

# README

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12/4/2021

Question 3:

In this brief report, I aim to show the JSE the concentration and commonality of returns within the Top 40 index (J200). In order to do so, I will look at using Principal Component Analysis and rolling constituent correlation perspective to illustrate. In my answer I will use stratification of monthly volatility by calculating the J200 index returns and comparing return source concentrations for periods of high volatility only.

```
# Importing the data
```

```
T40 <-readRDS("/Users/mathlogonolomashitisho/Desktop/Economics Masters /Semester 2/Fin Metrics/Financial
```

Let begin by stratifying the returns for times during high volatility.

```
library(rmsfun)
```

```
pacman::p_load("tidyr", "tbl2xts", "devtools", "lubridate", "readr", "PerformanceAnalytics", "ggplot2", "c
```

```
#Here , there is no need to consider before 2008 because the data has already provided for the survivin
```

```
T40 <- T40 %>% group_by(date) %>% dplyr::mutate(Return= coalesce(Return,0)) %>% dplyr::mutate(J200= coa
```

```
T40 <-
```

```
  T40 %>% group_by(Tickers) %>%
```

```
  mutate(Top = quantile(J200return, 0.99), Bot = quantile(J200return, 0.01)) %>%
```

```
  mutate(Return = ifelse(J200return> Top, Top,
```

```
                    ifelse(J200return < Bot, Bot, J200return))) %>% ungroup() %>% mutate(YearMonth
```

```
T40SD <-T40  %>%
```

```
  mutate(YearMonth = format(date, "%Y%B")) %>%
```

```
  group_by(YearMonth) %>% summarise(SD = sd(J200return)*sqrt(52)) %>%
```

```
# Top Decile Quantile overall (highly volatile month for ZAR:
```

```
  mutate(TopQtile = quantile(SD, 0.8),
```

```
        BotQtile = quantile(SD, 0.2))
```

```

Hi_Vol <- T40SD %>% filter(SD > TopQtile) %>% pull(YearMonth)

Low_Vol <- T40SD %>% filter(SD < BotQtile) %>% pull(YearMonth)

# Create generic function to compare performance:

Perf_comparisons <- function(Idxs, YMs, Alias){
  # For stepping through uncomment:
  YMs <- Hi_Vol

  Unconditional_SD <-

  Idxs %>%

    group_by(Tickers) %>%

    mutate(Full_SD = sd(Return) * sqrt(252)) %>%

    filter(YearMonth %in% YMs) %>%

    summarise(SD = sd(Return) * sqrt(252), across(.cols = starts_with("Full"), .fns = max)) %>%

    arrange(desc(SD)) %>% mutate(Period = Alias) %>%

    group_by(Tickers) %>%

    mutate(Ratio = SD / Full_SD)

    Unconditional_SD

}

```

```

#Extract the Tickers with monthly high volatility

perf_hi <- Perf_comparisons(T40, YMs = Hi_Vol, Alias = "High_Vol")

perf_lo <- Perf_comparisons(T40, YMs = Low_Vol, Alias = "Low_Vol")

```

PCA Analysis:

```

#Removing rows with missing values

perf_hi <- na.omit(perf_hi)

```

Mean centering J200returns

```

T40_Centered <- T40 %>% group_by(Tickers) %>% mutate(J200return_centered = J200return -
  mean(J200return)) %>% ungroup()

