

Ch13Teetor

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```
#-----Beyond Basic Numerics and Statistics-----#

#importing actual clickthrough data set to perform r commands and operations on

actualdata<-read.csv("C:/Users/AliDesktop/Desktop/Bit Briefcase/Big Data/Kaggle/CTR/train.csv",
                     nrow=1000)

#Checking variables of the clickthrough data set

names(actualdata)

## [1] "id"           "click"        "hour"
## [4] "C1"          "banner_pos"   "site_id"
## [7] "site_domain" "site_category" "app_id"
## [10] "app_domain"  "app_category" "device_id"
## [13] "device_ip"   "device_model" "device_type"
## [16] "device_conn_type" "C14"         "C15"
## [19] "C16"         "C17"         "C18"
## [22] "C19"         "C20"         "C21"

#Minimizing or Maximizing a Single-Parameter Function
f <- function(x) 3*x^4 - 2*x^3 + 3*x^2 - 4*x + 5
optimize(f, lower=-20, upper=20) #minimise

## $minimum
## [1] 0.5972778
##
## $objective
## [1] 3.636756

#Performing Principal Component Analysis-using prcomp function
r <- prcomp(~actualdata$C14+actualdata$C17+actualdata$C18+actualdata$C19)
summary(r) #summary shows the proportion of variation captured by each component

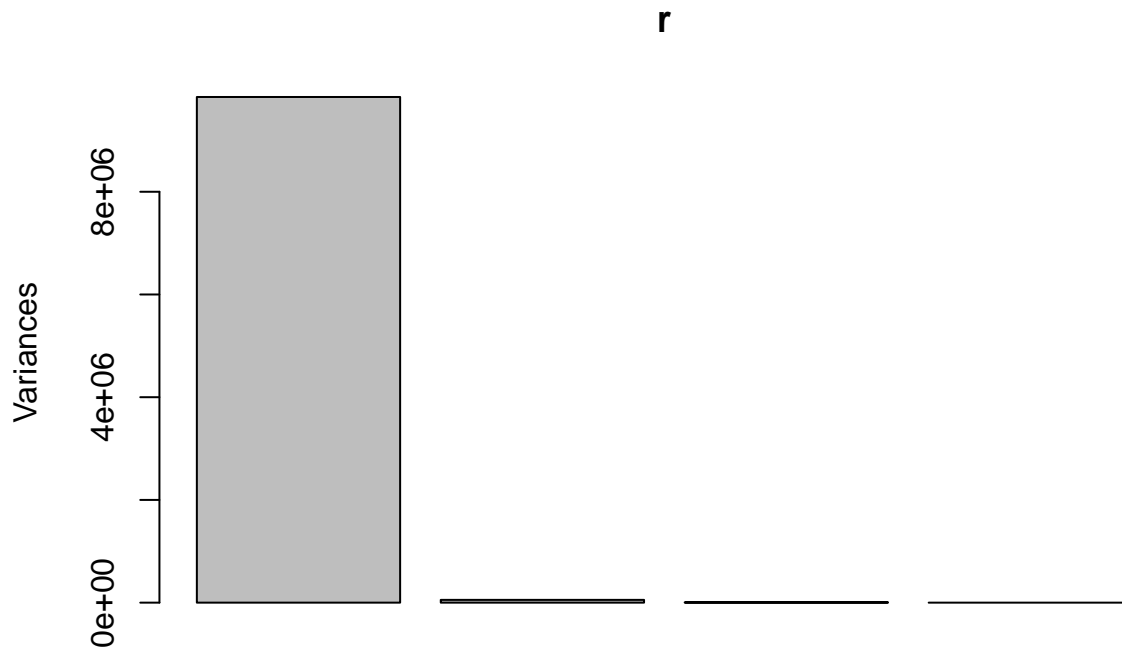
## Importance of components:
##
##          PC1          PC2          PC3          PC4
## Standard deviation 3137.7415 232.61219 82.26517 1.157
## Proportion of Variance 0.9939 0.00546 0.00068 0.000
## Cumulative Proportion 0.9939 0.99932 1.00000 1.000

