

Referências Bibliográficas

AFIFI, A.A. e CLARK, V. (1984). Computer – Aided Multivariate Analysis. Lifetime Learning Publications. Belm. California.

ANDERSON, T.W. (1984). An Introduction to Multivariate Statistical Analysis. 2ed. New York : John Wiley & Sons.

BARROSO, L.P. (2003). Tópicos de Análise Multivariada. Universidade de São Paulo.

BURT, C. (1941). The Vector of the Mind: An Introduction to Factor Analysis in Psychology. New York: Mc.Millan.

CAZAR, R.A. (2003). An Exercise on Chemometrics for a Quantitative Analysis Course. Madison. **Journal of Chemical Education**.

CHATFIELD, C. e COLLINS, A.J. (1980). Introduction to Multivariate Analysis. Chapman and Hall. New York.

CLIFF, N. e HAMBURGE, C.D. (1967). The Study of Sampling Errors in Factor Analysis by Means of Artificial Experiments. **Psychological Bulletin** 68: 430-45.

COOLEY, W.W. e LOHNES, P.R. (1971). Multivariate Data Analysis. John Wiley. New York.

COSTA, Giovani Glaucio de O. (2003). Busca de Fatores Associados à Prática de Atos Infracionais por Parte de Adolescentes no Estado do Rio de Janeiro: Um Estudo Preliminar, Estudo Orientado, PUC-RIO.

DAVID, A.A.; KUMAR, V. and GEORGE, S. Day. (1984). Marketing Research.

DILLON, W.R. e GOLDSTEIN, M. (1984). Multivariate Analysis: Methods and Applications. New York: John Wiley & Sons.

EFRON, B. (1979). Bootstrap Methods: Another Look at the Jackknife, **The Annals of Statistic**, 7, 1-26.

EFRON, B. (1980). Computer Intensive Methods in Statistics” in Some Recent Advance in Statistic, Ed. J. Tiago de Oliveira e B. Epstein, Academia das Ciências de Lisboa, Lisboa.

EFRON, B. (1982). The Jackknife, the Bootstrap, and other Resampling Methods, CBNS 38, SIAM-NSF.

EVERITT, B.S. (1978). Graphical Techniques for Multivariate Data. Heinemann Educational Books. London.

FERREIRA, D.F. Análise Multivariada. Minas Gerais: Universidade Federal de Lavras.

GNANADESIKAN, R. (1997). Methods for Statistical Data Analysis of Multivariate Observations. John Wiley. New York.

HAIR, J.F. Jr.; ANDERSON, R.E.; TATHAN, R.L. e BLACK, W.C. (2005). Trad. Sant'Anna, Adonai Schlup ; Neto, Anselmo Chaves. Análise Multivariada de Dados. 5. ed. Porto Alegre : Bookman.

HAIR, J.F. Jr.; ANDERSON, R.E.; TATHAN, R.L. e BLACK, W.C. (1998). Multivariate Data Analysis. 5th ed. Upper Saddle River : Prentice Hall.

HARMAN, H.H. (1967). Modern Factor Analysis . 2 ed. Chicago: University of Chicago.

HAWKINS, D.M. Topics in Multivariate Analysis. Cambridge. University Press: Cambridge.

HOGAN, D.J. (2002). Ordenação Multivariada na Ecologia e seu Uso em Ciências Ambientais. São Paulo: Ambiente e Sociedade.

HOMPSON, B. (1984). Canonical Correlation Analysis: Uses and Interpretation. Beverly Hills: Sage Pub.

JOBSON, J. (1992) D. Applied Multivariate Data Analysis. Vol II: Categorical and Multivariate Methods. Springer Verlag. New York.

JOHNSON, D.E. (1998). Applied Multivariate Methods for Data Analysis. Pacific Grove: Duxbury Press..

JOHNSON, R A. and WICHERN, D.W. (1998). Applied Multivariate Statistical Analysis. 4ed . Upper Saddle River: Prentice Hall.

JOHNSON, R.A. and WICHERN, D.W. (1982). Applied Multivariate Statistical Analysis. Prentice Hall, Inc. Englewood Cliffs, New Jersey.