```

Here we are only interested in Hi volatility periods as stratified above.

```
pacman::p_load(fEcofin)

# prcomp requires wide, numeric data:

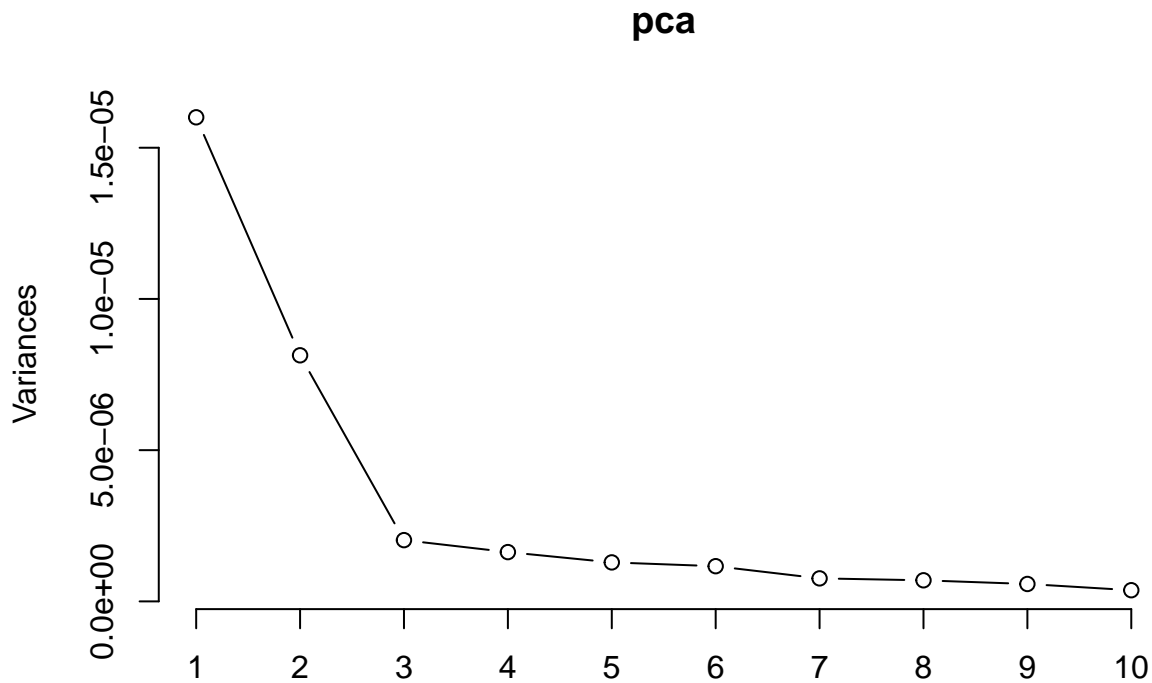
data_wide <- T40_Centered %>% filter(T40_Centered$Tickers %in% perf_hi$Tickers) %>% dplyr::mutate(Returns = J200return_centered) %>% select(-date)

# make sure to account for only those in portfolio
temp<-data_wide[, colSums(is.na(data_wide)) != nrow(data_wide)]

data_wide <- temp %>%
  mutate(
    across(everything(), replace_na, 0)
  )

pca <- prcomp(data_wide)
# We have already centered our data, but we could also use
# these built in measures: prcomp(data_wide, center = TRUE,
# scale. = TRUE)

plot(pca, type = "l")
```



```
summary(pca)
```

```
## Importance of components:
##              PC1      PC2      PC3      PC4      PC5      PC6
## Standard deviation 0.004001 0.002852 0.001423 0.001276 0.001135 0.001079
## Proportion of Variance 0.467120 0.237460 0.059110 0.047560 0.037620 0.033960
## Cumulative Proportion 0.467120 0.704580 0.763700 0.811250 0.848870 0.882840
```