#above summary shows that first component PC1 captures ~99.5% of the variance and other
# components capturing remaining

#The PCA recasts data into a vector space where the 1st dimension captures the most variance
#and the second dimension captures the second most, and so forth.
print(r)
```

```
## Standard deviations:
## [1] 3137.741496 232.612190 82.265173 1.157193
##
## Rotation:
##          PC1          PC2          PC3          PC4
## actualdata$C14 9.929803e-01 0.006271059 0.118113084 -0.0002597791
## actualdata$C17 1.182074e-01 -0.087520751 -0.989121214 0.0025186994
## actualdata$C18 4.627909e-05 0.001345340 -0.002659899 -0.9999955564
## actualdata$C19 4.134454e-03 0.996142049 -0.087643873 0.0015734716
```

*#in order to view a bar chart of the variances of the principal components,
use the below command*
`plot(r)`



#in order to rotate our actualdata to the principal components, use the below command-
`predict(r)`

```
##          PC1          PC2          PC3          PC4
## 1 -2025.59605 -80.39039 13.895863 0.55014573
## 2 -2027.58201 -80.40293 13.659637 0.55066529
## 3 -2027.58201 -80.40293 13.659637 0.55066529
## 4 -2025.59605 -80.39039 13.895863 0.55014573
## 5 1290.22331 -98.19903 -32.090643 0.80196076
## 6 -797.55799 306.20376 -52.496282 1.30367842
## 7 2669.96157 -100.68295 -40.873256 0.88583331
```

## 8	2942.91290	-102.57408	-49.544673	-2.08102705
## 9	-2024.60307	-80.38412	14.013976	0.54988595
## 10	4007.44361	20.88308	-56.590347	-1.84692790
## 11	30.88167	-85.65913	5.350226	-1.33904846
## 12	-2030.56096	-80.42174	13.305298	0.55144463
## 13	2881.97075	-88.14652	157.244630	0.38553480
## 14	2073.25935	550.38810	-62.624469	1.79199025
## 15	3294.20405	409.91659	-49.867187	1.62557873
## 16	-2032.54692	-80.43428	13.069071	0.55196419
## 17	204.86570	-90.65079	-43.174253	-1.20864131
## 18	-2023.61009	-80.37785	14.132089	0.54962617
## 19	-11245.42011	-33.03439	71.519994	-1.96611500
## 20	3548.29217	19.47147	-48.655709	-1.89118164
## 21	2660.03177	-100.74566	-42.054387	0.88843110
## 22	-2025.59605	-80.39039	13.895863	0.55014573
## 23	2673.93350	-100.65786	-40.400804	0.88479420
## 24	-2030.56096	-80.42174	13.305298	0.55144463
## 25	-2025.59605	-80.39039	13.895863	0.55014573
## 26	3982.71172	-110.49561	-44.888706	-2.05594755
## 27	3294.20405	409.91659	-49.867187	1.62557873
## 28	3928.63731	151.29274	-61.370932	-1.66241301
## 29	2673.93350	-100.65786	-40.400804	0.88479420
## 30	-2025.59605	-80.39039	13.895863	0.55014573
## 31	-2023.61009	-80.37785	14.132089	0.54962617
## 32	-882.62901	-87.29083	-10.667500	-2.34091901
## 33	-2023.61009	-80.37785	14.132089	0.54962617
## 34	-2032.54692	-80.43428	13.069071	0.55196419
## 35	2669.96157	-100.68295	-40.873256	0.88583331
## 36	1284.96936	157.04231	-52.276769	-1.80121460
## 37	2881.97075	-88.14652	157.244630	0.38553480
## 38	1290.22331	-98.19903	-32.090643	0.80196076
## 39	1290.22331	-98.19903	-32.090643	0.80196076
## 40	-2025.59605	-80.39039	13.895863	0.55014573
## 41	-2027.58201	-80.40293	13.659637	0.55066529
## 42	2176.76552	547.15803	-94.452607	1.87709489
## 43	-2023.61009	-80.37785	14.132089	0.54962617
## 44	3982.71172	-110.49561	-44.888706	-2.05594755
## 45	3973.74633	29.06810	-10.792028	-1.95929517
## 46	-1517.35387	47.42585	-15.539795	-2.17609326
## 47	-2030.56096	-80.42174	13.305298	0.55144463
## 48	-678.86265	-86.61075	-38.945215	-1.25535325
## 49	-797.55799	306.20376	-52.496282	1.30367842
## 50	2660.03177	-100.74566	-42.054387	0.88843110
## 51	2673.93350	-100.65786	-40.400804	0.88479420
## 52	-2026.58903	-80.39666	13.777750	0.55040551
## 53	1290.22331	-98.19903	-32.090643	0.80196076
## 54	-2030.56096	-80.42174	13.305298	0.55144463
## 55	-2030.56096	-80.42174	13.305298	0.55144463
## 56	-1106.29302	-83.04182	-18.563990	-2.32454967
## 57	2944.89886	-102.56153	-49.308447	-2.08154661
## 58	-2027.58201	-80.40293	13.659637	0.55066529
## 59	1290.22331	-98.19903	-32.090643	0.80196076
## 60	1290.22331	-98.19903	-32.090643	0.80196076
## 61	2881.97075	-88.14652	157.244630	0.38553480

## 872	2817.57725	22.18290	-52.308747	-1.91163404
## 873	-460.78493	-79.75959	49.189310	-2.47047803
## 874	204.86570	-90.65079	-43.174253	-1.20864131
## 875	3928.63731	151.29274	-61.370932	-1.66241301
## 876	-2025.59605	-80.39039	13.895863	0.55014573
## 877	-797.55799	306.20376	-52.496282	1.30367842
## 878	2881.97075	-88.14652	157.244630	0.38553480
## 879	-2025.59605	-80.39039	13.895863	0.55014573
## 880	-2028.57499	-80.40920	13.541524	0.55092507
## 881	2073.25935	550.38810	-62.624469	1.79199025
## 882	2944.89886	-102.56153	-49.308447	-2.08154661
## 883	2074.25233	550.39438	-62.506356	1.79173047
## 884	-2023.61009	-80.37785	14.132089	0.54962617
## 885	-2026.58903	-80.39666	13.777750	0.55040551
## 886	