

Happiest Countries

Rory Quinlan

Set Up

```
library(tidyverse)
library(dplyr)
library(factoextra)
library(cluster)
library(gridExtra)

# Load data
happiness_0 = read_csv("happiness_2018.csv", show_col_types = FALSE)

happiness_1 = happiness_0 %>%
  mutate(Corruption = as.numeric(`Perceptions of corruption`))

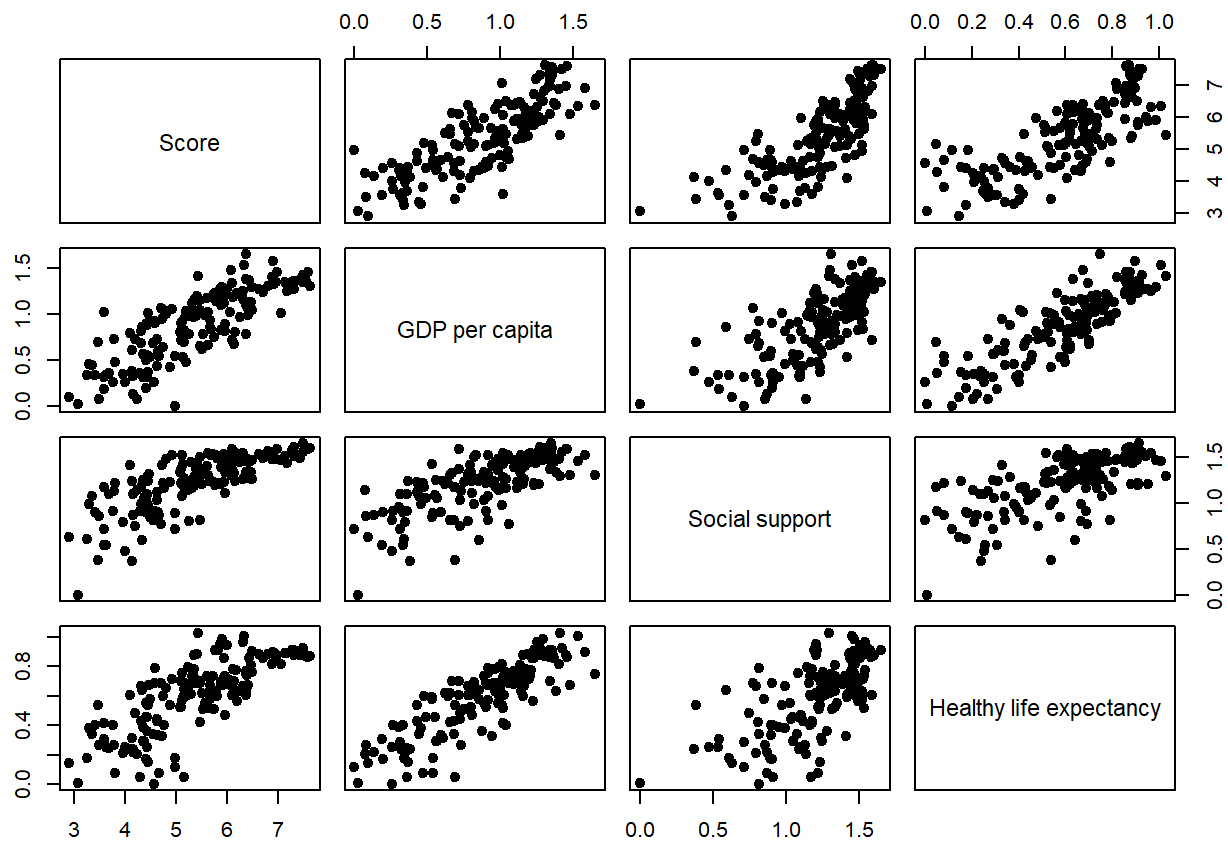
# Remove NA values
happiness_2 = happiness_1 %>% drop_na()
dim(happiness_1)
```

```
## [1] 156 10
```

```
happiness_1 = happiness_2 %>%
  select(-c(`Overall rank`, `Country or region`, `Perceptions of corruption`))
```