##		PC7	PC8	PC9	PC10	PC11
##	Standard deviation	0.0008722	0.0008349	0.0007587	0.0006079	0.0004087
##	Proportion of Variance	0.0222000	0.0203400	0.0168000	0.0107900	0.0048700
##	Cumulative Proportion	0.9050400	0.9253800	0.9421800	0.9529600	0.9578400
##		PC12	PC13	PC14	PC15	PC16
##	Standard deviation	0.0003829	0.0003571	0.0003364	0.0003143	0.000282
##	Proportion of Variance	0.0042800	0.0037200	0.0033000	0.0028800	0.002320
##	Cumulative Proportion	0.9621100	0.9658400	0.9691400	0.9720200	0.974340
##		PC17	PC18	PC19	PC20	PC21
##	Standard deviation	0.0002768	0.0002703	0.0002572	0.000231	0.0002264
##	Proportion of Variance	0.0022400	0.0021300	0.0019300	0.001560	0.0015000
##	Cumulative Proportion	0.9765800	0.9787100	0.9806400	0.982200	0.9836900
##		PC22	PC23	PC24	PC25	PC26
##	Standard deviation	0.0002146	0.0002088	0.0001835	0.0001755	0.0001707
##	Proportion of Variance	0.0013400	0.0012700	0.0009800	0.0009000	0.0008500
##	Cumulative Proportion	0.9850400	0.9863100	0.9872900	0.9881900	0.9890400
##		PC27	PC28	PC29	PC30	PC31
##	Standard deviation	0.0001667	0.0001534	0.0001494	0.0001406	0.0001396
##	Proportion of Variance	0.0008100	0.0006900	0.0006500	0.0005800	0.0005700
##	Cumulative Proportion	0.9898500	0.9905400	0.9911900	0.9917700	0.9923400
##		PC32	PC33	PC34	PC35	PC36
##	Standard deviation	0.0001355	0.0001313	0.0001274	0.000125	0.000119
##	Proportion of Variance	0.0005400	0.0005000	0.0004700	0.000460	0.000410
##	Cumulative Proportion	0.9928700	0.9933800	0.9938500	0.994310	0.994720
##		PC37	PC38	PC39	PC40	PC41
##	Standard deviation	0.0001164	0.0001123	0.0001066	0.0001058	0.0001013
##	Proportion of Variance	0.0004000	0.0003700	0.0003300	0.0003300	0.0003000
##	Cumulative Proportion	0.9951200	0.9954800	0.9958200	0.9961400	0.9964400
##		PC42	PC43	PC44	PC45	PC46
##	Standard deviation	9.439e-05	9.071e-05	8.701e-05	8.538e-05	8.122e-05
##	Proportion of Variance	2.600e-04	2.400e-04	2.200e-04	2.100e-04	1.900e-04
##	Cumulative Proportion	9.967e-01	9.969e-01	9.972e-01	9.974e-01	9.976e-01
##		PC47	PC48	PC49	PC50	PC51
##	Standard deviation	7.946e-05	7.675e-05	7.607e-05	7.534e-05	7.418e-05
##	Proportion of Variance	1.800e-04	1.700e-04	1.700e-04	1.700e-04	1.600e-04
##	Cumulative Proportion	9.978e-01	9.979e-01	9.981e-01	9.983e-01	9.984e-01
##		PC52	PC53	PC54	PC55	PC56
##	Standard deviation	6.831e-05	0.0000662	6.387e-05	6.198e-05	5.748e-05
##	Proportion of Variance	1.400e-04	0.0001300	1.200e-04	1.100e-04	1.000e-04
##	Cumulative Proportion	9.986e-01	0.9986800	9.988e-01	9.989e-01	9.990e-01
##		PC57	PC58	PC59	PC60	PC61
##	Standard deviation	5.676e-05	5.547e-05	5.454e-05	5.088e-05	4.965e-05
##	Proportion of Variance	9.000e-05	9.000e-05	9.000e-05	8.000e-05	7.000e-05
##	Cumulative Proportion	9.991e-01	9.992e-01	9.993e-01	9.994e-01	9.994e-01
##		PC62	PC63	PC64	PC65	PC66
##	Standard deviation	4.769e-05	4.712e-05	4.393e-05	4.341e-05	4.172e-05
##	Proportion of Variance	7.000e-05	6.000e-05	6.000e-05	5.000e-05	5.000e-05
##	Cumulative Proportion	9.995e-01	9.996e-01	9.996e-01	9.997e-01	9.997e-01
##		PC67	PC68	PC69	PC70	PC71
##	Standard deviation	0.0000383	3.321e-05	0.0000328	3.133e-05	2.967e-05
##	Proportion of Variance	0.0000400	3.000e-05	0.0000300	3.000e-05	3.000e-05
##	Cumulative Proportion	0.9997700	9.998e-01	0.9998300	9.999e-01	9.999e-01
##		PC72	PC73	PC74	PC75	PC76
##	Standard deviation	2.761e-05	2.621e-05	2.522e-05	2.469e-05	1.898e-05

```

## Proportion of Variance 2.000e-05 2.000e-05 2.000e-05 2.000e-05 1.000e-05
## Cumulative Proportion 9.999e-01 9.999e-01 9.999e-01 1.000e+00 1.000e+00
##          PC77          PC78          PC79          PC80          PC81
## Standard deviation    1.724e-05 1.681e-05 1.436e-05 1.146e-05 3.254e-06
## Proportion of Variance 1.000e-05 1.000e-05 1.000e-05 0.000e+00 0.000e+00
## Cumulative Proportion 1.000e+00 1.000e+00 1.000e+00 1.000e+00 1.000e+00
##          PC82
## Standard deviation    1.04e-06
## Proportion of Variance 0.00e+00
## Cumulative Proportion 1.00e+00

```

AS can be seen in the skree plot above, the elbow of the plot is at 2. This means that there are 2 components in the high volatility periods that can explain the sources of returns.