1290.22331	-98.19903	-32.090643	0.80196076
## 887	-550.75205	35.86443	-55.705012	-2.04007781
## 888	2944.89886	-102.56153	-49.308447	-2.08154661
## 889	-59.10135	-87.99275	-25.416715	-1.26451501
## 890	3294.20405	409.91659	-49.867187	1.62557873
## 891	-2023.61009	-80.37785	14.132089	0.54962617
## 892	-2025.59605	-80.39039	13.895863	0.55014573
## 893	-59.10135	-87.99275	-25.416715	-1.26451501
## 894	3982.71172	-110.49561	-44.888706	-2.05594755
## 895	-2029.56798	-80.41547	13.423411	0.55118485
## 896	2669.96157	-100.68295	-40.873256	0.88583331
## 897	204.86570	-90.65079	-43.174253	-1.20864131
## 898	-2024.60307	-80.38412	14.013976	0.54988595
## 899	2074.25233	550.39438	-62.506356	1.79173047
## 900	1969.48929	164.93198	-67.211680	-0.71919119
## 901	3928.63731	151.29274	-61.370932	-1.66241301
## 902	2660.03177	-100.74566	-42.054387	0.88843110
## 903	-2029.56798	-80.41547	13.423411	0.55118485
## 904	-2023.61009	-80.37785	14.132089	0.54962617
## 905	2673.93350	-100.65786	-40.400804	0.88479420
## 906	2669.96157	-100.68295	-40.873256	0.88583331
## 907	3928.63731	151.29274	-61.370932	-1.66241301
## 908	1290.22331	-98.19903	-32.090643	0.80196076
## 909	-1517.35387	47.42585	-15.539795	-2.17609326
## 910	2881.97075	-88.14652	157.244630	0.38553480
## 911	2881.97075	-88.14652	157.244630	0.38553480
## 912	-2028.57499	-80.40920	13.541524	0.55092507
## 913	-2025.59605	-80.39039	13.895863	0.55014573
## 914	2198.37847	70.89216	426.439365	-3.14818679
## 915	2881.97075	-88.14652	157.244630	0.38553480
## 916	-2032.54692	-80.43428	13.069071	0.55196419
## 917	-2024.60307	-80.38412	14.013976	0.54988595
## 918	-2024.60307	-80.38412	14.013976	0.54988595
## 919	-2023.61009	-80.37785	14.132089	0.54962617
## 920	-2023.61009	-80.37785	14.132089	0.54962617
## 921	1290.22331	-98.19903	-32.090643	0.80196076
## 922	-2029.56798	-80.41547	13.423411	0.55118485
## 923	2449.30742	-99.68921	-40.041728	-2.12526661
## 924	3294.20405	409.91659	-49.867187	1.62557873
## 925	-2027.58201	-80.40293	13.659637	0.55066529

## 926	3928.63731	151.29274	-61.370932	-1.66241301
## 927	-1106.29302	-83.04182	-18.563990	-2.32454967
## 928	2669.96157	-100.68295	-40.873256	0.88583331
## 929	-2025.59605	-80.39039	13.895863	0.55014573
## 930	-2029.56798	-80.41547	13.423411	0.55118485
## 931	-2032.54692	-80.43428	13.069071	0.55196419
## 932	-2023.61009	-80.37785	14.132089	0.54962617
## 933	1290.22331	-98.19903	-32.090643	0.80196076
## 934	2881.97075	-88.14652	157.244630	0.38553480
## 935	-2023.61009	-80.37785	14.132089	0.54962617
## 936	1290.22331	-98.19903	-32.090643	0.80196076
## 937	-2030.56096	-80.42174	13.305298	0.55144463
## 938	-2029.56798	-80.41547	13.423411	0.55118485
## 939	2817.57725	22.18290	-52.308747	-1.91163404
## 940	-2024.60307	-80.38412	14.013976	0.54988595
## 941	2669.96157	-100.68295	-40.873256	0.88583331
## 942	-2028.57499	-80.40920	13.541524	0.55092507
## 943	-2032.54692	-80.43428	13.069071	0.55196419
## 944	-1517.35387	47.42585	-15.539795	-2.17609326
## 945	-2026.58903	-80.39666	13.777750	0.55040551
## 946	-2025.59605	-80.39039	13.895863	0.55014573
## 947	2048.75475	287.98135	-79.749693	-1.51663335
## 948	2673.93350	-100.65786	-40.400804	0.88479420
## 949	-2025.59605	-80.39039	13.895863	0.55014573
## 950	2673.93350	-100.65786	-40.400804	0.88479420
## 951	204.86570	-90.65079	-43.174253	-1.20864131
## 952	4007.44361	20.88308	-56.590347	-1.84692790
## 953	-2032.54692	-80.43428	13.069071	0.55196419
## 954	3928.63731	151.29274	-61.370932	-1.66241301
## 955	-2025.59605	-80.39039	13.895863	0.55014573
## 956	1290.22331	-98.19903	-32.090643	0.80196076
## 957	-2030.56096	-80.42174	13.305298	0.55144463
## 958	1290.22331	-98.19903	-32.090643	0.80196076
## 959	2673.93350	-100.65786	-40.400804	0.88479420
## 960	3929.74850	151.21149	-62.241940	-1.66015409
## 961	1312.20371	-94.16132	-30.837172	-1.19493291
## 962	3929.74850	151.21149	-62.241940	-1.66015409
## 963	43.31918	932.76506	-102.481724	0.31848309
## 964	-2028.57499	-80.40920	13.541524	0.55092507
## 965	2561.85665	1612.95198	353.664566	2.13682601
## 966	-2026.58903	-80.39666	13.777750	0.55040551
## 967	-2029.56798	-80.41547	13.423411	0.55118485
## 968	2673.93350	-100.65786	-40.400804	0.88479420
## 969	2881.97075	-88.14652	157.244630	0.38553480
## 970	-2023.61009	-80.37785	14.132089	0.54962617
## 971	2673.93350	-100.65786	-40.400804	0.88479420
## 972	-797.55799	306.20376	-52.496282	1.30367842
## 973	43.31918	932.76506	-102.481724	0.31848309
## 974	-2026.58903	-80.39666	13.777750	0.55040551
## 975	3928.63731	151.29274	-61.370932	-1.66241301
## 976	-473.67049	-88.75596	-53.664759	-2.20959491
## 977	-2032.54692	-80.43428	13.069071	0.55196419
## 978	2881.97075	-88.14652	157.244630	0.38553480
## 979	-2027.58201	-80.40293	13.659637	0.55066529

