

# Principal Component Analysis (PCA) dan Factor Analysis (FA) pada Data Penjualan Mobil

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## R Markdown

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When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
df <- read.csv("Clean_Car_sales.csv")
head(df)
```

```
##   Sales_in_thousands X__year_resale_value Price_in_thousands Engine_size
## 1             16.919             16.360             21.50000             1.8
## 2             39.384             19.875             28.40000             3.2
## 3             14.114             18.225             27.39075             3.2
## 4              8.588             29.725             42.00000             3.5
## 5             20.397             22.255             23.99000             1.8
## 6             18.780             23.555             33.95000             2.8
##   Horsepower Wheelbase Width Length Curb_weight Fuel_capacity Fuel_efficiency
## 1         140      101.2   67.3  172.4         2.639          13.2             28
## 2         225      108.1   70.3  192.9         3.517          17.2             25
## 3         225      106.9   70.6  192.0         3.470          17.2             26
## 4         210      114.6   71.4  196.6         3.850          18.0             22
## 5         150      102.6   68.2  178.0         2.998          16.4             27
## 6         200      108.7   76.1  192.0         3.561          18.5             22
##   Power_perf_factor
## 1         58.28015
## 2         91.37078
## 3         77.04359
## 4         91.38978
## 5         62.77764
## 6         84.56511
```

```
str(df)
```

```
## 'data.frame':   157 obs. of  12 variables:
##  $ Sales_in_thousands : num  16.92 39.38 14.11 8.59 20.4 ...
##  $ X__year_resale_value: num  16.4 19.9 18.2 29.7 22.3 ...
##  $ Price_in_thousands : num  21.5 28.4 27.4 42 24 ...
##  $ Engine_size         : num  1.8 3.2 3.2 3.5 1.8 2.8 4.2 2.5 2.8 2.8 ...
##  $ Horsepower          : num  140 225 225 210 150 200 310 170 193 193 ...
##  $ Wheelbase           : num  101 108 107 115 103 ...
##  $ Width               : num  67.3 70.3 70.6 71.4 68.2 76.1 74 68.4 68.5 70.9 ...
```

```
## $ Length          : num  172 193 192 197 178 ...
## $ Curb_weight      : num   2.64 3.52 3.47 3.85 3 ...
## $ Fuel_capacity    : num  13.2 17.2 17.2 18 16.4 18.5 23.7 16.6 16.6 18.5 ...
## $ Fuel_efficiency  : num   28 25 26 22 27 22 21 26 24 25 ...
## $ Power_perf_factor : num  58.3 91.4 77 91.4 62.8 ...
```

```
summary(df)
```

```
## Sales_in_thousands X__year_resale_value Price_in_thousands Engine_size
## Min.   : 0.11      Min.   : 5.16      Min.   : 9.235      Min.   :1.000
## 1st Qu.: 14.11     1st Qu.:12.54      1st Qu.:18.145     1st Qu.:2.300
## Median : 29.45     Median :17.71      Median :23.400     Median :3.000
## Mean   : 53.00     Mean   :18.07      Mean   :27.391     Mean   :3.061
## 3rd Qu.: 67.96     3rd Qu.:18.14      3rd Qu.:31.930     3rd Qu.:3.500
## Max.   :540.56     Max.   :67.55      Max.   :85.500     Max.   :8.000
## Horsepower      Wheelbase      Width      Length
## Min.   : 55.0     Min.   : 92.6     Min.   :62.60     Min.   :149.4
## 1st Qu.:150.0     1st Qu.:103.0     1st Qu.:68.40     1st Qu.:177.6
## Median :180.0     Median :107.0     Median :70.60     Median :187.8
## Mean   :185.9     Mean   :107.5     Mean   :71.15     Mean   :187.3
## 3rd Qu.:215.0     3rd Qu.:112.2     3rd Qu.:73.40     3rd Qu.:196.1
## Max.   :450.0     Max.   :138.7     Max.   :79.90     Max.   :224.5
## Curb_weight      Fuel_capacity      Fuel_efficiency Power_perf_factor
## Min.   :1.895     Min.   :10.30     Min.   :15.00     Min.   : 23.28
## 1st Qu.:2.975     1st Qu.:15.80     1st Qu.:21.00     1st Qu.: 60.73
## Median :3.368     Median :17.20     Median :24.00     Median : 72.29
## Mean   :3.378     Mean   :17.95     Mean   :23.84     Mean   : 77.04
## 3rd Qu.:3.778     3rd Qu.:19.50     3rd Qu.:26.00     3rd Qu.: 89.40
## Max.   :5.572     Max.   :32.00     Max.   :45.00     Max.   :188.14
```