KAISER, H.F. (1958). The Varimax Criterion for Analytic Rotation in Factor Analysis. **Psychometrika**, 23, 187-200.

KAISER, H.F. (1974). A Second-Generation Little Jiffy. **Psychometrika** 35: 401-15

KAUFMAN, L. and ROUSSEEUW, P.J. (1990). Finding Groups in Data: An Introduction to Cluster Analysis. New York . Jonh Wile & Sons.

LEBART, L.; MORINEAU, A. and PIRON, M. (1998). Statistique Exploratoire Multidimensionnelle. França.

- LONG, J.S. (1983). *Confirmation Factor Analysis*. Beverly Hills: Sage Pub.
- MARDIA, K.V.; KENT, J.T. and BIBBY, J.M. (1989). *Multivariate Analysis*. London: Academic Press.
- MOITA NETO, J.M. e MOITA, G.C. (1998). *Uma Introdução à Análise de Dados Multivaridos*. São Paulo: Química Nova.
- MORRISON, D. (3rd ed.). (1982). *Multivariate Statistical Methods*. McGraw-Hill. New York.
- PEREIRA, A. (1999). *Guia Prático de Utilização do SPSS: Análise de Dados para Ciências Sociais e Psicologia*. Lisboa: Edições Silabo.
- PRADO, P.I.; LEWWINSOHN, T.M.; CARMO, R.L; PROENÇA, I.M.D. (1988). Estimativas Jackknife e Bootstrap para o Enviesamento e Desvio-padrão do índice de Gini. Doc.Trabalho N° 67, Cemapre, I.S.E , Lisboa.
- REIS, E. (1997). *Estatística Multivariada Aplicada*. Lisboa: Edições Silabo.
- REYMENT, R. e JORESOG, K.G. (1996). *Applied Factor Analysis in the Natural Science*. Cambridge. Cambridge University Press.
- ROSA, P.T.M. (2000). Modelos de “Credit Scoring”: Regressão Logística, CHAID e REAL . Dissertação de Mestrado. IME – Universidade de São Paulo.
- RUMMEL, R.J. (1970) *Applied Factor Analysis*. Evanston: Northwestern University Press.
- SAS Institute Inc. *SAS Procedures Guide, Version 8*. Cary, N.C.(1999). SAS Institute Inc.
- SAPORTA, G. (1996). *Probabilités Analyse Des Données Et Statistique*. França.
- SEBER, G.A.F. (1984). *Multivariate Observations*. John Wiley. New York.
- SIEGEL, S. (1975). *Estatística Não-paramétrica para Ciências do Comportamento*. São Paulo: McGraw-Hill Editora
- SHARMA, S. (1996). *Applied Multivariate Techniques*. New York: John Wile & Sons.
- SPEARMAN, C. (1940). General Intelligence Objectively Determined and Measured. **American Journal of Psychology**, 15: 201-293.
- STEWART, D.W. (1981). The applications and misapplications of factor analysis in marketing research. **Journal of Marketing Research**, 18, p.51-62, Feb.

TABACHNICK, B.G. and FIDEL, L.S. (2001). Using Multivariate Statistics . 4th ed. Allyn 7 Bacon , Boston.

THURSTONE, L.L. (1993). Multiple Factor Analysis. Chicago: Univ. Chicago Press.

THURSTONE, L.L. (1993). The Vector of Mind. Chicago: Univ. Chicago Press.

