

# **HASIL PRAKTIKUM STATISTIKA**

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**D4 SDT B**

**POLITEKNIK ELEKTRONIKA NEGERI  
SURABAYA**

# Percobaan 1

## 1. Hitung PCA & Standarisasi dataset

### Coding dan Output

```
> #Hitung PCA
> library(DT) #Menampilkan tabel agar mudah dilihat di browser
> library(factoextra) #Visualisasi
> PCAdf <- read.csv("C:/Users/fawza/Documents/Kuliah PENS/Semester 2/Permodelan Statistika Terapan/Dataset/data_pca.csv")
> PCAdf
  X1 X2 X3
1  2  4  1
2  3  3  3
3  5  2  4
4  6  1  7
5  8  0 10
> #Standarisasi dataset
> sdf<- scale(x = PCAdf)
> sdf
      X1      X2      X3
[1,] -1.17279094  1.2649111 -1.1313708
[2,] -0.75393703  0.6324555 -0.5656854
[3,]  0.08377078  0.0000000 -0.2828427
[4,]  0.50262469 -0.6324555  0.5656854
[5,]  1.34033251 -1.2649111  1.4142136
attr(,"scaled:center")
  X1 X2 X3
4.8 2.0 5.0
attr(,"scaled:scale")
      X1      X2      X3
2.387467 1.581139 3.535534
```

### Hasil Analisis Singkat

Kode diatas menampilkan dataset PCAdf lalu menerapkan Standarisasi pada setiap variabel sdf.

Standarisasi dataset PCAdf mengubah setiap variabel sehingga memiliki rata-rata nol dan variansi satu. Hal ini membantu dalam membandingkan data dari kolom yang berbeda.

## 2. Hitung Kovarians

### Coding dan Output

```
> #Hitung matriks kovarians (ragam-peragam)
> kov <- cov(sdf)
> kov
      X1      X2      X3
X1  1.0000000 -0.9933993  0.9773756
X2 -0.9933993  1.0000000 -0.9838699
X3  0.9773756 -0.9838699  1.0000000
```

### Hasil Analisis Singkat

Kode diatas menampilkan hasil dari hitung kovarian antar variable pada dataset.

Hasil matriks kovarians menunjukkan terdapat hubungan yang kuat antara variabel-variabel dalam dataset sdf

- X1 dan X2 memiliki hubungan negatif yang kuat (kovarians -0.9933993).
- X1 dan X3 memiliki hubungan positif yang kuat (kovarians 0.9773756).
- X2 dan X3 memiliki hubungan negatif yang kuat (kovarians -0.9838699).

## 3. Hitung Eigen

### Coding dan Output

```
> #Hitung vektor eigen dan nilai eigen
> vn_eigen <- eigen(kov)
> vn_eigen
eigen() decomposition
$values
[1] 2.969777889 0.024415078 0.005807033

$vectors
      [,1]      [,2]      [,3]
[1,] 0.5775521 -0.5442442 0.6084668
[2,] -0.5788138 0.2526048 0.7753486
[3,] 0.5756806 0.7999932 0.1691236
```

### Hasil Analisis Singkat

Nilai eigen matriks kovariansi mengukur pentingnya setiap komponen utama. Dari kode tersebut didapat nilai eigen dan vektor dari kovarian yang telah dihitung sebelumnya. Ini berguna untuk mengevaluasi berapa banyak variasi dalam data yang dapat dijelaskan.

## 4. PCA Fungsi prcomp

### Coding dan Output

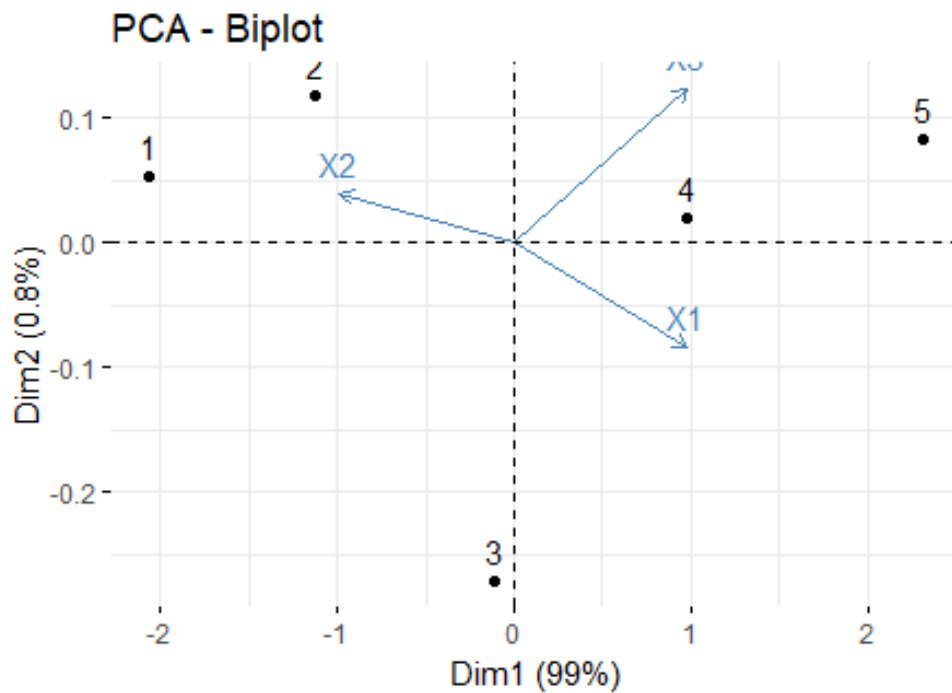
```
> #PCA dengan fungsi prcomp
> pca <- prcomp(x = PCAdf, scale. = TRUE, center = TRUE)
> names(pca)
[1] "sdev"      "rotation"  "center"    "scale"     "x"
> #nilai standar deviasi dari komponen utama
> pca$sdev
[1] 1.72330435 0.15625325 0.07620389
> #nilai data setelah dimasukkan ke dalam persamaan komponen utama (hasil transformasi)
> pca$x
      PC1      PC2      PC3
[1,] -2.0608041 0.05271830 0.075801120
[2,] -1.1271660 0.11754267 -0.064042914
[3,] -0.1144451 -0.27186400 0.003136354
[4,] 0.9820201 0.01923261 -0.088872297
[5,] 2.3203951 0.08237041 0.073977737
> #nilai loadings
> pca$rotation
      PC1      PC2      PC3
X1 0.5775521 -0.5442442 0.6084668
X2 -0.5788138 0.2526048 0.7753486
X3 0.5756806 0.7999932 0.1691236
> #menentukan jumlah komponen utama
> summary(pca)
Importance of components:
              PC1      PC2      PC3
Standard deviation 1.7233 0.15625 0.07620
Proportion of Variance 0.9899 0.00814 0.00194
Cumulative Proportion 0.9899 0.99806 1.00000
> #persamaan komponen utama
> pca$rotation
      PC1      PC2      PC3
X1 0.5775521 -0.5442442 0.6084668
X2 -0.5788138 0.2526048 0.7753486
X3 0.5756806 0.7999932 0.1691236
> #rekonstruksi data
> head(pca$x[,1:2])
      PC1      PC2
[1,] -2.0608041 0.05271830
[2,] -1.1271660 0.11754267
[3,] -0.1144451 -0.27186400
[4,] 0.9820201 0.01923261
[5,] 2.3203951 0.08237041
```

### Hasil Analisis Singkat

Kode diatas menampilkan hasil PCA dengan fungsi prcomp (principal component). Dengan menerapkan beberapa parameter seperti dilakukan standarisasi sehingga memiliki variansi satu dan rata-rata nol sebelum PCA dilakukan. Lalu dilakukan rotasi untuk hasil transformasi dari x1 x3 ke pc1 pc3. Lalu untuk jumlah komponen

## 5. Visualisasi PCA

### Coding dan Output



### Hasil Analisis Singkat

Kode diatas menampilkan biplot dari hasil hitung PCA yang telah dilakukan. Dalam biplot diatas Dim 1 didominasi oleh X1 dan X3 dan Dim 2 didominasi oleh X2.

Ini adalah gambaran menggunakan vektor. Jadi ini semacam vektor dari x1, x2, dan x3 yang dilihat dari 2D pakai biplot.