```
sum(is.na(df))
```

```
## [1] 0
```

```
p <- ncol(df)
print(p)
```

```
## [1] 12
```

```
library(psych)
r <- cor(df)
KMO(r)
```

```
## Kaiser-Meyer-Olkin factor adequacy
## Call: KMO(r = r)
## Overall MSA = 0.83
## MSA for each item =
## Sales_in_thousands X__year_resale_value Price_in_thousands
##           0.72           0.93           0.67
## Engine_size      Horsepower      Wheelbase
##           0.95           0.72           0.82
## Width           Length           Curb_weight
##           0.97           0.84           0.88
## Fuel_capacity    Fuel_efficiency Power_perf_factor
##           0.90           0.92           0.71
```

```
bartlett.test(df)
```

```
##
## Bartlett test of homogeneity of variances
##
## data: df
## Bartlett's K-squared = 4692, df = 11, p-value < 2.2e-16
df<- scale(df)

pca_result <- prcomp(df, scale. = TRUE)

summary(pca_result)

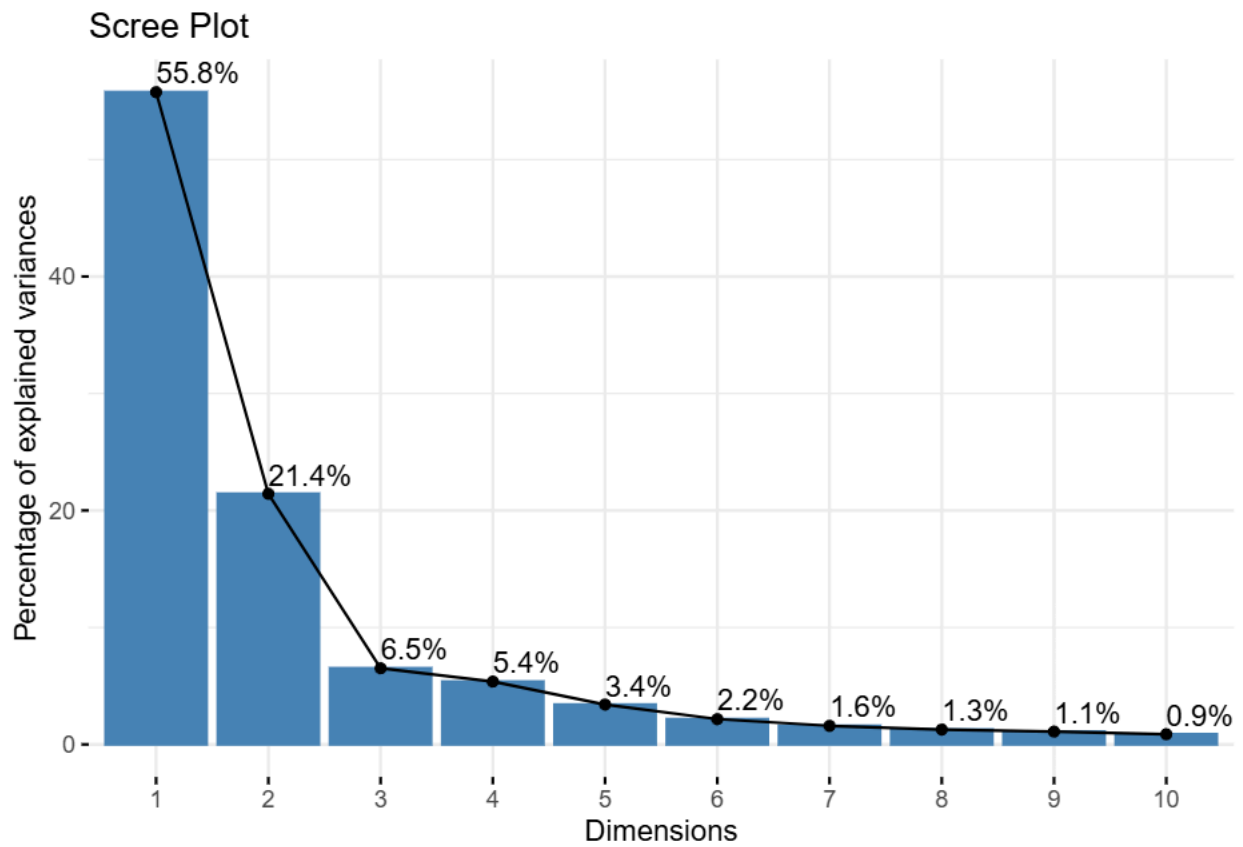
## Importance of components:
##
```

	PC1	PC2	PC3	PC4	PC5	PC6	PC7
## Standard deviation	2.5867	1.6035	0.88398	0.80266	0.63921	0.5091	0.43652
## Proportion of Variance	0.5576	0.2143	0.06512	0.05369	0.03405	0.0216	0.01588
## Cumulative Proportion	0.5576	0.7719	0.83699	0.89067	0.92472	0.9463	0.96220

