 1772.8 2443.6

## [2,] 1613.63 1688.5 1750.5 2460.2

## [3,] 1606.51 1678.6 1718.0 2448.2

## [4,] 1621.04 1684.1 1708.1 2470.4

## [5,] 1618.16 1686.6 1723.1 2484.7

## [6,] 1610.61 1671.6 1714.3 2466.8
```

- Suppose P_t is the closing price of a stock indices on day t.
- The daily return r_t is defined as

$$r_t = \log(P_t) - \log(P_{t-1}).$$

- a. Draw time-series plot of P_t for all four markets
- b. Draw time-series plot of r_t for all four markets
- c. Draw histogram of P_t for all four markets
- d. Draw histogram of \boldsymbol{r}_t for all four markets
- e. Suppose you invested \$ 1000 in each market indices on day 1. Plot how your investment grows on the same plot for all four markets. Make your plot using ggplot.
- f. Check which market outperform others during the same time?

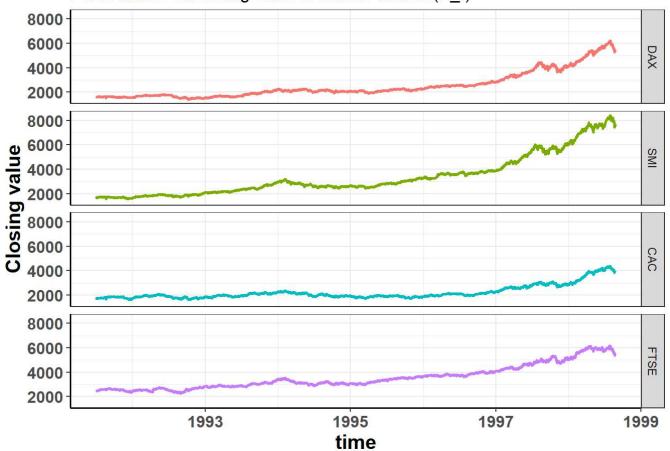
Make all your plots using ggplot.