## Percobaan 2

### 1. Komputasi PCA dan scree plot

### Coding dan Output

```
# 3.2 Percobaan ke-2: Studi Kasus 1
library(factoextra)
data(decathlon2)
decathlon2.active <- decathlon2[1:23, 1:10]
head(decathlon2.active[, 1:6])

> # Komputasi PCA
> res.pca <- prcomp(decathlon2.active, scale = TRUE)
> summary(res.pca)
Importance of components:

            PC1      PC2      PC3      PC4      PC5      PC6      PC7      PC8      PC9
Standard deviation  2.0308  1.3559  1.1132  0.90523  0.83759  0.65029  0.55007  0.52390  0.39398
Proportion of Variance 0.4124  0.1839  0.1239  0.08194  0.07016  0.04229  0.03026  0.02745  0.01552
Cumulative Proportion 0.4124  0.5963  0.7202  0.80213  0.87229  0.91458  0.94483  0.97228  0.98780

            PC10
Standard deviation  0.3492
Proportion of Variance 0.0122
Cumulative Proportion 1.0000

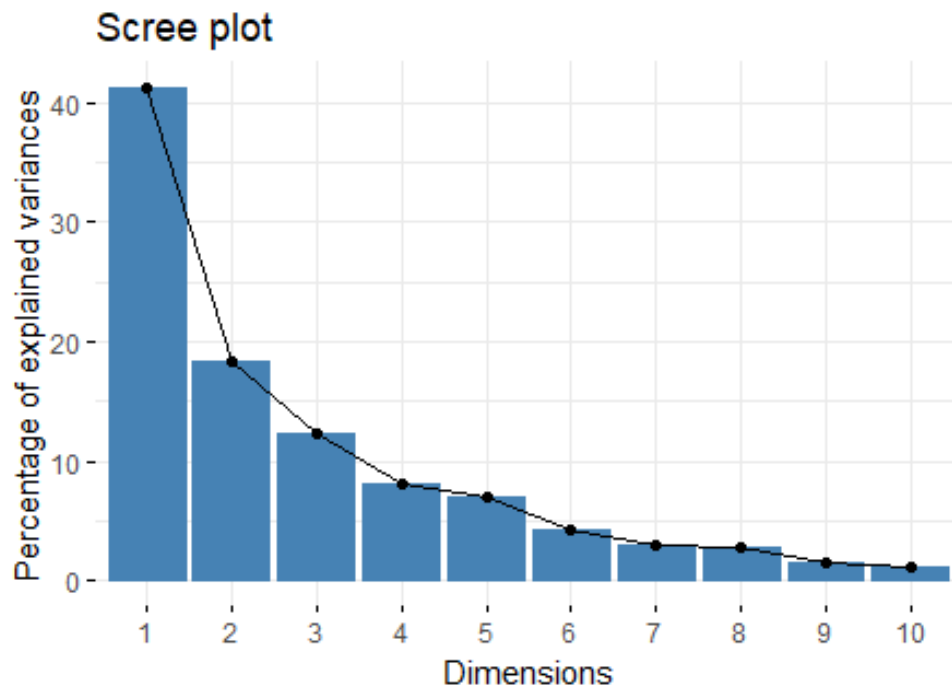
> res.pca
Standard deviations (1, ..., p=10):
[1] 2.0308159 1.3559244 1.1131668 0.9052294 0.8375875 0.6502944 0.5500742 0.5238988
[9] 0.3939758 0.3492435

Rotation (n x k) = (10 x 10):

            PC1      PC2      PC3      PC4      PC5      PC6
X100m      -0.418859080 -0.13230683 -0.27089959  0.03708806 -0.2321476 -0.054398099
Long.jump   0.391064807  0.20713320  0.17117519 -0.12746997  0.2783669  0.051865558
Shot.put    0.361388111  0.06298590 -0.46497777  0.14191803 -0.2970589  0.368739186
High.jump   0.300413236 -0.34309742 -0.29652805  0.15968342  0.4807859  0.437716883
X400m      -0.345478567  0.21400770 -0.25470839  0.47592968  0.1240569  0.075796432
X110m.hurdle -0.376265119 -0.01824645 -0.40325254 -0.01866477  0.2676975 -0.004048005
Discus      0.365965721  0.03662510 -0.15857927  0.43636361 -0.4873988 -0.305315353
Pole.vault  -0.106985591  0.59549862 -0.08449563 -0.37447391 -0.2646712  0.503563524
Javeline    0.210864329  0.28475723 -0.54270782 -0.36646463  0.2361698 -0.556821016
X1500m      0.002106782  0.57855748  0.19715884  0.49491281  0.3142987 -0.064663250

            PC7      PC8      PC9      PC10
X100m      -0.16604375 -0.19988005 -0.76924639  0.12718339
Long.jump   -0.28056361 -0.75850657 -0.13094589  0.08509665
Shot.put    -0.01797323  0.04649571  0.12129309  0.62263702
High.jump   0.05118848  0.16111045 -0.28463225 -0.38244596
X400m       0.52012255 -0.44579641  0.20854176 -0.09784197
X110m.hurdle -0.67276768 -0.01592804  0.41058421 -0.04475363
Discus      -0.25946615 -0.07550934  0.03391600 -0.49418361
Pole.vault  -0.01889413  0.06282691 -0.06540692 -0.39288155
Javeline    0.24281145  0.10086127 -0.10268134 -0.01103627
X1500m      -0.20245828  0.37119711 -0.25950868  0.17991689

> #Visualize eigenvalues (scree plot). Show the percentage of variances explained by each principal component.
> fviz_eig(res.pca)
```



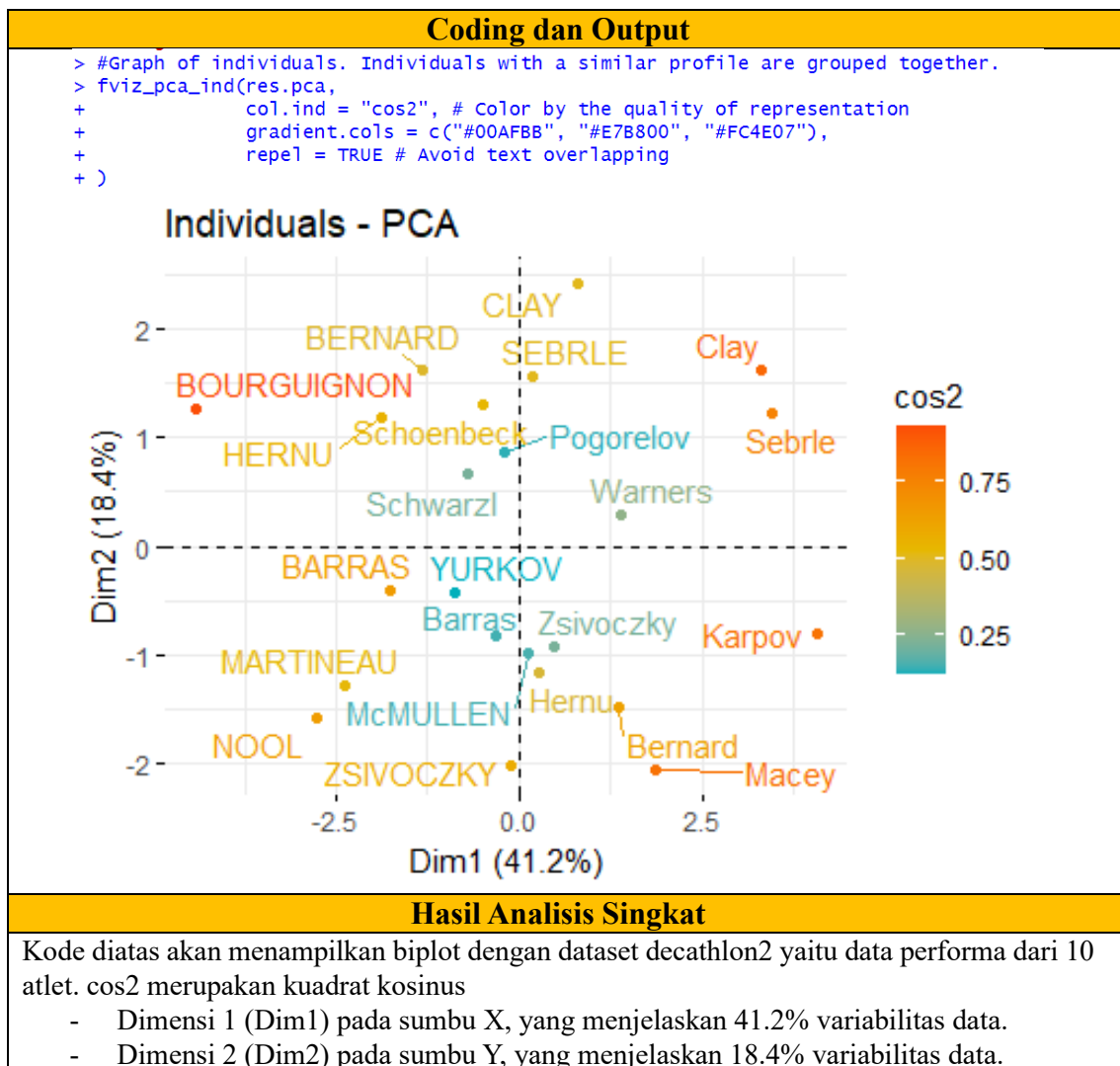
**Hasil Analisis Singkat**

Kode di atas menggunakan dataset decathlon2 dari beberapa atletik dengan mengambil 23 (individu) baris pertama dan 10 (atribut) kolom pertama. Datasetnya kemudian diterapkan PCA parameter standarisasi aktif agar memiliki varians unit sebelum analisis. Kemudian dengan fungsi summary() untuk menampilkan ringkasan statistik dari data tersebut.

Dalam analisis subjektif. Idealnya mengambil komponen utama yang kumulatifnya mendekati 100. Kemudian menampilkan hasil rotasi/hasil transformasi ke PCA.

Disini untuk menentukan jumlah komponen yang akan kita pakai, menggunakan Scree Plot (Visualisasi agent value, merupakan visualisasi dari agent tertinggi ke terendah.). Pada analisis komponen utama (PCA), untuk mengetahui komponen yang mampu menunjukkan perubahan nilai eigen yang besar yaitu dengan melihat scree plot yang dapat menggantikan variabel-variabel yang berpengaruh terhadap dataset decathlon2. Dilihat dari grafik bahwa semakin banyak jumlah komponen yang akan kita pakai, semakin kecil persentase dari total varians dalam data yang dijelaskan oleh setiap komponen utama.

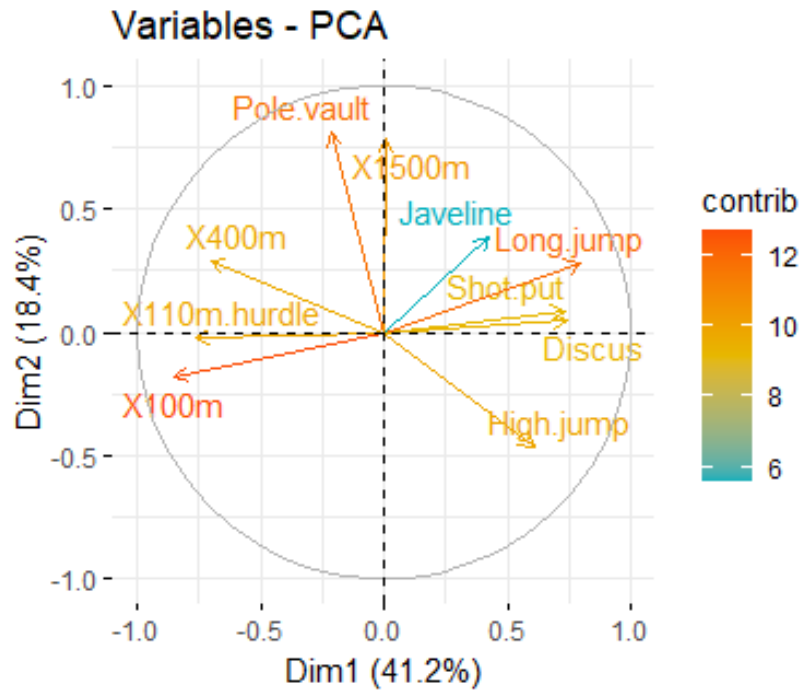
## 2. Individuals – PCA



## 3. Variables – BCA

### Coding dan Output

```
> #Visualize eigenvalues (scree plot). Show the percentage of variances
> #explained by each principal component.
> fviz_eig(res.pca)
```