```
## 980    -2025.59605   -80.39039    13.895863   0.55014573
## 981     -473.67049   -88.75596   -53.664759  -2.20959491
## 982      165.34514   -82.31224   -41.560626  -2.20354860
## 983    -2026.58903   -80.39666    13.777750   0.55040551
## 984     2669.96157  -100.68295   -40.873256   0.88583331
## 985    -2027.58201   -80.40293    13.659637   0.55066529
## 986     2670.95455  -100.67668   -40.755143   0.88557353
## 987     3280.51078    670.72422   -75.590699   2.04424368
## 988     2669.96157  -100.68295   -40.873256   0.88583331
## 989    -2027.58201   -80.40293    13.659637   0.55066529
## 990     2669.96157  -100.68295   -40.873256   0.88583331
## 991    -2026.58903   -80.39666    13.777750   0.55040551
## 992     2881.97075   -88.14652   157.244630   0.38553480
## 993    -2025.59605   -80.39039    13.895863   0.55014573
## 994   -11412.06571   -23.99814    75.066434  -2.97110801
## 995     3928.63731   151.29274   -61.370932  -1.66241301
## 996    -2023.61009   -80.37785    14.132089   0.54962617
## 997     2881.97075   -88.14652   157.244630   0.38553480
## 998     1290.22331   -98.19903   -32.090643   0.80196076
## 999    -2028.57499   -80.40920    13.541524   0.55092507
## 1000   -2030.56096   -80.42174    13.305298   0.55144463
```