Anexo 1

Os Programas *Bootstrap* e *Jackknife* em R 2.1.1

O programa abaixo referê-se ao usado com a base de dados “Modo de Vida”.

```
#####  
#####  
# PROGRAMA FEITO EM 29/SETEMBRO/2005 - ÚLTIMA MODIFICAÇÃO  
08/OUTUBRO/2005  
#####  
#####  
# Usando a ajuda  
help("factanal")  
# Lendo os dados  
# Informe o número de variáveis  
m      <- 11  
# Número de variáveis  
x      <- matrix(scan("c:/Tese  
Gcosta/basepaises.txt"),,,m,T)  
# Análise Fatorial - guardando os dois fatores  
cargas <- factanal(x,factor=2)$loadings[,1:2]  
#####  
#####  
# INÍCIO DO BOOTSTRAP  
#####  
#####  
n      <- length(x[,1])                                #  
Tamanho da amostra original (40)  
N      <- length(x[,1])                                #  
Tamanho da re-amostra          (40)  
M      <- 1000                                         #  
Número de replicações          (1000)  
#####  
#####  
# Função auxiliar para gerar números de 1 até o tamanho  
da amostra  
rp <- function(N,n){  
y <- rep(NA,N)
```

```

for(i in 1:N){
  aux <- runif(1)
  lb <- 0
  ub <- 1/n
  for(j in 1:n){
    if(lb<aux && aux <= ub ) y[i] <- j
    lb <- lb + 1/n
    ub <- ub + 1/n
  }
}
return(y)
}
#=====
=====
F1 <- matrix(NA,length(x[1,]),M)
F2 <- matrix(NA,length(x[1,]),M)
# Fazendo ...
for(i in 1:M){
  w <- rp(N,n)
# Gerando valores da reamostra
  z <- x[w,]
# Alocando a reamostra
  y <- factanal(z,factor=2)$loading[,1:2]
# Análise Fatorial
  F1[,i] <- y[,1]
  F2[,i] <- y[,2]
}
F1.MEAN <- mean(F1[1,])
F2.MEAN <- mean(F2[1,])
F1.VAR <- var(F1[1,])
F2.VAR <- var(F2[1,])
alpha <- 0.05
q1 <- round(M*alpha/2)
q2 <- M - q1 + 1
F1.LI <- sort(F1[1,])[q1]
F1.LS <- sort(F1[1,])[q2]
F2.LI <- sort(F2[1,])[q1]
F2.LS <- sort(F2[1,])[q2]
for(i in 2:m){
F1.MEAN <- c(F1.MEAN,mean(F1[i,]))

```

```

F2.MEAN <- c(F2.MEAN,mean(F2[i,]))
F1.VAR <- c(F1.VAR,var(F1[i,]))
F2.VAR <- c(F2.VAR,var(F2[i,]))
F1.LI <- c(F1.LI,sort(F1[i,])[q1])
F1.LS <- c(F1.LS,sort(F1[i,])[q2])
F2.LI <- c(F2.LI,sort(F2[i,])[q1])
F2.LS <- c(F2.LS,sort(F2[i,])[q2])
}
F1.VIES <- cargas[,1] - F1.MEAN
F2.VIES <- cargas[,2] - F2.MEAN
F1.EMQ <- F1.VAR - F1.VIES^2
F2.EMQ <- F2.VAR - F2.VIES^2
F1.SISIC<- rep(NA,m)
F2.SISIC<- rep(NA,m)
for(i in 1:m){
  if(F1.LI[i]<= 0 && 0 <= F1.LS[i]){
    F1.SISIC[i] <- "NSIG"
  }
  else{
    F1.SISIC[i] <- "SIG"
  }
  if(F2.LI[i]<= 0 && 0 <= F2.LS[i]){
    F2.SISIC[i] <- "NSIG"
  }
  else{
    F2.SISIC[i] <- "SIG"
  }
}
# Salvando os dados num arquivo - distribuição amostral
options(digits=5)
w <- cbind(t(F1),t(F2))
write(t(w),file="c:/Tese Gcosta/dist-boot-
summary.txt",ncolumn=(2*m))
# Padronização ...
for(i in 1:m){
  F1[i,] <- (F1[i,] - F1.MEAN[i])/sqrt(F1.VAR[i]) #
Padronização
  F2[i,] <- (F2[i,] - F2.MEAN[i])/sqrt(F2.VAR[i]) #
Padronização
}