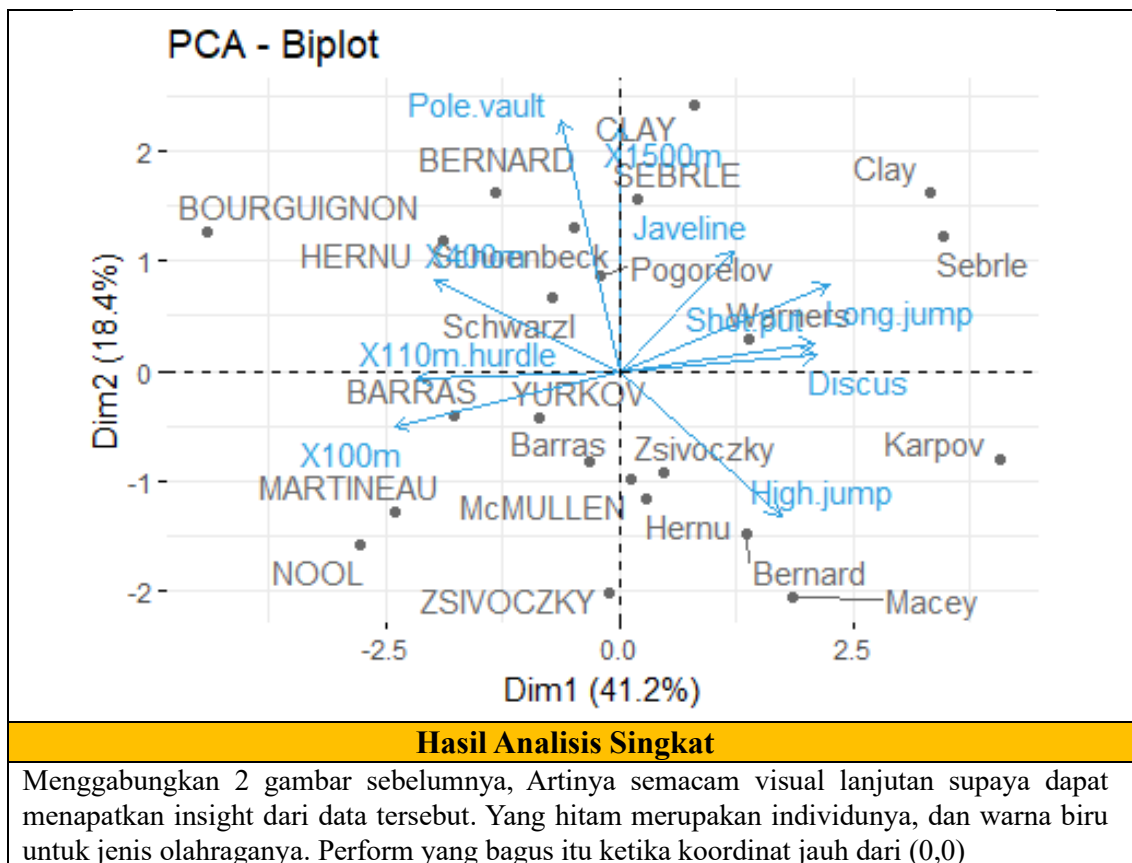
### Hasil Analisis Singkat

Kode diatas akan menampilkan biplot dari 10 atribut data pada dataset decathlon2 dan arah panah menunjukan korelasi dari setiap variable. Di tampilan 2D, arah vektornya memperngaruhi korelasi sehingga semakin panjang vektornya maka dia mempunyai pengaruh yang besar, dan mendominasi tiap jenis olahraga

## 4. PCA – Biplot

### Coding dan Output

```
> #Biplot of individuals and variables
> fviz_pca_biplot(res.pca, repel = TRUE,
+                 col.var = "#2E9FDF", # Variables color
+                 col.ind = "#696969" # Individuals color
+ )
> |
```



## 5. Eigenvalues

Coding dan Output						
<pre> &gt; library(factoextra) &gt; # Eigenvalues &gt; eig.val &lt;- get_eigenvalue(res.pca) &gt; eig.val eigenvalue variance.percent cumulative.variance.percent Dim.1 4.1242133 41.242133 41.24213 Dim.2 1.8385309 18.385309 59.62744 Dim.3 1.2391403 12.391403 72.01885 Dim.4 0.8194402 8.194402 80.21325 Dim.5 0.7015528 7.015528 87.22878 Dim.6 0.4228828 4.228828 91.45760 Dim.7 0.3025817 3.025817 94.48342 Dim.8 0.2744700 2.744700 97.22812 Dim.9 0.1552169 1.552169 98.78029 Dim.10 0.1219710 1.219710 100.00000 &gt; # Results for Variables &gt; res.var &lt;- get_pca_var(res.pca) &gt; res.var\$coord # Coordinates </pre>						
	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
X100m	-0.850625692	-0.17939806	-0.30155643	0.03357320	-0.1944440	-0.035374780
Long.jump	0.794180641	0.28085695	0.19054653	-0.11538956	0.2331567	0.033727883
Shot.put	0.733912733	0.08540412	-0.51759781	0.12846837	-0.2488129	0.239789034
High.jump	0.610083985	-0.46521415	-0.33008517	0.14455012	0.4027002	0.284644846
X400m	-0.701603377	0.29017826	-0.28353292	0.43082552	0.1039085	0.049289996
X110m.hurdle	-0.764125197	-0.02474081	-0.44888733	-0.01689589	0.2242200	-0.002632395
Discus	0.743209016	0.04966086	-0.17652518	0.39500915	-0.4082391	-0.198544870
Pole.vault	-0.217268042	0.80745110	-0.09405773	-0.33898477	-0.2216853	0.327464549
Javeline	0.428226639	0.38610928	-0.60412432	-0.33173454	0.1978128	-0.362097598
X1500m	0.004278487	0.78448019	0.21947068	0.44800961	0.2632527	-0.042050151



```

          Dim.7      Dim.8      Dim.9      Dim.10
X100m      -0.091336386 -0.104716925 -0.30306448  0.044417974
Long.jump  -0.154330810 -0.397380703 -0.05158951  0.029719453
Shot.put   -0.009886612  0.024359049  0.04778655  0.217451948
High.jump   0.028157465  0.084405578 -0.11213822 -0.133566774
X400m      0.286106008 -0.233552216  0.08216041 -0.034170673
X110m.hurdle -0.370072158 -0.008344682  0.16176025 -0.015629914
Discus     -0.142725641 -0.039559255  0.01336209 -0.172590426
Pole.vault -0.010393176  0.032914942 -0.02576874 -0.137211339
Javeline   0.133564318  0.052841099 -0.04045397 -0.003854347
X1500m     -0.111367083  0.194469730 -0.10224014  0.062834809
> res.var$contrib # Contributions to the PCs
          Dim.1      Dim.2      Dim.3      Dim.4      Dim.5      Dim.6
X100m      1.754429e+01  1.7505098  7.3386590  0.13755240  5.389252  0.295915322
Long.jump  1.529317e+01  4.2904162  2.9300944  1.62485936  7.748815  0.269003613
Shot.put   1.306014e+01  0.3967224  21.6204325  2.01407269  8.824401  13.596858744
High.jump  9.024811e+00  11.7715838  8.7928883  2.54987951  23.115504  19.159607001
X400m      1.193554e+01  4.5799296  6.4876363  22.65090599  1.539012  0.574509906
X110m.hurdle 1.415754e+01  0.0332933  16.2612611  0.03483735  7.166193  0.001638634
Discus     1.339309e+01  0.1341398  2.5147385  19.04132022  23.755756  9.321746508
Pole.vault 1.144592e+00  35.4618611  0.7139512  14.02307063  7.005084  25.357622290
Javeline   4.446377e+00  8.1086683  29.4531777  13.42963254  5.577615  31.004964393
X1500m     4.438531e-04  33.4728757  3.8871610  24.49386930  9.878367  0.418133591
          Dim.7      Dim.8      Dim.9      Dim.10
X100m      2.75705260  3.99520353  59.1740009  1.61756139
Long.jump  7.87159392  57.53322220  1.7146826  0.72414393
Shot.put   0.03230371  0.21618512  1.4712015  38.76768578
High.jump  0.26202607  2.59565787  8.1015517  14.62649091
X400m      27.05274658  19.87344405  4.3489667  0.95730504
X110m.hurdle 45.26163460  0.02537025  16.8579392  0.20028870
Discus     6.73226823  0.57016606  0.1150295  24.42174410
Pole.vault 0.03569883  0.39472201  0.4278065  15.43559151
Javeline   5.89573984  1.01729950  1.0543458  0.01217993
X1500m     4.09893563  13.77872941  6.7344755  3.23700871
> res.var$cos2 # Quality of representation
          Dim.1      Dim.2      Dim.3      Dim.4      Dim.5      Dim.6
X100m      7.235641e-01  0.0321836641  0.090936280  0.0011271597  0.03780845  1.251375e-03
Long.jump  6.307229e-01  0.0788806285  0.036307981  0.0133147506  0.05436203  1.137570e-03
Shot.put   5.386279e-01  0.0072938636  0.267907488  0.0165041211  0.06190783  5.749878e-02
High.jump  3.722025e-01  0.2164242070  0.108956221  0.0208947375  0.16216747  8.102269e-02
X400m      4.922473e-01  0.0842034209  0.080390914  0.1856106269  0.01079698  2.429504e-03
X110m.hurdle 5.838873e-01  0.0006121077  0.201499837  0.0002854712  0.05027463  6.929502e-06
Discus     5.523596e-01  0.0024662013  0.031161138  0.1560322304  0.16665918  3.942007e-02
Pole.vault 4.720540e-02  0.6519772763  0.008846856  0.1149106765  0.04914437  1.072330e-01
Javeline   1.833781e-01  0.1490803723  0.364966189  0.1100478063  0.03912992  1.311147e-01
X1500m     1.830545e-05  0.6154091638  0.048167378  0.2007126089  0.06930197  1.768215e-03
          Dim.7      Dim.8      Dim.9      Dim.10
X100m      0.0083423353  1.096563e-02  0.0918480768  1.972956e-03
Long.jump  0.0238179990  1.579114e-01  0.0026614779  8.832459e-04
Shot.put   0.0000977451  5.933633e-04  0.0022835540  4.728535e-02
High.jump  0.0007928428  7.124302e-03  0.0125749811  1.784008e-02
X400m      0.0818566479  5.454664e-02  0.0067503333  1.167635e-03
X110m.hurdle 0.1369534023  6.963371e-05  0.0261663784  2.442942e-04
Discus     0.0203706085  1.564935e-03  0.0001785453  2.978746e-02
Pole.vault 0.0001080181  1.083393e-03  0.0006640282  1.882695e-02
Javeline   0.0178394271  2.792182e-03  0.0016365234  1.485599e-05
X1500m     0.0124026272  3.781848e-02  0.0104530472  3.948213e-03
> # Results for individuals
> res.ind <- get_pca_ind(res.pca)
> res.ind$coord # Coordinates