```
##
```

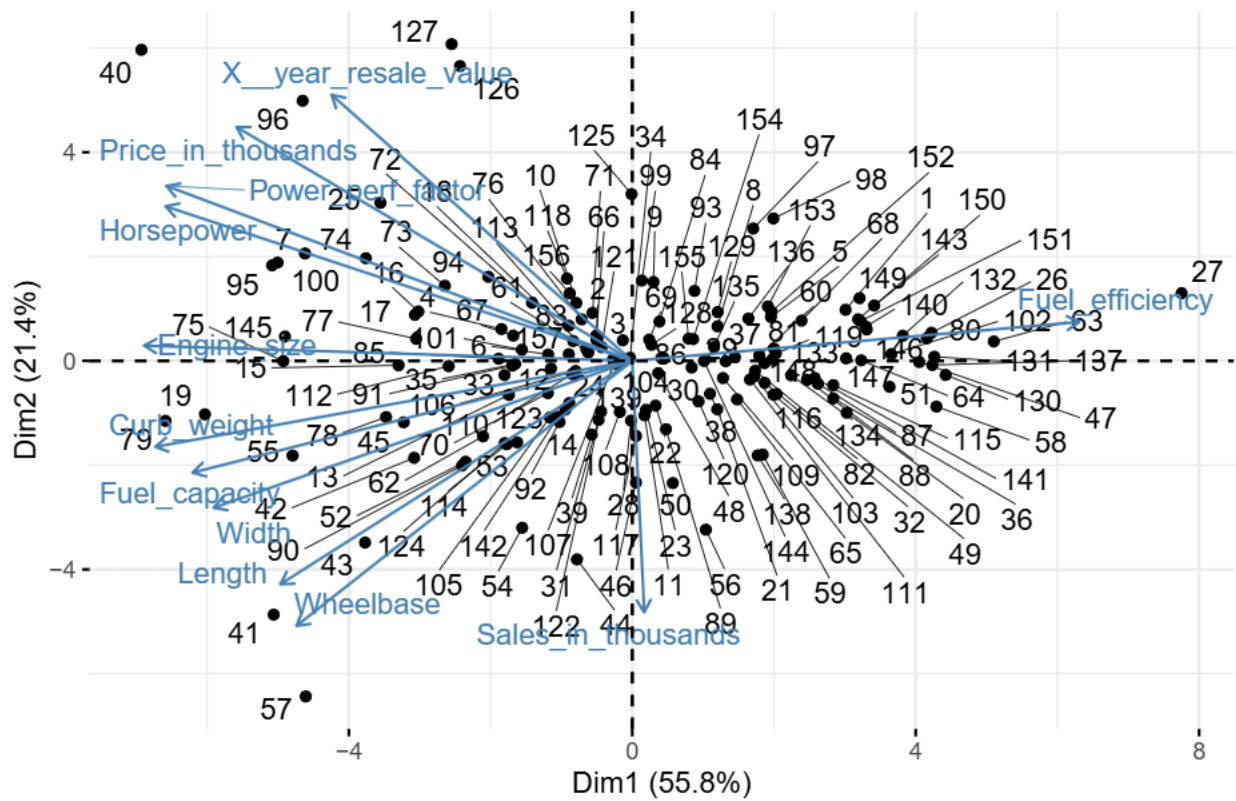
	PC8	PC9	PC10	PC11	PC12
## Standard deviation	0.39003	0.36294	0.32362	0.25277	0.03383
## Proportion of Variance	0.01268	0.01098	0.00873	0.00532	0.00010
## Cumulative Proportion	0.97488	0.98585	0.99458	0.99990	1.00000