```
time = data.frame(time = time(EuStockMarkets))
data = data.frame(EuStockMarkets)
n = nrow(data)
for(i in 1:4){
 temp_dr = c(0)
 temp_inv = c(1000)
 for(j in 2:n){
    k = log(data[j,i]/data[j-1,i])
   temp_dr = c(temp_dr,k)
    temp_inv = c(temp_inv,temp_inv[j-1]*exp(k))
 }
 if(i==1){
   dr df = data.frame(IDX dr = temp dr)
    investment df = data.frame(IDX inv = temp inv)
 }
 else{
    dr_df[paste(names(data)[i], 'dr', sep = '_')] = data.frame(temp_dr)
    investment_df[paste(names(data)[i], 'inv', sep = '_')] = data.frame(temp_inv)
 }
}
inv_melt=melt(cbind(time,investment_df),id=c('time'),value.name = "value")
pt melt = melt(cbind(time, data), id = c("time"), value.name = "value")
dr_melt = melt(cbind(time, dr_df), id = c("time"), value.name = "value")
```

a.

Don't know how to automatically pick scale for object of type ts. Defaulting to continuou s.

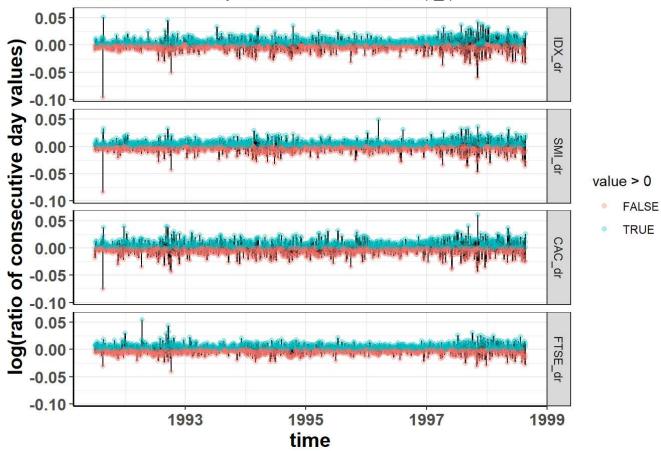
Performance on closing value of various Indices(P t)



b.

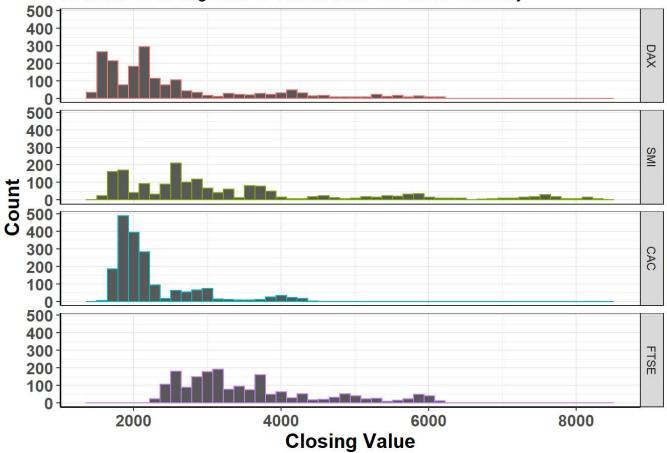
Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.

Performance on daily return of various Indices(r_t)



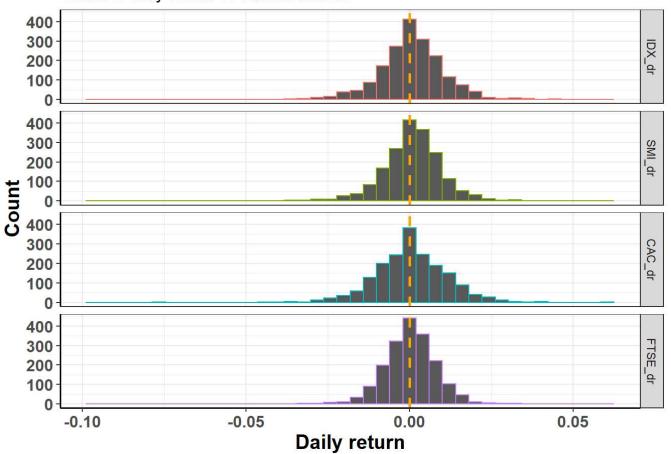
C.

Deviation of closing value of various indices from its centrality



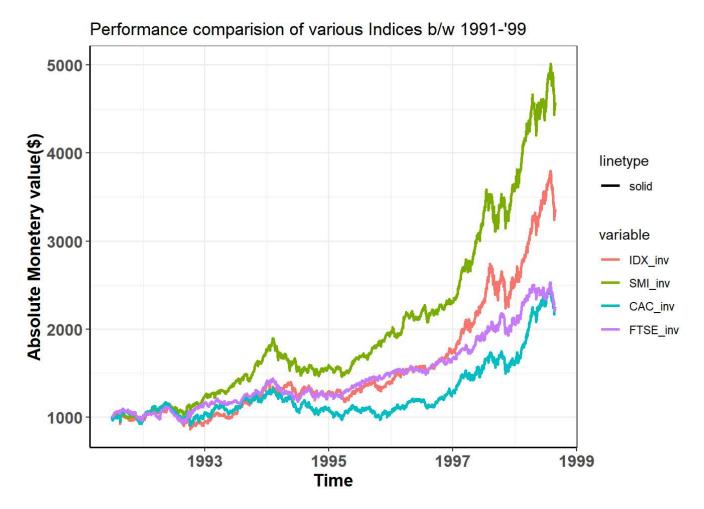
d.

Count of daily return of various indices



e.

Don't know how to automatically pick scale for object of type ts. Defaulting to continuou s.



f. By analyzing the last graph(e), we get to know that, the SMI(Swiss Market Index) outperforms all the the other indices (DAX, CAC and FTSE) for almost the entire time-span given(1991-1999).