```
#-----Performing Simple Orthogonal Regression-Also called as total least squares-----
#To create a linear model using orthogonal regression in which variances of C18 and C19
# are treated symmetrically in order to implement a basic orthogonal regression in R,
# we perform PCA
```

```
r <- prcomp( ~ actualdata$C18 + actualdata$C19 )
#Now, using the rotations to compute the slope:
slope <- r$rotation[2,1] / r$rotation[1,1]
#Now, calculating the intercept from the slope:
intercept <- r$center[2] - slope*r$center[1]
```

```
#-----Finding Clusters in the Data-----
```

```
#creating a subset of the actual clickthrough data set to include only numerical variables
# to understand clustering
```

```
#d<-dist(x)      #Compute distances between observations
#hc <- hclust(d)  #Form hierarchical clusters
```

```
#the result clust below is the vector of numbers between 1 and 3, one for each observation in x
#Each number classifies its corresponding observation into one of the n clusters.
#clust <- cutree(hc, k=3)  #Organize them into the 3 largest clusters
```

```
#-----Predicting a Binary-Valued Variable (Logistic Regression)-----
#A regression model to predict the probability of a binary event occurring
```

```
# install.packages("faraway")
# library(faraway)
#
# #Faraway gives an example of predicting a binary-valued variable:
# #test from the dataset pima is true if the patient tested positive for diabetes.
```

```

#
# data(pima, package="faraway")
# b <- factor(pima$test)
#
# #The predictors are diastolic blood pressure and body mass index (BMI).
# m <- glm(b ~ diastolic + bmi, family=binomial, data=pima)
#
# summary(m)    #results show that only the bmi variable is significant, p-value for it is
# # 1.95e-14
#
# #Since only bmi variable is significant, a reduced model can be created like below:
# m.red <- glm(b ~ bmi, family=binomial, data=pima)
#
#
# #Now using the model to calculate the probability that someone with an avg BMI(32.0)
# # will test positive for diabetes
# newdata <- data.frame(bmi=32.0)
#
# predict(m.red, type="response", newdata=newdata)
# #According to this model, the probability is about 33.3%