```

```

# Salvando os dados num arquivo - distribuição amostral
padronizada
options(digits=5)
w <- cbind(t(F1),t(F2))
write(t(w),file="c:/Tese Gcosta/dist-boot-padro-
summary.txt",ncolumn=(2*m))
# Calculando G
g <- cargas/sqrt(cbind(F1.VAR,F2.VAR))
p.valor <- matrix(NA,m,2)
for(j in 1:m){
  cont1 <- 0
  cont2 <- 0
  for(i in 1:M){
    # Fator 1
    if(F1[j,i] >= g[j,1] && g[j,1]>= 0 ) cont1 <- cont1 +
1 # Se g >= 0 ele só fará este
    if(F1[j,i] <= g[j,1] && g[j,1] < 0 ) cont1 <- cont1 +
1 # Se g < 0 ele só fará este
    # Fator 2
    if(F2[j,i] >= g[j,2] && g[j,2]>= 0 ) cont2 <- cont2 +
1 # Se g >= 0 ele só fará este
    if(F2[j,i] <= g[j,2] && g[j,2] < 0 ) cont2 <- cont2 +
1 # Se g < 0 ele só fará este
  }
  p.valor[j,1] <- cont1/M
  p.valor[j,2] <- cont2/M
}
F1.SIGVP <- rep(NA,m)
F2.SIGVP <- rep(NA,m)
for(i in 1:m){
  if( p.valor[i,1] > 0.05 ){
    F1.SIGVP[i] <- "NSIG"
  }else{
    F1.SIGVP[i] <- "SIG"
  }
  if( p.valor[i,2] > 0.05 ){
    F2.SIGVP[i] <- "NSIG"
  }else{
    F2.SIGVP[i] <- "SIG"
  }
}

```



```

}
p.valor
F1.VP <- p.valor[,1]
F2.VP <- p.valor[,2]
F1.C <- cargas[,1]
F2.C <- cargas[,2]
F1.summary <-
data.frame(cbind(F1.C,F1.MEAN,F1.VAR,F1.LI,F1.LS,F1.VIES,
F1.EMQ,F1.SISIC,F1.VP,F1.SIGVP))
F2.summary <-
data.frame(cbind(F2.C,F2.MEAN,F2.VAR,F2.LI,F2.LS,F2.VIES,
F2.EMQ,F2.SISIC,F2.VP,F2.SIGVP))
options(digits=5)
F1.summary          # RESUMO DO FATOR 1
F2.summary          # RESUMO DO FATOR 2
# Salvando os dados num arquivo - resumo estatístico
options(digits=5)
F.summary <-
rbind(cbind(F1.C,F1.MEAN,F1.VAR,F1.LI,F1.LS,F1.VIES,F1.EM
Q,F1.SISIC,F1.VP,F1.SIGVP),

cbind(F2.C,F2.MEAN,F2.VAR,F2.LI,F2.LS,F2.VIES,F2.EMQ,F2.S
ISIC,F2.VP,F2.SIGVP))
write(t(F.summary),file="c:/Tese Gcosta/boot-
summary.txt",ncolumn=10)
#=====
# FIM DO BOOTSTRAP
#=====
# Para aqui
#=====
# INÍCIO DO JACKKNIFE
#=====
=====

```

```

n      <- length(x[,1]) # Tamanho da amostra original (40)
N      <- length(x[,1]) # Tamanho da re-amostra      (40)
M      <- length(x[,1]) # Número de replicações     (40)
#=====
=====
F1  <- matrix(NA,length(x[1,]),M)
F2  <- matrix(NA,length(x[1,]),M)
# Fazendo ...
for(i in 1:M){
  if(i==1) w <- seq(2,N,1)
# Gerando valores da reamostra
  if(i==N) w <- seq(1,N-1,1)
# Gerando valores da reamostra
  if(i!=1 && i!=N) w <- c(seq(1,(i-1),1),seq((i+1),N,1))
# Gerando valores da reamostra
  z      <- x[w,]
# Alocando a reamostra
  y      <- factanal(z,factor=2)$loading[,1:2]
# Análise Fatorial
  F1[,i] <- y[,1]
  F2[,i] <- y[,2]
}
F1.MEAN <- mean(F1[1,])
F2.MEAN <- mean(F2[1,])
F1.VAR  <- var(F1[1,])
F2.VAR  <- var(F2[1,])
alpha   <- 0.05
q1      <- round(M*alpha/2)
q2      <- M - q1 + 1
F1.LI   <- sort(F1[1,])[q1]
F1.LS   <- sort(F1[1,])[q2]
F2.LI   <- sort(F2[1,])[q1]
F2.LS   <- sort(F2[1,])[q2]
for(i in 2:m){
F1.MEAN <- c(F1.MEAN,mean(F1[i,]))
F2.MEAN <- c(F2.MEAN,mean(F2[i,]))
F1.VAR  <- c(F1.VAR,var(F1[i,]))
F2.VAR  <- c(F2.VAR,var(F2[i,]))
F1.LI   <- c(F1.LI,sort(F1[i,])[q1])
F1.LS   <- c(F1.LS,sort(F1[i,])[q2])