```

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.1912074	1.5541282	-0.62836882	0.08205241	1.1426139415	0.46389755
CLAY	0.7901217	2.4204156	1.35688701	1.26984296	-0.8068483724	-1.30420016
BERNARD	-1.3292592	1.6118687	-0.19614996	-1.92092203	0.0823428202	0.40062867
YURKOV	-0.8694134	-0.4328779	-2.47398223	0.69723814	0.3988584116	-0.10286344
ZSIVOCZKY	-0.1057450	-2.0233632	1.30493117	-0.09929630	-0.1970241089	-0.89554111
McMULLEN	0.1185550	-0.9916237	0.84355824	1.31215266	1.5858708644	-0.18657283
MARTINEAU	-2.3923532	-1.2849234	-0.89816842	0.37309771	-2.2433515889	0.45666350
HERNU	-1.8910497	1.1784614	-0.15641037	0.89130068	-0.1267412520	-0.43623496
BARRAS	-1.7744575	-0.4125321	0.65817750	0.22872866	-0.2338366980	-0.09026010
NOOL	-2.7770058	-1.5726757	0.60724821	-1.55548081	1.4241839810	-0.49716399
BOURGUIGNON	-4.4137335	1.2635770	-0.01003734	0.66675478	0.4191518468	0.08200220
Sebrle	3.4514485	1.2169193	-1.67816711	-0.80870696	-0.0250530746	0.08279306
Clay	3.3162243	1.6232908	-0.61840443	-0.31679906	0.5691645854	-0.77715960
Karpov	4.0703560	-0.7983510	1.01501662	0.31336354	-0.7974259553	0.32958134
Macey	1.8484623	-2.0638828	-0.97928455	0.58469073	-0.0002157834	0.19728082
Warners	1.3873514	0.2819083	1.99969621	-1.01959817	-0.0405401497	0.55673300
Zsivoczky	0.4715533	-0.9267436	-1.72815525	-0.18483138	0.4073029909	0.11383190
Hernu	0.2763118	-1.1657260	0.17056375	-0.84869401	-0.6894795441	0.33168404
Bernard	1.3672590	-1.4780354	0.83137913	0.74531557	0.8598016482	0.32806564
Schwarzl	-0.7102777	0.6584251	1.04075176	-0.92717510	-0.2887568007	0.68891640
Pogorelov	-0.2143524	0.8610557	0.29761010	1.35560294	-0.0150531057	1.59379599
Schoenbeck	-0.4953166	1.3000530	0.10300360	-0.24927712	-0.6452257128	-0.16172381
Barras	-0.3158867	-0.8193681	-0.86169481	-0.58935985	-0.7797389436	-1.17415412
	Dim.7	Dim.8	Dim.9	Dim.10		
SEBRLE	-0.20796012	0.043460568	-0.659344137	0.03273238		
CLAY	-0.21291866	0.617240611	-0.060125359	-0.31716015		
BERNARD	-0.40643754	0.703856040	0.170083313	-0.09908142		
YURKOV	-0.32487448	0.114996135	-0.109524039	-0.11969720		
ZSIVOCZKY	0.08825624	-0.202341299	-0.523103099	-0.34842265		
McMULLEN	0.47828432	0.293089967	-0.105623196	-0.39317797		
MARTINEAU	-0.29975522	-0.291628488	-0.223417655	-0.61640509		
HERNU	-0.56609980	-1.529404317	0.006184409	0.55368016		
BARRAS	0.21594095	0.682583078	-0.669282042	0.53085420		
NOOL	-0.53205687	-0.433385655	-0.115777808	-0.09622142		
BOURGUIGNON	-0.59833739	0.563619921	0.525814030	0.05855882		
Sebrle	0.01016177	-0.030585843	-0.847210682	0.21970353		
Clay	0.25750851	-0.580638301	0.409776590	-0.61601933		
Karpov	-1.36365568	0.345306381	0.193055107	0.21721852		
Macey	-0.26927772	-0.363219506	0.368260269	0.21249474		
Warners	-0.26739400	-0.109470797	0.180283071	0.24208420		
Zsivoczky	0.03991159	0.538039776	0.585966156	-0.14271715		
Hernu	0.44308686	0.247293566	0.066908586	-0.20868256		
Bernard	0.36357920	0.006165316	0.279488675	0.32067773		
Schwarzl	0.56568604	-0.687053339	-0.008358849	-0.30211546		
Pogorelov	0.78370119	-0.037623661	-0.130531397	-0.03697576		
Schoenbeck	0.85752368	-0.255850722	0.564222295	0.29680481		
Barras	0.94512710	0.365550568	0.102255763	0.61186706		

> res.ind\$contrib # Contributions to the PCs

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.03854254	5.7118249	1.385418e+00	0.03572215	8.091161e+00	2.21256620
CLAY	0.65814114	13.8541889	6.460097e+00	8.55568792	4.034555e+00	17.48801877
BERNARD	1.86273218	6.1441319	1.349983e-01	19.57827284	4.202070e-02	1.65019840
YURKOV	0.79686310	0.4431309	2.147558e+01	2.57939100	9.859373e-01	0.10878629
ZSIVOCZKY	0.01178829	9.6816398	5.974848e+00	0.05231437	2.405750e-01	8.24561722
McMULLEN	0.01481737	2.3253860	2.496789e+00	9.13531719	1.558646e+01	0.35788945
MARTINEAU	6.03367104	3.9044125	2.830527e+00	0.73858431	3.118936e+01	2.14409841
HERNU	3.76996156	3.2842176	8.583863e-02	4.21505626	9.955149e-02	1.95655942
BARRAS	3.31942012	0.4024544	1.519980e+00	0.27758505	3.388731e-01	0.08376135
NOOL	8.12988880	5.8489726	1.293851e+00	12.83761115	1.257025e+01	2.54127369
BOURGUIGNON	20.53729577	3.7757623	3.534995e-04	2.35877858	1.088816e+00	0.06913582
Sebrle	12.55838616	3.5020697	9.881482e+00	3.47006223	3.889859e-03	0.07047579
Clay	11.59361384	6.2315181	1.341828e+00	0.53250375	2.007648e+00	6.20972751
Karpov	17.46609555	1.5072627	3.614914e+00	0.52101693	3.940874e+00	1.11680500
Macey	3.60207087	10.0732890	3.364879e+00	1.81387486	2.885677e-07	0.40014909
Warners	2.02910262	0.1879390	1.403071e+01	5.51585696	1.018550e-02	3.18673563
Zsivoczky	0.23441891	2.0310492	1.047894e+01	0.18126182	1.028128e+00	0.13322327
Hernu	0.08048777	3.2136178	1.020764e-01	3.82170515	2.946148e+00	1.13110069
Bernard	1.97075488	5.1661961	2.425213e+00	2.94737426	4.581507e+00	1.10655655
Schwarzl	0.53184785	1.0252129	3.800546e+00	4.56119277	5.167449e-01	4.87961053
Pogorelov	0.04843819	1.7533304	3.107757e-01	9.75034337	1.404313e-03	26.11665608
Schoenbeck	0.25864068	3.9969003	3.722687e-02	0.32970059	2.580092e+00	0.26890572
Barras	0.10519467	1.5876667	2.605305e+00	1.84296038	3.767994e+00	14.17432302
	Dim.7	Dim.8	Dim.9	Dim.10		
SEBRLE	0.621426384	2.992045e-02	12.177477305	0.03819185		
CLAY	0.651413899	6.035125e+00	0.101262442	3.58568943		
BERNARD	2.373652810	7.847747e+00	0.810319793	0.34994507		
YURKOV	1.516564073	2.094806e-01	0.336009790	0.51072064		
ZSIVOCZKY	0.111923276	6.485544e-01	7.664919832	4.32741147		
McMULLEN	3.287016354	1.360753e+00	0.312501167	5.51053518		
MARTINEAU	1.291109482	1.347216e+00	1.398195851	13.54402896		
HERNU	4.604850849	3.705288e+01	0.001071345	10.92781554		
NOOL	4.067669683	2.975270e+00	0.375477289	0.33003418		
BOURGUIGNON	5.144247534	5.032108e+00	7.744571086	0.12223626		
Sebrle	0.001483775	1.481898e-02	20.105546253	1.72063803		
Clay	0.952824148	5.340583e+00	4.703566841	13.52708188		
Karpov	26.720158115	1.888802e+00	1.043988269	1.68193477		
Macey	1.041910483	2.089853e+00	3.798767930	1.60957713		
Warners	1.027384225	1.898339e-01	0.910422384	2.08904756		
Zsivoczky	0.022889042	4.585705e+00	9.617852173	0.72605208		
Hernu	2.821027418	9.687304e-01	0.125399768	1.55234328		
Bernard	1.899449022	6.021268e-04	2.188071254	3.66566729		
Schwarzl	4.598122119	7.477531e+00	0.001957159	3.25357879		
Pogorelov	8.825322559	2.242329e-02	0.477268755	0.04873597		
Schoenbeck	10.566272800	1.036933e+00	8.917302863	3.14020004		
Barras	12.835417603	2.116763e+00	0.292892746	13.34533825		
> res.ind\$cos2 # Quality of representation						
	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.007530179	0.49747323	8.132523e-02	0.001386688	2.689027e-01	0.0443241299
CLAY	0.048701249	0.45701660	1.436281e-01	0.125791741	5.078506e-02	0.1326907339
BERNARD	0.197199804	0.28996555	4.294015e-03	0.411819183	7.567259e-04	0.0179131165
YURKOV	0.096109800	0.02382571	7.782303e-01	0.061812637	2.022798e-02	0.0013453555
ZSIVOCZKY	0.001574385	0.57641944	2.397542e-01	0.001388216	5.465497e-03	0.1129176906
McMULLEN	0.002175437	0.15219499	1.101379e-01	0.266486530	3.892621e-01	0.0053876990
MARTINEAU	0.404013915	0.11654676	5.694575e-02	0.009826320	3.552552e-01	0.0147210347
HERNU	0.399282749	0.15506199	2.731529e-03	0.088699901	1.793538e-03	0.0212478795
BARRAS	0.616241975	0.03330700	8.478249e-02	0.010239088	1.070152e-02	0.0015944528
NOOL	0.489872515	0.15711146	2.342405e-02	0.153694675	1.288433e-01	0.0157010551
BOURGUIGNON	0.859698130	0.07045912	4.446015e-06	0.019618511	7.753120e-03	0.0002967459
Sebrle	0.675380606	0.08395940	1.596674e-01	0.037079012	3.558507e-05	0.0003886276
Clay	0.687592867	0.16475409	2.391051e-02	0.006274965	2.025440e-02	0.0377627839
Karpov	0.783666922	0.03014772	4.873187e-02	0.004644764	3.007790e-02	0.0051379747
Macey	0.363436037	0.45308203	1.020057e-01	0.036362957	4.952707e-09	0.0041397727
Warners	0.255651956	0.01055582	5.311341e-01	0.138081100	2.182965e-04	0.0411689767
Zsivoczky	0.045053176	0.17401353	6.051030e-01	0.006921739	3.361236e-02	0.0026253777