#-----Factor Analysis-----

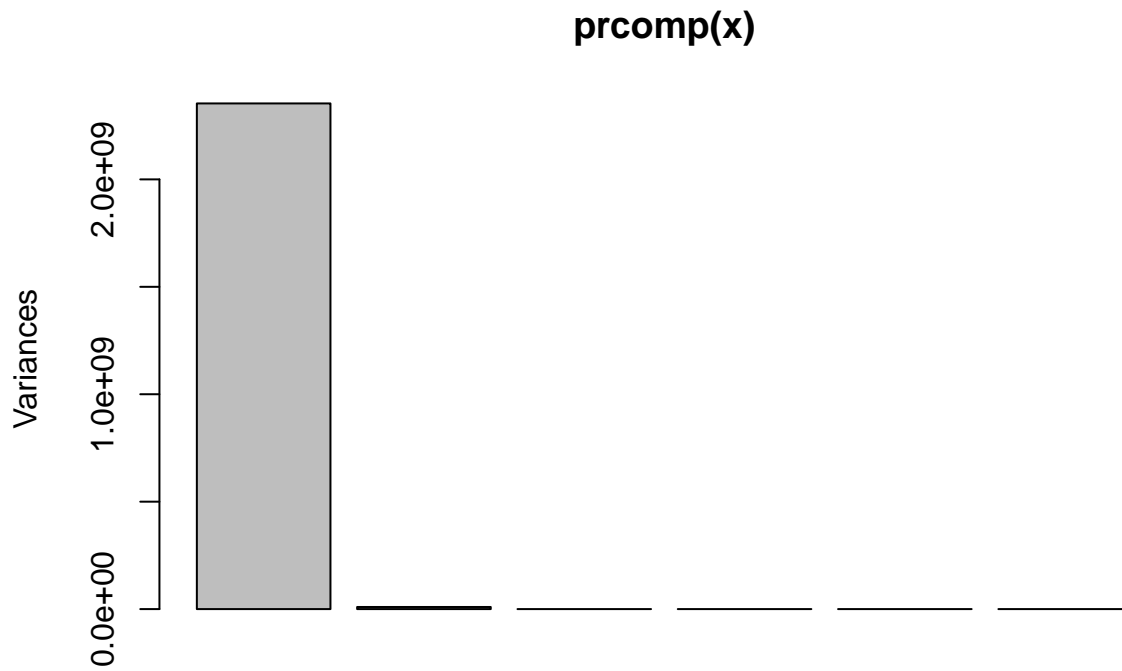
#in order to discover what the variables in a dataset have in common, we use the factanal
# function

#creating a subset of the actual clickthrough data set to include only numerical variables
# since factor analysis is for numerical ones

x<-data.frame(actualdata$C14, actualdata$C18, actualdata$C19, actualdata$C17, actualdata$C20,
              actualdata$C21)

#Plotting the PCA to see the variance captured by the components
plot(prcomp(x))

```



```
factanal(x,factors=2) #The p-value is 9.4e-12. Small p-value (<0.05) indicates that the two
```

```
##
## Call:
## factanal(x = x, factors = 2)
##
## Uniquenesses:
## actualdata.C14 actualdata.C18 actualdata.C19 actualdata.C17 actualdata.C20
##          0.005          0.464          0.858          0.045          0.914
## actualdata.C21
##          0.220
##
## Loadings:
##          Factor1 Factor2
## actualdata.C14  0.997
## actualdata.C18  0.135  0.719
## actualdata.C19          0.372
## actualdata.C17  0.977
## actualdata.C20          0.290
## actualdata.C21  0.377 -0.799
##
##          Factor1 Factor2
## SS loadings    2.115  1.378
## Proportion Var  0.353  0.230
## Cumulative Var  0.353  0.582
```



```
##  
## Test of the hypothesis that 2 factors are sufficient.  
## The chi square statistic is 95.99 on 4 degrees of freedom.  
## The p-value is 7.03e-20
```

```
# factors are insufficient
```

```
#In cases where p-value>0.05, it will help us to conclude that factors are sufficient  
#and % of individual variance and cumulative variance they explain
```