Explore Data

```
# CorreLogram
pairs(happiness_1[,1:4], pch = 19)
```



PCA

```
# Create A PCA
pc.happiness = happiness_1 %>%
  select(-Score) %>%
  prcomp(scale=TRUE)
pc.happiness
```

```
## Standard deviations (1, .., p=6):
## [1] 1.7286165 1.1856234 0.7717936 0.7501125 0.5653144 0.3581444
##
## Rotation (n x k) = (6 x 6):
##
```

	PC1	PC2	PC3	PC4
## GDP per capita	-0.5126579	0.2666787	-0.1164253	-0.17098649
## Social support	-0.4720852	0.2376977	-0.1824560	0.27652670
## Healthy life expectancy	-0.5038326	0.2424268	-0.1581973	-0.20850817
## Freedom to make life choices	-0.3752315	-0.3544097	0.4418517	0.68603484
## Generosity	-0.1219418	-0.6758505	-0.7238170	0.03330243
## Corruption	-0.3237308	-0.4808653	0.4571483	-0.61568506

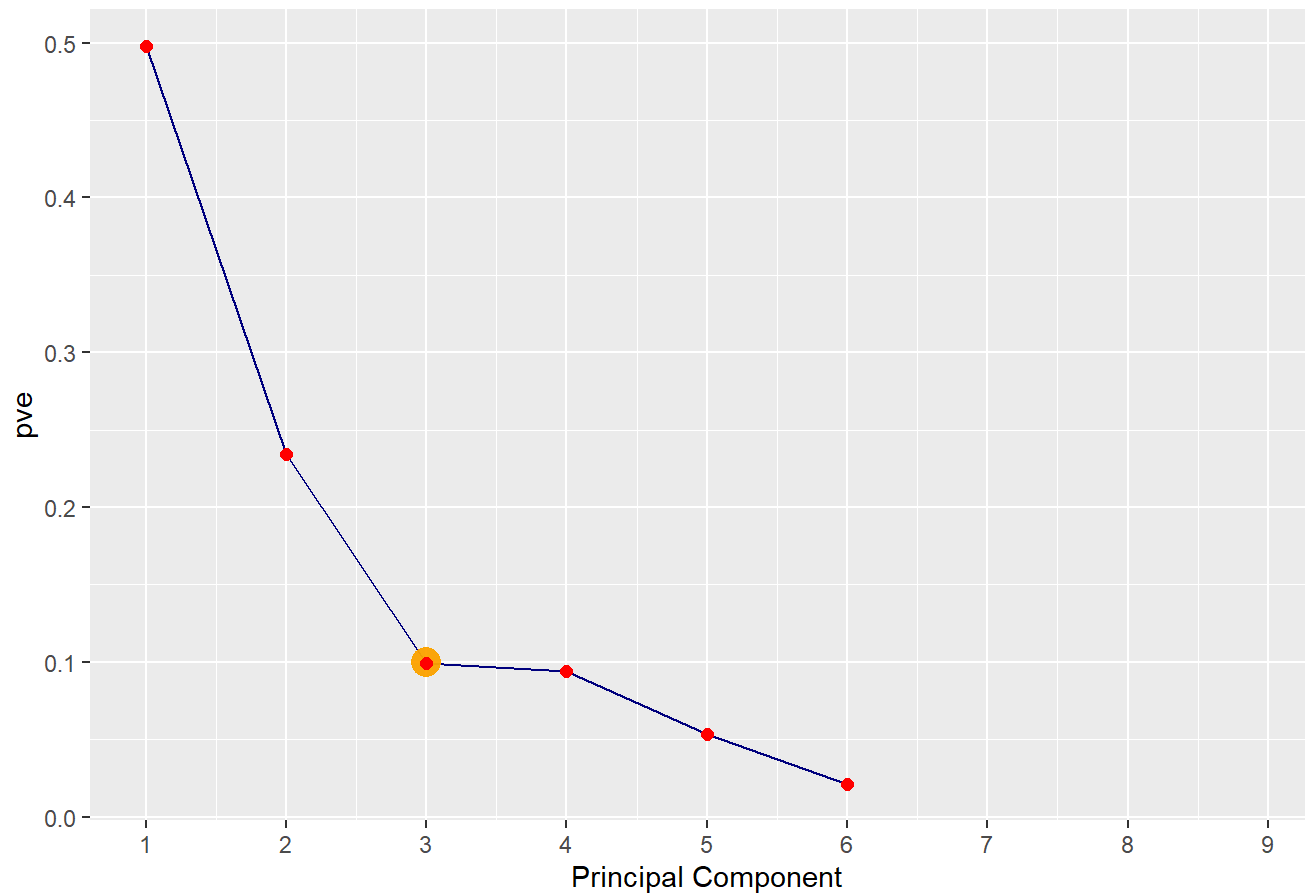
```
##
```

	PC5	PC6
## GDP per capita	0.23122805	-0.75485540
## Social support	-0.76997346	0.13423480
## Healthy life expectancy	0.45787741	0.63970950
## Freedom to make life choices	0.25983295	-0.01432444
## Generosity	0.04285037	