Hernu	0.024824321	0.44184663	9.459148e-03	0.234196727	1.545686e-01	0.0357707217
Bernard	0.289347476	0.33813318	1.069834e-01	0.085980212	1.144234e-01	0.0166586433
Schwarzl	0.116721435	0.10030142	2.506043e-01	0.198892209	1.929118e-02	0.1098063093
Pogorelov	0.007803472	0.12591966	1.504272e-02	0.312101619	3.848427e-05	0.4314162233
Schoenbeck	0.067070098	0.46204603	2.900467e-03	0.016987442	1.138116e-01	0.0071500829
Barras	0.018972684	0.12765099	1.411800e-01	0.066043061	1.156018e-01	0.2621297474

	Dim.7	Dim.8	Dim.9	Dim.10
SEBRLE	8.907507e-03	3.890334e-04	8.954067e-02	0.0002206741
CLAY	3.536548e-03	2.972084e-02	2.820119e-04	0.0078471026
BERNARD	1.843634e-02	5.529104e-02	3.228572e-03	0.0010956493
YURKOV	1.341980e-02	1.681440e-03	1.525225e-03	0.0018217256
ZSIVOCZKY	1.096685e-03	5.764478e-03	3.852703e-02	0.0170924251
McMULLEN	3.540616e-02	1.329562e-02	1.726733e-03	0.0239268142
MARTINEAU	6.342774e-03	6.003515e-03	3.523552e-03	0.0268211980
HERNU	3.578167e-02	2.611676e-01	4.270425e-06	0.0342288717
BARRAS	9.126203e-03	9.118662e-02	8.766746e-02	0.0551531863
NOOL	1.798232e-02	1.193105e-02	8.514912e-04	0.0005881295
BOURGUIGNON	1.579887e-02	1.401866e-02	1.220108e-02	0.0001513277
Sebrle	5.854423e-06	5.303795e-05	4.069384e-02	0.0027366539
Clay	4.145976e-03	2.107924e-02	1.049876e-02	0.0237264222
Karpov	8.795817e-02	5.639959e-03	1.762907e-03	0.0022318265
Macey	7.712721e-03	1.403282e-02	1.442502e-02	0.0048028954
Warners	9.496848e-03	1.591742e-03	4.317040e-03	0.0077841113
Zsivoczky	3.227467e-04	5.865332e-02	6.956790e-02	0.0041268259
Hernu	6.383462e-02	1.988402e-02	1.455601e-03	0.0141595965
Bernard	2.046050e-02	5.883405e-06	1.209056e-02	0.0159167991
Schwarzl	7.403638e-02	1.092132e-01	1.616543e-05	0.0211173850
Pogorelov	1.043115e-01	2.404103e-04	2.893750e-03	0.0002322016
Schoenbeck	2.010275e-01	1.789520e-02	8.702893e-02	0.0240826922
Barras	1.698426e-01	2.540745e-02	1.988116e-03	0.0711836486

> res.var\$contrib # Contributions to the PCs

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
X100m	1.754429e+01	1.7505098	7.3386590	0.13755240	5.389252	0.295915322
Long.jump	1.529317e+01	4.2904162	2.9300944	1.62485936	7.748815	0.269003613
Shot.put	1.306014e+01	0.3967224	21.6204325	2.01407269	8.824401	13.596858744
High.jump	9.024811e+00	11.7715838	8.7928883	2.54987951	23.115504	19.159607001
X400m	1.193554e+01	4.5799296	6.4876363	22.65090599	1.539012	0.574509906
X110m.hurdle	1.415754e+01	0.0332933	16.2612611	0.03483735	7.166193	0.001638634
Discus	1.339309e+01	0.1341398	2.5147385	19.04132022	23.755756	9.321746508
Pole.vault	1.144592e+00	35.4618611	0.7139512	14.02307063	7.005084	25.357622290
Javeline	4.446377e+00	8.1086683	29.4531777	13.42963254	5.577615	31.004964393
X1500m	4.438531e-04	33.4728757	3.8871610	24.49386930	9.878367	0.418133591

	Dim.7	Dim.8	Dim.9	Dim.10
X100m	2.75705260	3.99520353	59.1740009	1.61756139
Long.jump	7.87159392	57.53322220	1.7146826	0.72414393
Shot.put	0.03230371	0.21618512	1.4712015	38.76768578
High.jump	0.26202607	2.59565787	8.1015517	14.62649091
X400m	27.05274658	19.87344405	4.3489667	0.95730504
X110m.hurdle	45.26163460	0.02537025	16.8579392	0.20028870
Discus	6.73226823	0.57016606	0.1150295	24.42174410
Pole.vault	0.03569883	0.39472201	0.4278065	15.43559151
Javeline	5.89573984	1.01729950	1.0543458	0.01217993
X1500m	4.09893563	13.77872941	6.7344755	3.23700871

>

> res.var\$cos2 # Quality of representation

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
X100m	7.235641e-01	0.0321836641	0.090936280	0.0011271597	0.03780845	1.251375e-03
Long.jump	6.307229e-01	0.0788806285	0.036307981	0.0133147506	0.05436203	1.137570e-03
Shot.put	5.386279e-01	0.0072938636	0.267907488	0.0165041211	0.06190783	5.749878e-02
High.jump	3.722025e-01	0.2164242070	0.108956221	0.0208947375	0.16216747	8.102269e-02
X400m	4.922473e-01	0.0842034209	0.080390914	0.1856106269	0.01079698	2.429504e-03
X110m.hurdle	5.838873e-01	0.0006121077	0.201499837	0.0002854712	0.05027463	6.929502e-06
Discus	5.523596e-01	0.0024662013	0.031161138	0.1560322304	0.16665918	3.942007e-02
Pole.vault	4.720540e-02	0.6519772763	0.008846856	0.1149106765	0.04914437	1.072330e-01
Javeline	1.833781e-01	0.1490803723	0.364966189	0.1100478063	0.03912992	1.311147e-01
X1500m	1.830545e-05	0.6154091638	0.048167378	0.2007126089	0.06930197	1.768215e-03

	Dim.7	Dim.8	Dim.9	Dim.10
X100m	0.0083423353	1.096563e-02	0.0918480768	1.972956e-03
Long.jump	0.0238179990	1.579114e-01	0.0026614779	8.832459e-04
Shot.put	0.0000977451	5.933633e-04	0.0022835540	4.728535e-02
High.jump	0.0007928428	7.124302e-03	0.0125749811	1.784008e-02
X400m	0.0818566479	5.454664e-02	0.0067503333	1.167635e-03
X110m.hurdle	0.1369534023	6.963371e-05	0.0261663784	2.442942e-04
Discus	0.0203706085	1.564935e-03	0.0001785453	2.978746e-02
Pole.vault	0.0001080181	1.083393e-03	0.0006640282	1.882695e-02
Javeline	0.0178394271	2.792182e-03	0.0016365234	1.485599e-05
X1500m	0.0124026272	3.781848e-02	0.0104530472	3.948213e-03

>

> # Results for individuals

> res.ind <- get\_pca\_ind(res.pca)