```

```

F2.LI    <- c(F2.LI,sort(F2[i,])[q1])
F2.LS    <- c(F2.LS,sort(F2[i,])[q2])
}
F1.VIES  <- cargas[,1] - F1.MEAN
F2.VIES  <- cargas[,2] - F2.MEAN
F1.EMQ   <- F1.VAR - F1.VIES^2
F2.EMQ   <- F2.VAR - F2.VIES^2
F1.SISIC<- rep(NA,m)
F2.SISIC<- rep(NA,m)
for(i in 1:m){
  if(F1.LI[i]<= 0 && 0 <= F1.LS[i]){
    F1.SISIC[i] <- "NSIG"
  }
  else{
    F1.SISIC[i] <- "SIG"
  }
  if(F2.LI[i]<= 0 && 0 <= F2.LS[i]){
    F2.SISIC[i] <- "NSIG"
  }
  else{
    F2.SISIC[i] <- "SIG"
  }
}
# Salvando os dados num arquivo - distribuição amostral
options(digits=5)
w <- cbind(t(F1),t(F2))
write(t(w),file="c:/Tese Gcosta/dist-jack-
summary.txt",ncolumn=(2*m))
# Padronização ...
for(i in 1:m){
  F1[i,] <- (F1[i,] - F1.MEAN[i])/sqrt(F1.VAR[i]) #
Padronização
  F2[i,] <- (F2[i,] - F2.MEAN[i])/sqrt(F2.VAR[i]) #
Padronização
}
# Salvando os dados num arquivo - distribuição amostral
padronizada
options(digits=5)
w <- cbind(t(F1),t(F2))

```

```

write(t(w),file="c:/Tese Gcosta/dist-jack-padro-
summary.txt",ncolumn=(2*m))
# Calculando G
g <- cargas/sqrt(cbind(F1.VAR,F2.VAR))
p.valor <- matrix(NA,m,2)
for(j in 1:m){
  cont1 <- 0
  cont2 <- 0
  for(i in 1:M){
    # Fator 1
    if(F1[j,i] >= g[j,1] && g[j,1]>= 0 ) cont1 <- cont1 +
1 # Se g >= 0 ele só fará este
    if(F1[j,i] <= g[j,1] && g[j,1] < 0 ) cont1 <- cont1 +
1 # Se g < 0 ele só fará este
    # Fator 2
    if(F2[j,i] >= g[j,2] && g[j,2]>= 0 ) cont2 <- cont2 +
1 # Se g >= 0 ele só fará este
    if(F2[j,i] <= g[j,2] && g[j,2] < 0 ) cont2 <- cont2 +
1 # Se g < 0 ele só fará este
  }
  p.valor[j,1] <- cont1/M
  p.valor[j,2] <- cont2/M
}
F1.SIGVP <- rep(NA,m)
F2.SIGVP <- rep(NA,m)
for(i in 1:m){
  if( p.valor[i,1] > 0.05 ){
    F1.SIGVP[i] <- "NSIG"
  }else{
    F1.SIGVP[i] <- "SIG"
  }
  if( p.valor[i,2] > 0.05 ){
    F2.SIGVP[i] <- "NSIG"
  }else{
    F2.SIGVP[i] <- "SIG"
  }
}
p.valor
F1.VP <- p.valor[,1]
F2.VP <- p.valor[,2]