-0.03873052
## Corruption	-0.27326438	0.03522586

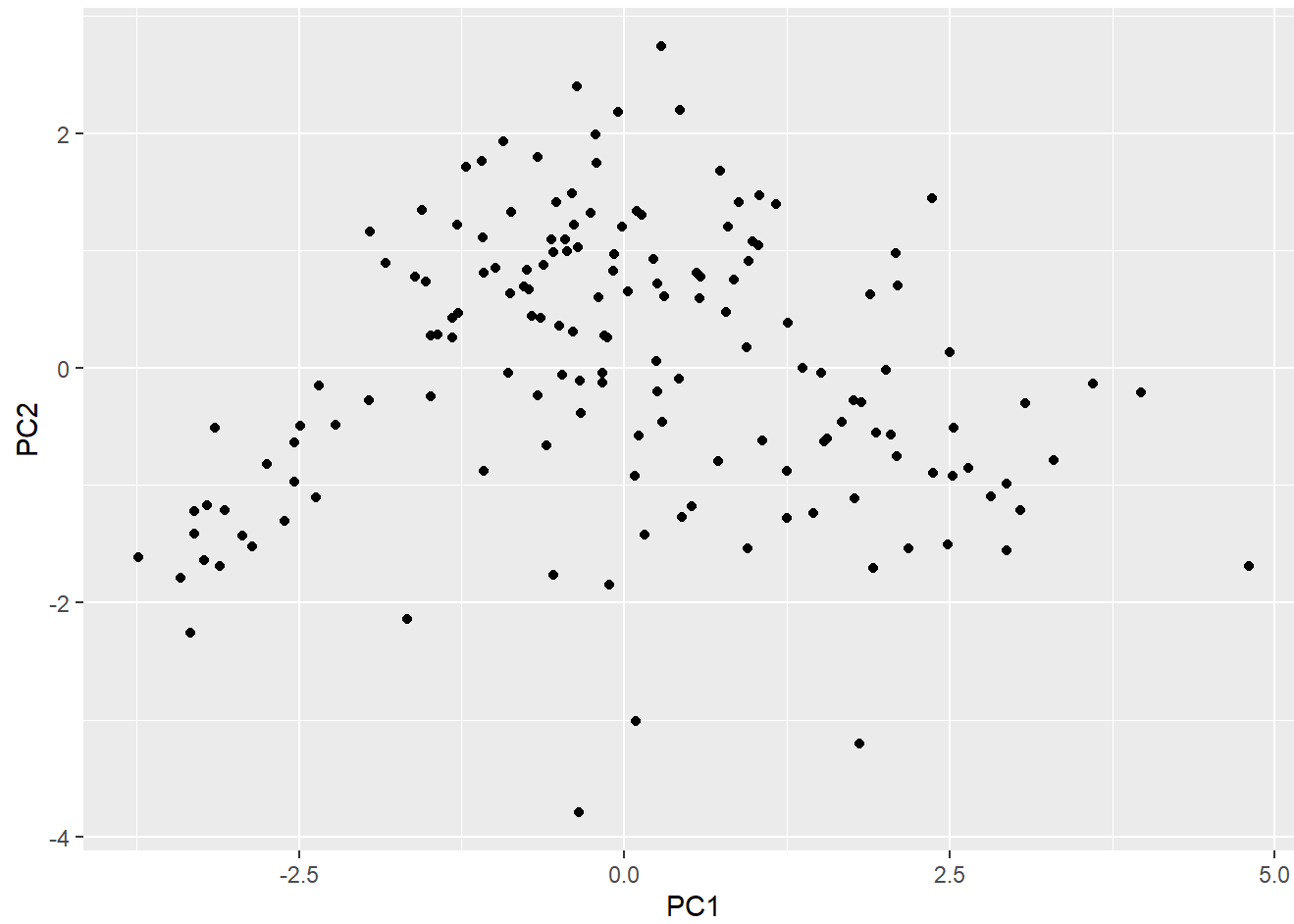
```
# Graph Explanatory power of variables
PRVar<- pc.happiness$sdev^2
PVE<- PRVar[1:9]/sum(PRVar)

PC=1:9
data=data.frame(PC, PVE)
ggplot(data=data, aes(x=PC, y=PVE))+
  geom_line(color="navy")+
  geom_point(aes(x=3,y=0.1),cex=5,color="orange",alpha=0.3)+
  geom_point(color="red",cex=2)+
  labs(title="Proportion of Variance Explained", x="Principal Component",y="pve")+
  scale_x_continuous(breaks = 1:9)
```