```
## Loading required package: ggplot2
##
## Attaching package: 'ggplot2'
## The following objects are masked from 'package:psych':
##
##    %+%, alpha
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```



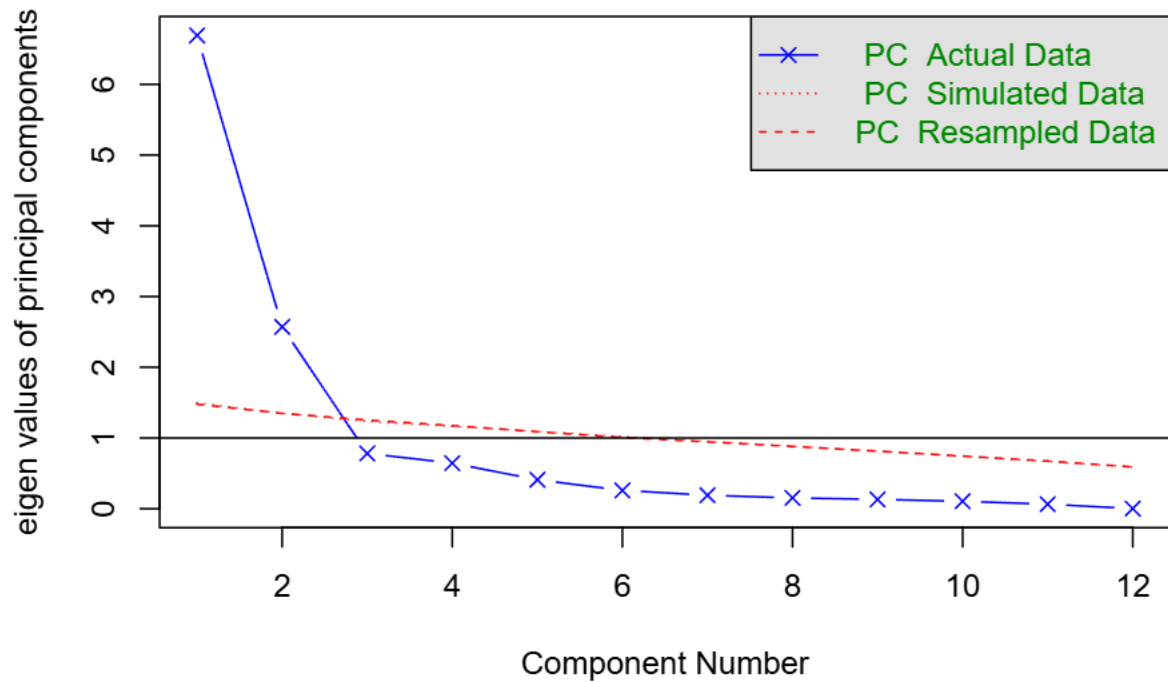
```
# Biplot PCA  
fviz_pca_biplot(pca_result, repel = TRUE)
```