```

```

F1.C <- cargas[,1]
F2.C <- cargas[,2]
F1.summary <-
data.frame(cbind(F1.C,F1.MEAN,F1.VAR,F1.LI,F1.LS,F1.VIES,
F1.EMQ,F1.SISIC,F1.VP,F1.SIGVP))
F2.summary <-
data.frame(cbind(F2.C,F2.MEAN,F2.VAR,F2.LI,F2.LS,F2.VIES,
F2.EMQ,F2.SISIC,F2.VP,F2.SIGVP))
options(digits=5)
F1.summary          # RESUMO DO FATOR 1
F2.summary          # RESUMO DO FATOR 2
# Salvando os dados num arquivo - resumo estatístico
options(digits=5)
F.summary <-
rbind(cbind(F1.C,F1.MEAN,F1.VAR,F1.LI,F1.LS,F1.VIES,F1.EM
Q,F1.SISIC,F1.VP,F1.SIGVP),

cbind(F2.C,F2.MEAN,F2.VAR,F2.LI,F2.LS,F2.VIES,F2.EMQ,F2.S
ISIC,F2.VP,F2.SIGVP))
write(t(F.summary),file="c:/Tese Gcosta/jack-
summary.txt",ncolumn=10)
#=====
=====
# FIM DO JACKKNIFE
#=====
=====

```

Anexo 2

O Programa Bootstrap em SAS V.8

O programa abaixo referê-se ao usado com a base de dados “Creme Dental”.

```
*libname in 'C:\Tese Giovani';
libname in2 'C:\Tese Giovani\Terceira Fase';
options nodate ls=max ps=max;

PROC IMPORT OUT= IN2.basecremedental
            DATAFILE= "C:\Tese Giovani\Terceira
Fase\basecremedental.xls"
            DBMS=EXCEL2000 REPLACE;
            GETNAMES=YES;
RUN;

data base(drop=F7);
  set in2.basecremedental;
run;

%macro sel(i,j);
data base&i;
  set base;
  if _n_=&i then delete;
run;

proc factor data=base&i
            method=principal
            rotate=Varimax
            outstat=perfil&i
            nfactors=2 noprint;
  var V1-V6;
run;

data perfil_f1_&i(keep=i f1_v01-f1_v06)
  perfil_f2_&i(keep=i f2_v01-f2_v06);
  set perfil&i;
```

```

if _type_='UNROTATE' and _name_ in ('Factor1'
'Factor2');

if _name_='Factor1' then do;
    f1_v01= v1 ;    f1_v02= v2 ;    f1_v03= v3 ;    f1_v04=
v4 ;    f1_v05= v5 ;    f1_v06= v6 ;
end;
else if _name_='Factor2' then do;
    f2_v01= v1 ;    f2_v02= v2 ;    f2_v03= v3 ;    f2_v04=
v4 ;    f2_v05= v5 ;    f2_v06= v6 ;
end;
    i=1;

        if _name_='Factor1' then output perfil_f1_&i;
else if _name_='Factor2' then output perfil_f2_&i;
run;

data amostra&i(drop=i);
    merge perfil_f1_&i perfil_f2_&i;
    by i;
    amostra=&i;
run;

proc delete data=perfil&i;
proc delete data=perfil_f1_&i;
proc delete data=perfil_f2_&i;
proc delete data=base&i;
run;
%mend sel;
%sel (    1    ,    87575    ); %sel (    2    ,    42631    );
%sel (    3    ,    98398    ); %sel (    4    ,    82401    );
%sel (    5    ,    75217    ); %sel (    6    ,    23436    );
%sel (    7    ,    78850    ); %sel (    8    ,    99497    );
%sel (    9    ,    75267    ); %sel (   10    ,    12243    );
%sel (   11    ,    55216    ); %sel (   12    ,    90000    );
%sel (   13    ,    10855    ); %sel (   14    ,    20001    );
%sel (   15    ,    44804    ); %sel (   16    ,    31726    );
%sel (   17    ,    93948    ); %sel (   18    ,    74881    );
%sel (   19    ,    20262    ); %sel (   20    ,    13439    );
%sel (   21    ,    3817    ); %sel (   22    ,    6551    );