> res.ind\$coord # Coordinates



	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.1912074	1.5541282	-0.62836882	0.08205241	1.1426139415	0.46389755
CLAY	0.7901217	2.4204156	1.35688701	1.26984296	-0.8068483724	-1.30420016
BERNARD	-1.3292592	1.6118687	-0.19614996	-1.92092203	0.0823428202	0.40062867
YURKOV	-0.8694134	-0.4328779	-2.47398223	0.69723814	0.3988584116	-0.10286344
ZSIVOCZKY	-0.1057450	-2.0233632	1.30493117	-0.09929630	-0.1970241089	-0.89554111
McMULLEN	0.1185550	-0.9916237	0.84355824	1.31215266	1.5858708644	-0.18657283
MARTINEAU	-2.3923532	-1.2849234	-0.89816842	0.37309771	-2.2433515889	0.45666350
HERNU	-1.8910497	1.1784614	-0.15641037	0.89130068	-0.1267412520	-0.43623496
BARRAS	-1.7744575	-0.4125321	0.65817750	0.22872866	-0.2338366980	-0.09026010
NOOL	-2.7770058	-1.5726757	0.60724821	-1.55548081	1.4241839810	-0.49716399
BOURGUIGNON	-4.4137335	1.2635770	-0.01003734	0.66675478	0.4191518468	0.08200220
Sebrle	3.4514485	1.2169193	-1.67816711	-0.80870696	-0.0250530746	0.08279306
Clay	3.3162243	1.6232908	-0.61840443	-0.31679906	0.5691645854	-0.77715960
Karpov	4.0703560	-0.7983510	1.01501662	0.31336354	-0.7974259553	0.32958134
Macey	1.8484623	-2.0638828	-0.97928455	0.58469073	-0.0002157834	0.19728082
Warners	1.3873514	0.2819083	1.99969621	-1.01959817	-0.0405401497	0.55673300
Zsivoczky	0.4715533	-0.9267436	-1.72815525	-0.18483138	0.4073029909	0.11383190
Hernu	0.2763118	-1.1657260	0.17056375	-0.84869401	-0.6894795441	0.33168404
Bernard	1.3672590	-1.4780354	0.83137913	0.74531557	0.8598016482	0.32806564
Schwarzl	-0.7102777	0.6584251	1.04075176	-0.92717510	-0.2887568007	0.68891640
Pogorelov	-0.2143524	0.8610557	0.29761010	1.35560294	-0.0150531057	1.59379599
Schoenbeck	-0.4953166	1.3000530	0.10300360	-0.24927712	-0.6452257128	-0.16172381
Barras	-0.3158867	-0.8193681	-0.86169481	-0.58935985	-0.7797389436	-1.17415412
	Dim.7	Dim.8	Dim.9	Dim.10		
SEBRLE	-0.20796012	0.043460568	-0.659344137	0.03273238		
CLAY	-0.21291866	0.617240611	-0.060125359	-0.31716015		
BERNARD	-0.40643754	0.703856040	0.170083313	-0.09908142		
YURKOV	-0.32487448	0.114996135	-0.109524039	-0.11969720		
ZSIVOCZKY	0.08825624	-0.202341299	-0.523103099	-0.34842265		
McMULLEN	0.47828432	0.293089967	-0.105623196	-0.39317797		
MARTINEAU	-0.29975522	-0.291628488	-0.223417655	-0.61640509		
HERNU	-0.56609980	-1.529404317	0.006184409	0.55368016		
BARRAS	0.21594095	0.682583078	-0.669282042	0.53085420		
NOOL	-0.53205687	-0.433385655	-0.115777808	-0.09622142		
BOURGUIGNON	-0.59833739	0.563619921	0.525814030	0.05855882		
Sebrle	0.01016177	-0.030585843	-0.847210682	0.21970353		
Clay	0.25750851	-0.580638301	0.409776590	-0.61601933		
Karpov	-1.36365568	0.345306381	0.193055107	0.21721852		
Macey	-0.26927772	-0.363219506	0.368260269	0.21249474		
Warners	-0.26739400	-0.109470797	0.180283071	0.24208420		
Zsivoczky	0.03991159	0.538039776	0.585966156	-0.14271715		
Hernu	0.44308686	0.247293566	0.066908586	-0.20868256		
Bernard	0.36357920	0.006165316	0.279488675	0.32067773		
Schwarzl	0.56568604	-0.687053339	-0.008358849	-0.30211546		
Pogorelov	0.78370119	-0.037623661	-0.130531397	-0.03697576		
Schoenbeck	0.85752368	-0.255850722	0.564222295	0.29680481		
Barras	0.94512710	0.365550568	0.102255763	0.61186706		
>						
> res.ind\$contrib # Contributions to the PCs						
	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.03854254	5.7118249	1.385418e+00	0.03572215	8.091161e+00	2.21256620
CLAY	0.65814114	13.8541889	6.460097e+00	8.55568792	4.034555e+00	17.48801877
BERNARD	1.86273218	6.1441319	1.349983e-01	19.57827284	4.202070e-02	1.65019840
YURKOV	0.79686310	0.4431309	2.147558e+01	2.57939100	9.859373e-01	0.10878629
ZSIVOCZKY	0.01178829	9.6816398	5.974848e+00	0.05231437	2.405750e-01	8.24561722
McMULLEN	0.01481737	2.3253860	2.496789e+00	9.13531719	1.558646e+01	0.35788945
MARTINEAU	6.03367104	3.9044125	2.830527e+00	0.73858431	3.118936e+01	2.14409841
HERNU	3.76996156	3.2842176	8.583863e-02	4.21505626	9.955149e-02	1.95655942
BARRAS	3.31942012	0.4024544	1.519980e+00	0.27758505	3.388731e-01	0.08376135
NOOL	8.12988880	5.8489726	1.293851e+00	12.83761115	1.257025e+01	2.54127369
BOURGUIGNON	20.53729577	3.7757623	3.534995e-04	2.35877858	1.088816e+00	0.06913582
Sebrle	12.55838616	3.5020697	9.881482e+00	3.47006223	3.889859e-03	0.07047579
Clay	11.59361384	6.2315181	1.341828e+00	0.53250375	2.007648e+00	6.20972751
Karpov	17.46609555	1.5072627	3.614914e+00	0.52101693	3.940874e+00	1.11680500
Macey	3.60207087	10.0732890	3.364879e+00	1.81387486	2.885677e-07	0.40014909
Warners	2.02910262	0.1879390	1.403071e+01	5.51585696	1.018550e-02	3.18673563
Zsivoczky	0.23441891	2.0310492	1.047894e+01	0.18126182	1.028128e+00	0.13322327
Hernu	0.08048777	3.2136178	1.020764e-01	3.82170515	2.946148e+00	1.13110069
Bernard	1.97075488	5.1661961	2.425213e+00	2.94737426	4.581507e+00	1.10655655
Schwarzl	0.53184785	1.0252129	3.800546e+00	4.56119277	5.167449e-01	4.87961053
Pogorelov	0.04843819	1.7533304	3.107757e-01	9.75034337	1.404313e-03	26.11665608
Schoenbeck	0.25864068	3.9969003	3.722687e-02	0.32970059	2.580092e+00	0.26890572
Barras	0.10519467	1.5876667	2.605305e+00	1.84296038	3.767994e+00	14.17432302

	Dim.7	Dim.8	Dim.9	Dim.10		
SEBRLE	0.621426384	2.992045e-02	12.177477305	0.03819185		
CLAY	0.651413899	6.035125e+00	0.101262442	3.58568943		
BERNARD	2.373652810	7.847747e+00	0.810319793	0.34994507		
YURKOV	1.516564073	2.094806e-01	0.336009790	0.51072064		
ZSIVOCZKY	0.111923276	6.485544e-01	7.664919832	4.32741147		
McMULLEN	3.287016354	1.360753e+00	0.312501167	5.51053518		
MARTINEAU	1.291109482	1.347216e+00	1.398195851	13.54402896		
HERNU	4.604850849	3.705288e+01	0.001071345	10.92781554		
BARRAS	0.670038259	7.380544e+00	12.547331617	10.04537028		
NOOL	4.067669683	2.975270e+00	0.375477289	0.33003418		
BOURGUIGNON	5.144247534	5.032108e+00	7.744571086	0.12223626		
Sebrle	0.001483775	1.481898e-02	20.105546253	1.72063803		
Clay	0.952824148	5.340583e+00	4.703566841	13.52708188		
Karpov	26.720158115	1.888802e+00	1.043988269	1.68193477		
Macey	1.041910483	2.089853e+00	3.798767930	1.60957713		
Warners	1.027384225	1.898339e-01	0.910422384	2.08904756		
Zsivoczky	0.022889042	4.585705e+00	9.617852173	0.72605208		
Hernu	2.821027418	9.687304e-01	0.125399768	1.55234328		
Bernard	1.899449022	6.021268e-04	2.188071254	3.66566729		
Schwarzl	4.598122119	7.477531e+00	0.001957159	3.25357879		
Pogorelov	8.825322559	2.242329e-02	0.477268755	0.04873597		
Schoenbeck	10.566272800	1.036933e+00	8.917302863	3.14020004		
Barras	12.835417603	2.116763e+00	0.292892746	13.34533825		
>						
> res.ind\$cos2 # Quality of representation						
	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
SEBRLE	0.007530179	0.49747323	8.132523e-02	0.001386688	2.689027e-01	0.0443241299
CLAY	0.048701249	0.45701660	1.436281e-01	0.125791741	5.078506e-02	0.1326907339
BERNARD	0.197199804	0.28996555	4.294015e-03	0.411819183	7.567259e-04	0.0179131165
YURKOV	0.096109800	0.02382571	7.782303e-01	0.061812637	2.022798e-02	0.0013453555
ZSIVOCZKY	0.001574385	0.57641944	2.397542e-01	0.001388216	5.465497e-03	0.1129176906
McMULLEN	0.002175437	0.15219499	1.101379e-01	0.266486530	3.892621e-01	0.0053876990
MARTINEAU	0.404013915	0.11654676	5.694575e-02	0.009826320	3.552552e-01	0.0147210347
HERNU	0.399282749	0.15506199	2.731529e-03	0.088699901	1.793538e-03	0.0212478795
BARRAS	0.616241975	0.03330700	8.478249e-02	0.010239088	1.070152e-02	0.0015944528
NOOL	0.489872515	0.15711146	2.342405e-02	0.153694675	1.288433e-01	0.0157010551
BOURGUIGNON	0.859698130	0.07045912	4.446015e-06	0.019618511	7.753120e-03	0.0002967459
Sebrle	0.675380606	0.08395940	1.596674e-01	0.037079012	3.558507e-05	0.0003886276
Clay	0.687592867	0.16475409	2.391051e-02	0.006274965	2.025440e-02	0.0377627839
Karpov	0.783666922	0.03014772	4.873187e-02	0.004644764	3.007790e-02	0.0051379747
Macey	0.363436037	0.45308203	1.020057e-01	0.036362957	4.952707e-09	0.0041397727
Warners	0.255651956	0.01055582	5.311341e-01	0.138081100	2.182965e-04	0.0411689767
Zsivoczky	0.045053176	0.17401353	6.051030e-01	0.006921739	3.361236e-02	0.0026253777
Hernu	0.024824321	0.44184663	9.459148e-03	0.234196727	1.545686e-01	0.0357707217
Bernard	0.289347476	0.33813318	1.069834e-01	0.085980212	1.144234e-01	0.0166586433
Schwarzl	0.116721435	0.10030142	2.506043e-01	0.198892209	1.929118e-02	0.1098063093
Pogorelov	0.007803472	0.12591966	1.504272e-02	0.312101619	3.848427e-05	0.4314162233
Schoenbeck	0.067070098	0.46204603	2.900467e-03	0.016987442	1.138116e-01	0.0071500829
Barras	0.018972684	0.12765099	1.411800e-01	0.066043061	1.156018e-01	0.2621297474
	Dim.7	Dim.8	Dim.9	Dim.10		
SEBRLE	8.907507e-03	3.890334e-04	8.954067e-02	0.0002206741		
CLAY	3.536548e-03	2.972084e-02	2.820119e-04	0.0078471026		
BERNARD	1.843634e-02	5.529104e-02	3.228572e-03	0.0010956493		
YURKOV	1.341980e-02	1.681440e-03	1.525225e-03	0.0018217256		
ZSIVOCZKY	1.096685e-03	5.764478e-03	3.852703e-02	0.0170924251		
McMULLEN	3.540616e-02	1.329562e-02	1.726733e-03	0.0239268142		
MARTINEAU	6.342774e-03	6.003515e-03	3.523552e-03	0.0268211980		
HERNU	3.578167e-02	2.611676e-01	4.270425e-06	0.0342288717		
BARRAS	9.126203e-03	9.118662e-02	8.766746e-02	0.0551531863		
NOOL	1.798232e-02	1.193105e-02	8.514912e-04	0.0005881295		
BOURGUIGNON	1.579887e-02	1.401866e-02	1.220108e-02	0.0001513277		
Sebrle	5.854423e-06	5.303795e-05	4.069384e-02	0.0027366539		
Clay	4.145976e-03	2.107924e-02	1.049876e-02	0.0237264222		
Karpov	8.795817e-02	5.639959e-03	1.762907e-03	0.0022318265		
Macey	7.712721e-03	1.403282e-02	1.442502e-02	0.0048028954		
Warners	9.496848e-03	1.591742e-03	4.317040e-03	0.0077841113		
Zsivoczky	3.227467e-04	5.865332e-02	6.956790e-02	0.0041268259		
Hernu	6.383462e-02	1.988402e-02	1.455601e-03	0.0141595965		
Bernard	2.046050e-02	5.883405e-06	1.209056e-02	0.0159167991		
Schwarzl	7.403638e-02	1.092132e-01	1.616543e-05	0.0211173850		
Pogorelov	1.043115e-01	2.404103e-04	2.893750e-03	0.0002322016		
Schoenbeck	2.010275e-01	1.789520e-02	8.702893e-02	0.0240826922		
Barras	1.698426e-01	2.540745e-02	1.988116e-03	0.0711836486		
>						