## PCA – Biplot



```
# Parallel Analysis
#fa.parallel(df, fm = "ml", fa = "both")
fa.parallel(df, fa = "pc")
```

## Parallel Analysis Scree Plots



```
## Parallel analysis suggests that the number of factors = NA and the number of components = 2
```

```
#Factor Analysis
```

```
varcov <- cov(df)
```

```
pc <- eigen(varcov)
```

```
pc$values
```

```
## [1] 6.691203830 2.571192763 0.781426944 0.644271014 0.408586245 0.259153970
```

```
## [7] 0.190545442 0.152124933 0.131725757 0.104730296 0.063894102 0.001144704
```

```
pc$vectors
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,]  0.008594942  0.39430485  0.744967314  0.49659316 -0.05791056 -0.03906725
## [2,] -0.215153270 -0.41725016  0.256403822  0.12725271  0.48280190 -0.31930644
## [3,] -0.283191520 -0.36679817  0.121181534  0.01806013  0.24902739 -0.05121043
## [4,] -0.349360433 -0.02387282  0.104837881  0.01435761 -0.47660410  0.20660608
## [5,] -0.333970559 -0.24240653  0.192115126 -0.15518349 -0.22100106  0.21228727
## [6,] -0.240296794  0.41567976  0.059565578 -0.20189642  0.39896923  0.15971796
## [7,] -0.299527557  0.23033279 -0.007239365 -0.27218856 -0.32314576 -0.80702331
## [8,] -0.251818759  0.35034241  0.128237780 -0.48095448  0.23287613  0.25105300
## [9,] -0.341327961  0.13486392 -0.292777115  0.15451925  0.06572817  0.02379266
## [10,] -0.314663595  0.17531663 -0.310570313  0.35037715  0.24910048 -0.11072841
## [11,]  0.319921952 -0.06100794  0.285302120 -0.45429037  0.15884398 -0.17053283
## [12,] -0.333329732 -0.27445477  0.181883371 -0.11576563 -0.12691551  0.15623800
##           [,7]      [,8]      [,9]      [,10]      [,11]
## [1,] -0.11639319  0.01099162 -0.114074540  0.09776134  0.04182092
## [2,]  0.48716925 -0.20959311 -0.009735307 -0.22350018  0.17694475
## [3,] -0.46868117  0.24514882 -0.228094637  0.01244330 -0.58471399
## [4,]  0.07302561 -0.46665744  0.148174828 -0.48952182 -0.33398623
## [5,] -0.03748809  0.10346328  0.197425917  0.27066942  0.42393388
```

```
## [6,] -0.07910795  0.37993197  0.424633385 -0.45306158  0.06545991
## [7,]  0.02573961  0.15644351 -0.006426979  0.03825343 -0.02699552
## [8,]  0.30291337 -0.23724368 -0.374494003  0.34830488 -0.19804493
## [9,] -0.34370379 -0.15814197 -0.561094016 -0.26680529  0.47075855
## [10,] -0.16988410 -0.39451345  0.453643716  0.42173305 -0.09256427
## [11,] -0.50996623 -0.49208122  0.143257130 -0.07304446  0.15403916
## [12,] -0.13031900  0.13956015  0.110408292  0.21217511  0.18878296
##           [,12]
## [1,]  1.185220e-03
## [2,] -1.107114e-03
## [3,]  1.782721e-01
## [4,]  1.032526e-02
## [5,]  6.088813e-01
## [6,]  1.272035e-03
## [7,]  3.277369e-03
## [8,] -1.244756e-03
## [9,] -7.725002e-03
## [10,] -2.609167e-05
## [11,] -6.514083e-03
## [12,] -7.728243e-01
```

```
sp = sum(pc$values[1:3])
```

```
L1 = sqrt(pc$values[1])*pc$vectors[,1]
L2 = sqrt(pc$values[2])*pc$vectors[,2]
L3 = sqrt(pc$values[3])*pc$vectors[,3]
```

```
L = cbind(L1,L2,L3)
L
```

```
##           L1           L2           L3
## [1,]  0.02223285  0.63226543  0.65853886
## [2,] -0.55654474 -0.66905810  0.22665676
## [3,] -0.73254174 -0.58815863  0.10712248
## [4,] -0.90370326 -0.03827991  0.09267496
## [5,] -0.86389371 -0.38869740  0.16982661
## [6,] -0.62158440  0.66653998  0.05265499
## [7,] -0.77479875  0.36933722 -0.00639948
## [8,] -0.65138868  0.56177193  0.11336009
## [9,] -0.88292537  0.21625347 -0.25881016
## [10,] -0.81395169  0.28111915 -0.27453905
## [11,]  0.82755367 -0.09782587  0.25220239
## [12,] -0.86223606 -0.44008656  0.16078191
```

```
#Factor Analysis
```