Proportion of Variance Explained

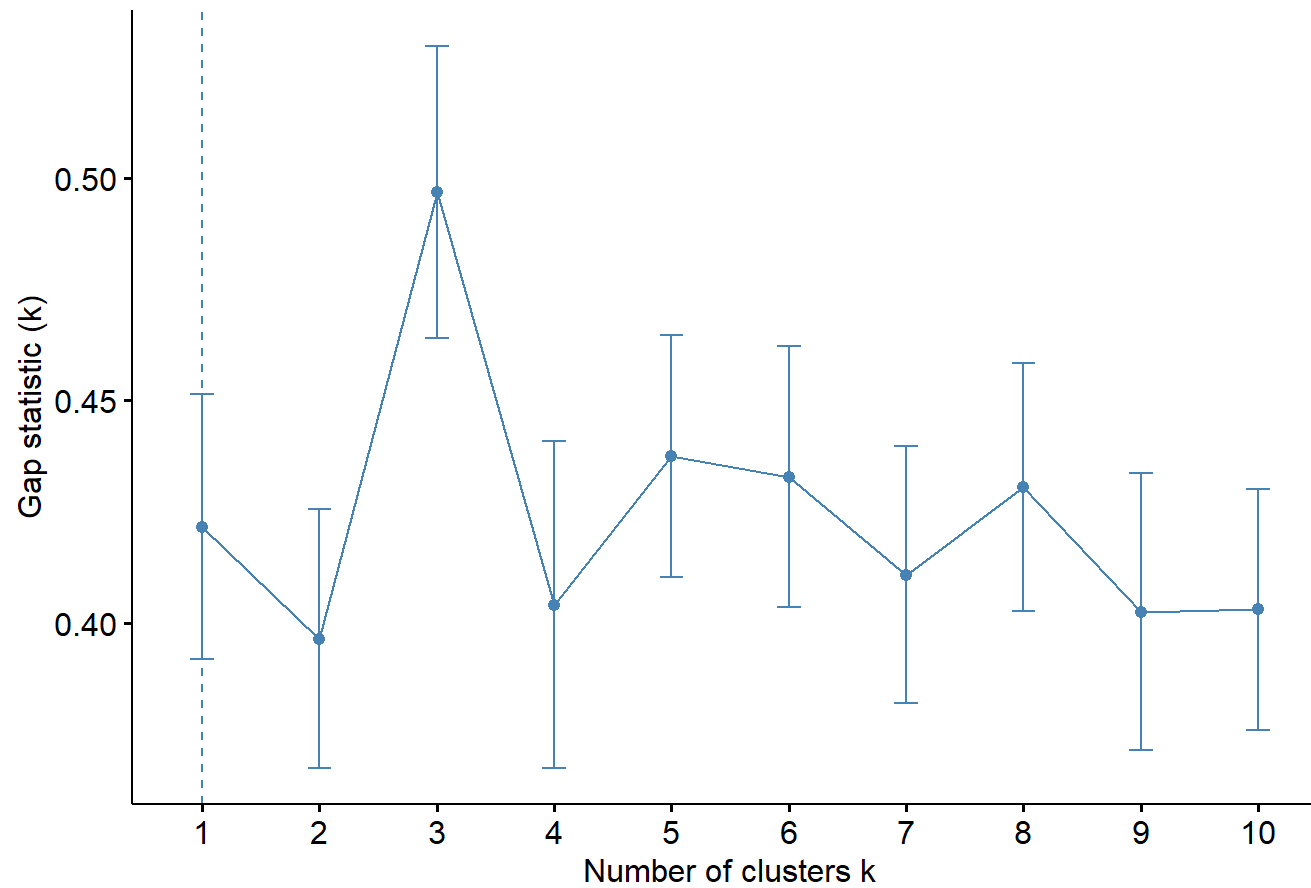


```
# View scatterplot of greatest contributors from PCA
PC12 <- pc.happiness$x %>% as_tibble() %>% select(1:2)
pc.happiness$x %>% as_tibble() %>% select(PC1, PC2) %>%
  bind_cols(happiness_1) %>%
  ggplot(aes(x = `PC1`, y = `PC2`)) + geom_point()
```



```
# Find optimal clusters  
happiness_4 = scale(happiness_1[, -1])  
fviz_nbclust(PC12, kmeans, method = "gap_stat")
```