### Hasil Analisis Singkat

Di atas adalah hasil dari tiap variabel jika ingin melihat kontribusi, koordinat, dari masing-masing personal atau individu yang mendominasi dari tiap tiap dimensi. Dimensi 1 nanti merupakan visual vektor. Untuk insightnya yang bisa ditarik dari

informasi di atas untuk pengambilan keputusan dari tiap komponen utama yang diambil

KARPOV memiliki skor PC1 dan PC3 yang tinggi, menunjukkan performa yang baik pada "kecepatan keseluruhan" dan "kekuatan keseluruhan".

WARNERS memiliki skor PC1 dan PC3 yang positif, menunjukkan performa yang baik pada "kecepatan keseluruhan" dan "kekuatan keseluruhan".

Nool memiliki skor PC1 yang negatif dan skor PC4 yang sangat negatif, menunjukkan performa yang kurang baik pada "kecepatan keseluruhan" dan "kemampuan atletik secara keseluruhan".

Drews memiliki skor PC1 dan PC3 yang positif, menunjukkan performa yang baik pada "kecepatan keseluruhan" dan "kekuatan keseluruhan".

## 6. Prediksi menggunakan PCA

### Coding dan Output

```
> # Data for the supplementary individuals
> ind.sup <- decathlon2[24:27, 1:10]
> ind.sup[, 1:6]
      X100m Long.jump Shot.put High.jump X400m X110m.hurdle
KARPOV  11.02    7.30    14.77    2.04 48.37    14.09
WARNERS  11.11    7.60    14.31    1.98 48.68    14.23
Nool    10.80    7.53    14.26    1.88 48.81    14.80
Drews   10.87    7.38    13.07    1.88 48.51    14.01

> ind.sup.coord <- predict(res.pca, newdata = ind.sup)
> ind.sup.coord[, 1:4]
      PC1      PC2      PC3      PC4
KARPOV  0.7772521  0.76237804 1.5971253 1.6863286
WARNERS -0.3779697 -0.11891968 1.7005146 -0.6908084
Nool    -0.5468405  1.93402211 0.4724184 -2.2283706
Drews   -1.0848227  0.01703198 2.9818031 -1.5006207
```

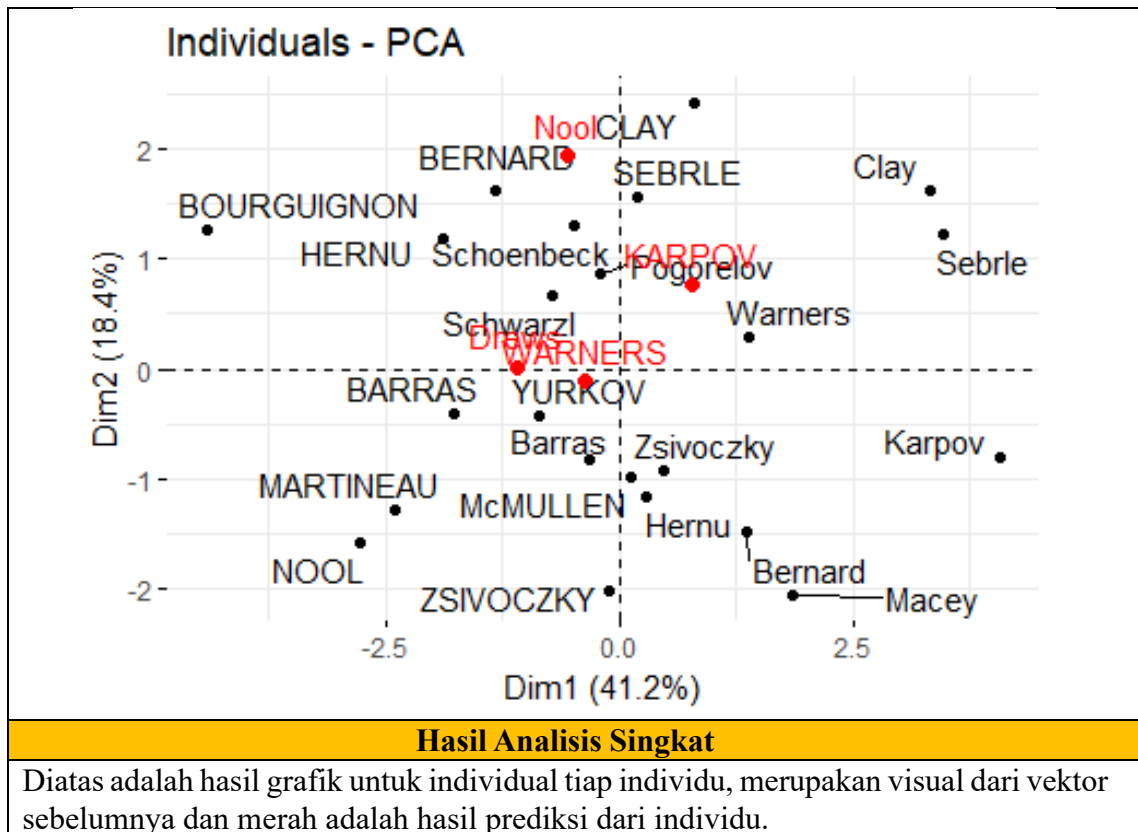
### Hasil Analisis Singkat

Kode di atas merupakan kode untuk memprediksi menggunakan PCA. Hasil prediksi menunjukkan bahwa KARPOV dan WARNERS memiliki performa yang lebih baik dibandingkan dengan Nool dan Drews.

## 7. Individuals PCA

### Coding dan Output

```
> # Plot of active individuals
> p <- fviz_pca_ind(res.pca, repel = TRUE)
> # Add supplementary individuals
> fviz_add(p, ind.sup.coord, color = "red")
```



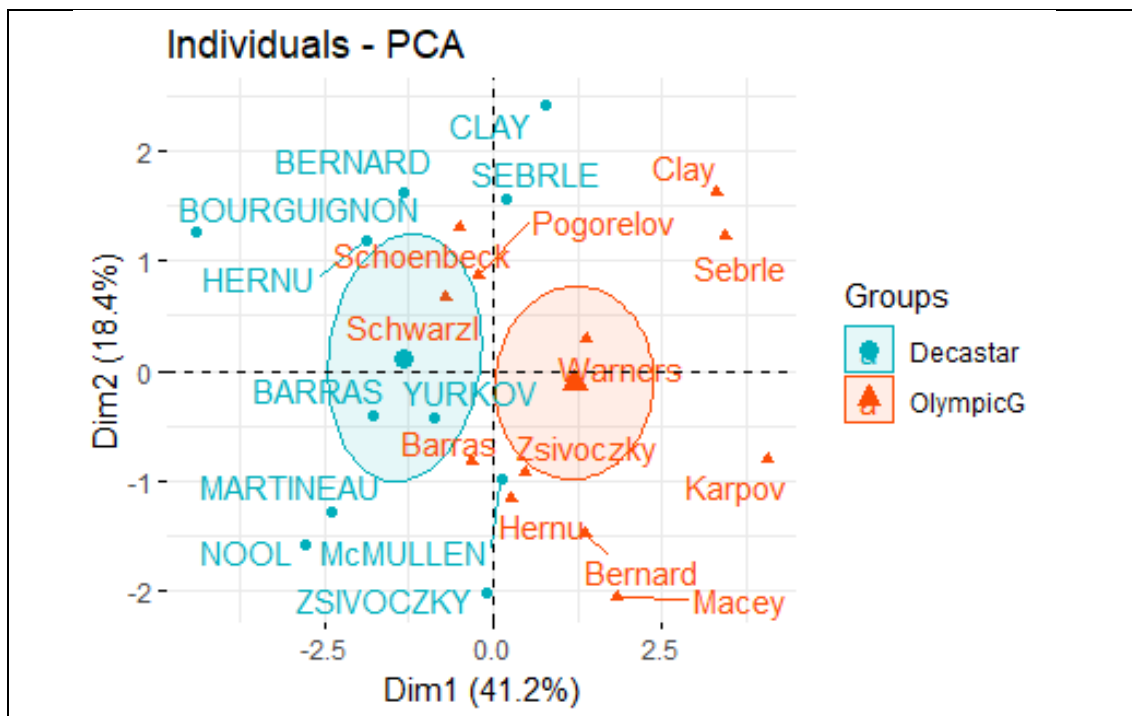
## 8. Variabel Pelengkap Individuals PCA

**Coding dan Output**

```
> ind.scaled <- scale(ind.sup,
+                     center = res.pca$center,
+                     scale = res.pca$scale)
> # Coordinates of the individuals
> coord_func <- function(ind, loadings){
+   r <- loadings*ind
+   apply(r, 2, sum)
+ }
> pca.loadings <- res.pca$rotation
> ind.sup.coord <- t(apply(ind.scaled, 1, coord_func, pca.loadings ))
> ind.sup.coord[, 1:4]
      PC1      PC2      PC3      PC4
KARPOV  0.7772521  0.76237804  1.5971253  1.6863286
WARNERS -0.3779697 -0.11891968  1.7005146 -0.6908084
Nool    -0.5468405  1.93402211  0.4724184 -2.2283706
Drews   -1.0848227  0.01703198  2.9818031 -1.5006207

> #VARIABEL PELENGKAP
> groups <- as.factor(decathlon2$Competition[1:23])
> fviz_pca_ind(res.pca,
+             col.ind = groups, # color by groups
+             palette = c("#00AFBB", "#FC4E07"),
+             addEllipses = TRUE, # Concentration ellipses
+             ellipse.type = "confidence",
+             legend.title = "Groups",
+             repel = TRUE
+ )
```





#### Hasil Analisis Singkat

Kode di atas melakukan Pemusatan (dengan parameter center) dan penskalaan individu (dengan parameter scale).