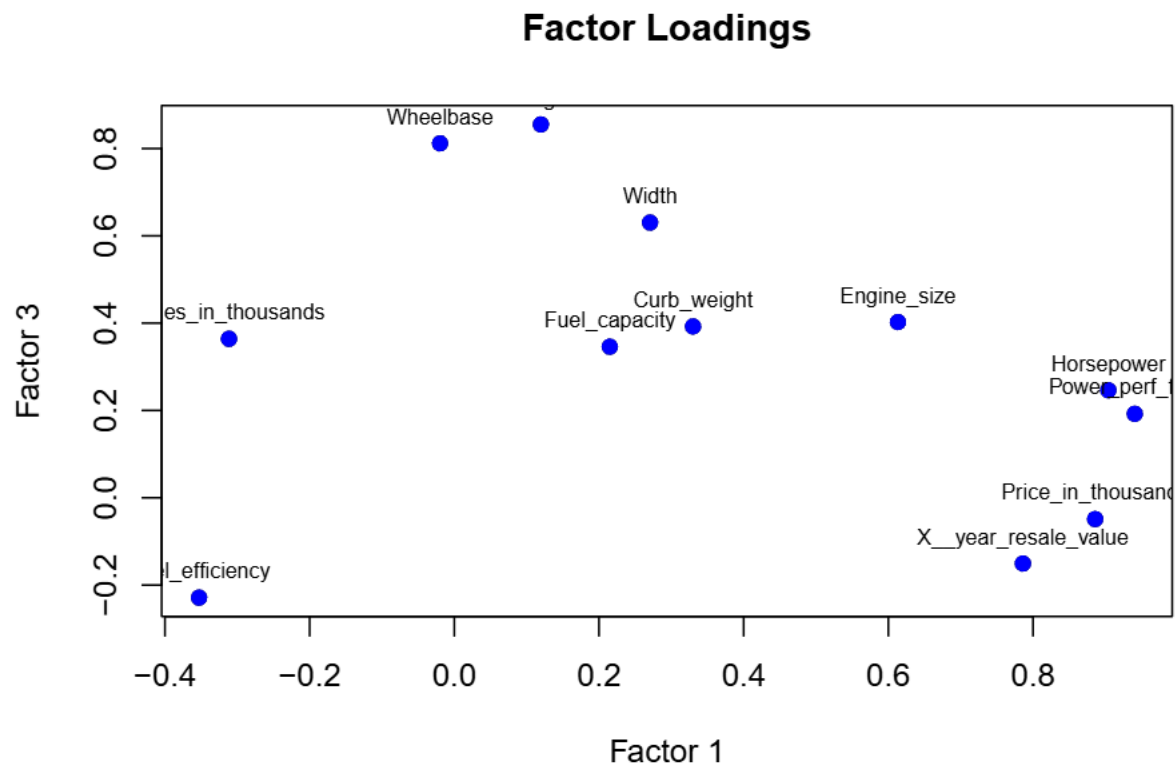
```
library(psych)
fa <- fa(r = cor(df),
         nfactors = 3,
         rotate = "varimax")
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```



```
load <- fa$loadings
plot(load[,c(1,3)], type="p", xlab="Factor 1", ylab="Factor 3", main="Factor Loadings")
points(load[,c(1,3)], pch=19, col="blue")
text(load[,c(1,3)], labels=colnames(df), cex=.7, pos=3)
```



```
fa.diagram(load)
```



## Factor Analysis