Optimal number of clusters



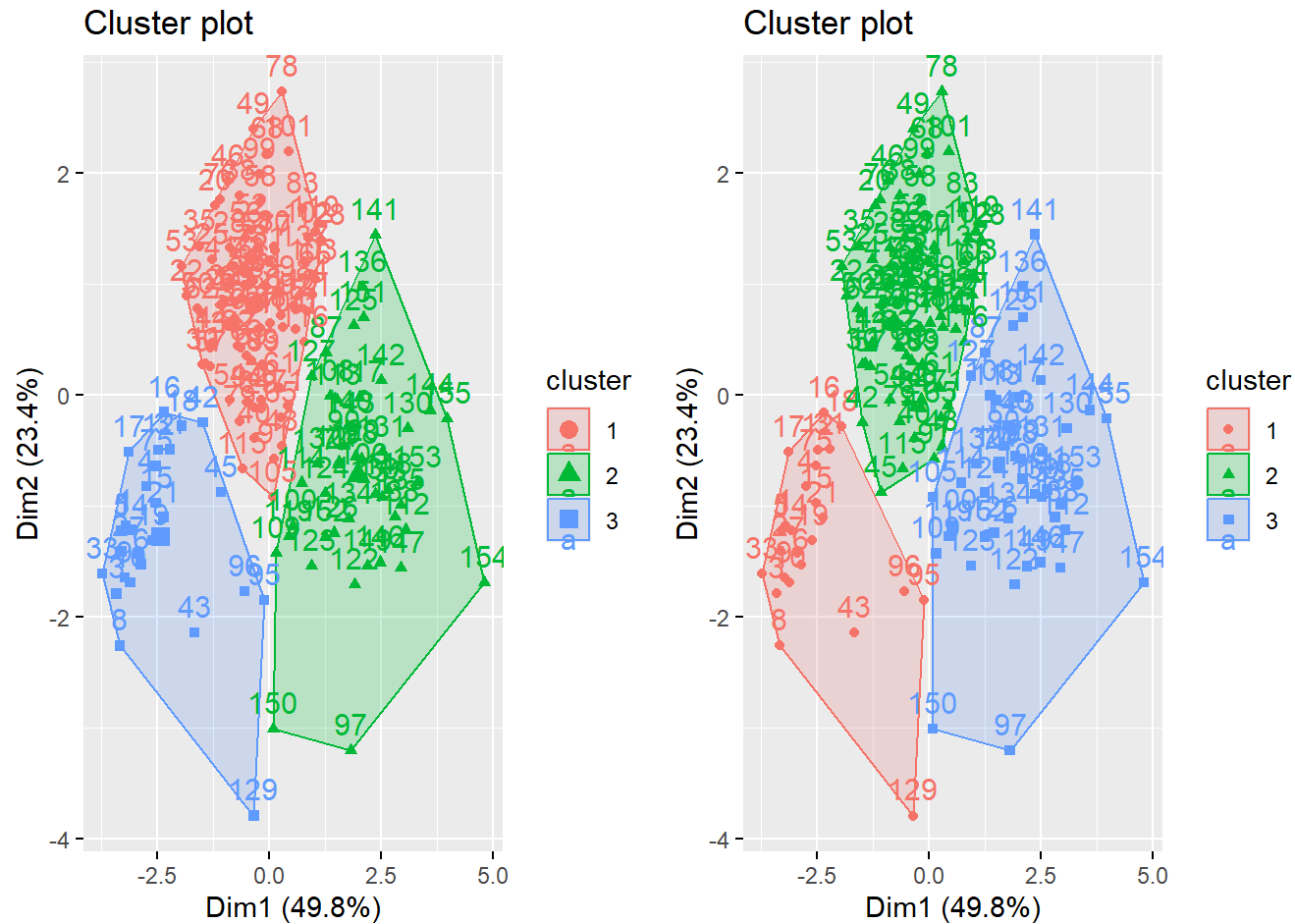
```
km_mod = kmeans(happiness_4, centers=3)
pam_mod = pam(happiness_4, 3)
```

```
# Create variable for cluster number in df
PC12_cluster = happiness_1 %>% mutate(cluster=factor(pam_mod$cluster))
```

```
# Compare kmean and pam
p1<-fviz_cluster(km_mod, data = happiness_4)

p2<- fviz_cluster(pam_mod, data = happiness_4)

grid.arrange(p1, p2, ncol=2)
```



```
# Create clusters with country name
```

```
PC12_cluster <- PC12 %>% mutate(cluster = factor(pam_mod$cluster), countryOrRegion = factor(happiness_2$`Country or region`))
```

```
# Filter and display the happiest countries
```

```
as.data.frame(PC12_cluster %>% filter(cluster == 1) %>% select(countryOrRegion))
```

```
##      countryOrRegion
## 1      Finland
## 2      Norway
## 3      Denmark
## 4      Iceland
## 5      Switzerland
## 6      Netherlands
## 7      Canada
## 8      New Zealand
## 9      Sweden
## 10     Australia
## 11     Austria
## 12     Ireland
## 13     Germany
## 14     Belgium
## 15     Luxembourg
## 16     United States
## 17     Israel
## 18     Malta
## 19     Qatar
## 20     Singapore
## 21     Uzbekistan
## 22     Hong Kong
## 23     Indonesia
## 24     Bhutan
## 25     Myanmar
```