Tabel menunjukkan skor individu tambahan pada masing-masing komponen utama. Nilai yang lebih tinggi menunjukkan kontribusi yang lebih besar terhadap komponen tersebut. Nilai positif menunjukkan hubungan positif dengan komponen, sedangkan nilai negatif menunjukkan hubungan negatif.

Grafik di atas adalah visual untuk pengelompokkan, disana terdapat pengrupan menjadi 2 yaitu Decastar dan OlympicG ada warna yaitu hijau ke biruan dan orange.

### 9. Hitung Koordinat Variabel Pengelompokkan

#### Coding dan Output

```
> #koordinat group
> library(magrittr) # for pipe %>%
> library(dplyr) # everything else
> # 1. Individual coordinates
> res.ind <- get_pca_ind(res.pca)
> # 2. Coordinate of groups
> coord.groups <- res.ind$coord %>%
+   as_data_frame() %>%
+   select(Dim.1, Dim.2) %>%
+   mutate(competition = groups) %>%
+   group_by(competition) %>%
+   summarise(
+     Dim.1 = mean(Dim.1),
+     Dim.2 = mean(Dim.2)
+   )
> coord.groups
# A tibble: 2 x 3
  competition Dim.1 Dim.2
  <fct>      <dbl> <dbl>
1 Decastar   -1.31  0.119
2 OlympicG    1.20 -0.109
```

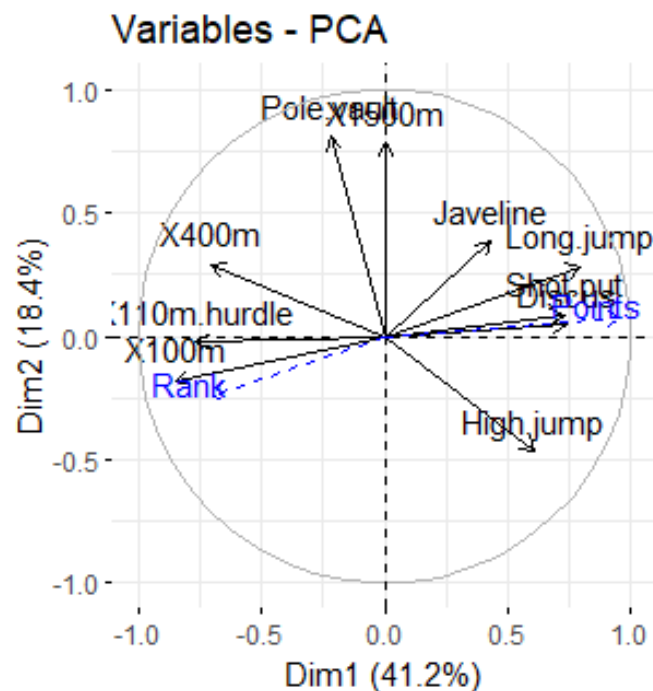
#### Hasil Analisis Singkat

Berdasarkan output, grup "OlympicG" memiliki centroid di posisi (1.204428, 0.1092046). Ini menunjukkan bahwa rata-rata atlet dalam grup tersebut memiliki skor tinggi pada PCA1 dan skor sedikit di atas rata-rata pada PCA 2.

## 10. Variable – PCA

### Coding dan Output

```
> #variabel kuantitatif
> quanti.sup <- decathlon2[1:23, 11:12, drop = FALSE]
> head(quanti.sup)
      Rank Points
SEBRLE    1  8217
CLAY      2  8122
BERNARD   4  8067
YURKOV    5  8036
ZSIVOCZKY 7  8004
McMULLEN  8  7995
> # Koordinat variabel kuantitatif
> # Predict coordinates and compute cos2
> quanti.coord <- cor(quanti.sup, res.pca$x)
> quanti.cos2 <- quanti.coord^2
> # Graph of variables including supplementary variables
> p <- fviz_pca_var(res.pca)
> fviz_add(p, quanti.coord, color = "blue", geom = "arrow")
```



### Hasil Analisis Singkat

Kode di atas menampilkan ranking untuk tiap individu. Dan grafik di atas merupakan gabungan dari hasil visualisasi sebelumnya. Untuk warna biru di atas merupakan gambaran arah vektor untuk pengukuran ranking dan point.

## 11. Menghitung Hasil PCA

### Coding dan Output

```

> # Helper function
> #::::::::::::::::::::::::::::::::::::
> var_coord_func <- function(loadings, comp.sdev){
+   loadings*comp.sdev
+ }
> # Compute Coordinates
> #::::::::::::::::::::::::::::::::::::
> loadings <- res.pca$rotation
> sdev <- res.pca$sdev
> var.coord <- t(apply(loadings, 1, var_coord_func, sdev))
> head(var.coord[, 1:4])

```

	PC1	PC2	PC3	PC4
X100m	-0.8506257	-0.17939806	-0.3015564	0.03357320
Long.jump	0.7941806	0.28085695	0.1905465	-0.11538956
Shot.put	0.7339127	0.08540412	-0.5175978	0.12846837
High.jump	0.6100840	-0.46521415	-0.3300852	0.14455012
X400m	-0.7016034	0.29017826	-0.2835329	0.43082552
X110m.hurdle	-0.7641252	-0.02474081	-0.4488873	-0.01689589

```

> # Compute Cos2
> #::::::::::::::::::::::::::::::::::::
> var.cos2 <- var.coord^2
> head(var.cos2[, 1:4])

```

	PC1	PC2	PC3	PC4
X100m	0.7235641	0.0321836641	0.09093628	0.0011271597
Long.jump	0.6307229	0.0788806285	0.03630798	0.0133147506
Shot.put	0.5386279	0.0072938636	0.26790749	0.0165041211
High.jump	0.3722025	0.2164242070	0.10895622	0.0208947375
X400m	0.4922473	0.0842034209	0.08039091	0.1856106269
X110m.hurdle	0.5838873	0.0006121077	0.20149984	0.0002854712

```

> # Compute contributions
> #::::::::::::::::::::::::::::::::::::
> comp.cos2 <- apply(var.cos2, 2, sum)
> contrib <- function(var.cos2, comp.cos2){var.cos2*100/comp.cos2}
> var.contrib <- t(apply(var.cos2,1, contrib, comp.cos2))
> head(var.contrib[, 1:4])

```

	PC1	PC2	PC3	PC4
X100m	17.544293	1.7505098	7.338659	0.13755240
Long.jump	15.293168	4.2904162	2.930094	1.62485936
Shot.put	13.060137	0.3967224	21.620432	2.01407269
High.jump	9.024811	11.7715838	8.792888	2.54987951
X400m	11.935544	4.5799296	6.487636	22.65090599
X110m.hurdle	14.157544	0.0332933	16.261261	0.03483735

```

> # Coordinates of individuals
> #::::::::::::::::::::::::::::::::::::
> ind.coord <- res.pca$x
> head(ind.coord[, 1:4])

```

	PC1	PC2	PC3	PC4
SEBRLE	0.1912074	1.5541282	-0.6283688	0.08205241
CLAY	0.7901217	2.4204156	1.3568870	1.26984296
BERNARD	-1.3292592	1.6118687	-0.1961500	-1.92092203
YURKOV	-0.8694134	-0.4328779	-2.4739822	0.69723814
ZSIVOCZKY	-0.1057450	-2.0233632	1.3049312	-0.09929630
McMULLEN	0.1185550	-0.9916237	0.8435582	1.31215266

```

> # Cos2 of individuals
> #::::::::::::::::::::::::::::::::::
> # 1. square of the distance between an individual and the
> # PCA center of gravity
> center <- res.pca$center
> scale<- res.pca$scale
> getdistance <- function(ind_row, center, scale){
+   return(sum(((ind_row-center)/scale)^2))
+ }
> d2 <- apply(decathlon2.active,1,getdistance, center, scale)
> # 2. Compute the cos2. The sum of each row is 1
> cos2 <- function(ind.coord, d2){return(ind.coord^2/d2)}
> ind.cos2 <- apply(ind.coord, 2, cos2, d2)
> head(ind.cos2[, 1:4])

```

	PC1	PC2	PC3	PC4
SEBRLE	0.007530179	0.49747323	0.081325232	0.001386688
CLAY	0.048701249	0.45701660	0.143628117	0.125791741
BERNARD	0.197199804	0.28996555	0.004294015	0.411819183
YURKOV	0.096109800	0.02382571	0.778230322	0.061812637
ZSIVOCZKY	0.001574385	0.57641944	0.239754152	0.001388216
McMULLEN	0.002175437	0.15219499	0.110137872	0.266486530

```

> # Contributions of individuals
> #::::::::::::::::::::::::::::::::::::
> contrib <- function(ind.coord, comp.sdev, n.ind){
+   100*(1/n.ind)*ind.coord^2/comp.sdev^2
+ }
> ind.contrib <- t(apply(ind.coord, 1, contrib,
+   res.pca$sdev, nrow(ind.coord)))
> head(ind.contrib[,1:4])

```

	PC1	PC2	PC3	PC4
SEBRLE	0.03854254	5.7118249	1.3854184	0.03572215
CLAY	0.65814114	13.8541889	6.4600973	8.55568792
BERNARD	1.86273218	6.1441319	0.1349983	19.57827284
YURKOV	0.79686310	0.4431309	21.4755770	2.57939100
ZSIVOCZKY	0.01178829	9.6816398	5.9748485	0.05231437
McMULLEN	0.01481737	2.3253860	2.4967890	9.13531719

### Hasil Analisis Singkat

Lalu diatas terdapat perhitungan untuk tiap komponen utama. Dengan pertimbangan pembobotan, didapat kesimpulan baik dari tiap individu dan tiap jenis olahraga untuk mencari komponen utama. Cos2 adalah kualitas representasi variabel dengan pembobotan untuk menghitung kontribusi dalam olahraga dan koordinat dari tiap individu