```

```

%sel ( 23 , 54403 ) ; %sel ( 24 , 11361 ) ;
%sel ( 25 , 97085 ) ; %sel ( 26 , 83919 ) ;
%sel ( 27 , 85594 ) ; %sel ( 28 , 35212 ) ;
%sel ( 29 , 32651 ) ; %sel ( 30 , 84167 ) ;
%sel ( 31 , 23953 ) ; %sel ( 32 , 56974 ) ;
%sel ( 33 , 59788 ) ; %sel ( 34 , 91566 ) ;
%sel ( 35 , 84992 ) ; %sel ( 36 , 20942 ) ;
%sel ( 37 , 84190 ) ; %sel ( 38 , 50512 ) ;
%sel ( 39 , 98086 ) ; %sel ( 40 , 23271 ) ;
%sel ( 41 , 59386 ) ; %sel ( 42 , 49980 ) ;
%sel ( 43 , 71145 ) ; %sel ( 44 , 29931 ) ;
%sel ( 45 , 64912 ) ; %sel ( 46 , 12436 ) ;
%sel ( 47 , 49793 ) ; %sel ( 48 , 33565 ) ;
%sel ( 49 , 57161 ) ; %sel ( 50 , 94119 ) ;
%sel ( 51 , 63066 ) ; %sel ( 52 , 88110 ) ;
%sel ( 53 , 44405 ) ; %sel ( 54 , 13065 ) ;
%sel ( 55 , 25214 ) ; %sel ( 56 , 47196 ) ;
%sel ( 57 , 29784 ) ; %sel ( 58 , 22568 ) ;
%sel ( 59 , 60846 ) ; %sel ( 60 , 83724 ) ;
%sel ( 61 , 72102 ) ; %sel ( 62 , 27420 ) ;
%sel ( 63 , 5148 ) ; %sel ( 64 , 77389 ) ;
%sel ( 65 , 48909 ) ; %sel ( 66 , 52879 ) ;
%sel ( 67 , 39849 ) ; %sel ( 68 , 72711 ) ;
%sel ( 69 , 14754 ) ; %sel ( 70 , 16441 ) ;
%sel ( 71 , 81171 ) ; %sel ( 72 , 91869 ) ;
%sel ( 73 , 64291 ) ; %sel ( 74 , 16099 ) ;
%sel ( 75 , 28189 ) ; %sel ( 76 , 58487 ) ;
%sel ( 77 , 16035 ) ; %sel ( 78 , 93280 ) ;
%sel ( 79 , 23064 ) ; %sel ( 80 , 28573 ) ;
%sel ( 81 , 32756 ) ; %sel ( 82 , 12601 ) ;
%sel ( 83 , 71664 ) ; %sel ( 84 , 31319 ) ;
%sel ( 85 , 3285 ) ; %sel ( 86 , 73965 ) ;
%sel ( 87 , 30113 ) ; %sel ( 88 , 56699 ) ;
%sel ( 89 , 59394 ) ; %sel ( 90 , 24559 ) ;
%sel ( 91 , 8943 ) ; %sel ( 92 , 49173 ) ;
%sel ( 93 , 43540 ) ; %sel ( 94 , 22720 ) ;
%sel ( 95 , 49001 ) ; %sel ( 96 , 51072 ) ;
%sel ( 97 , 91797 ) ; %sel ( 98 , 54853 ) ;
%sel ( 99 , 31344 ) ; %sel ( 100 , 44768 ) ;
%sel ( 101 , 80496 ) ; %sel ( 102 , 71547 ) ;

```



```

%sel ( 103 , 83316 ) ; %sel ( 104 , 64796 ) ;
%sel ( 105 , 79490 ) ; %sel ( 106 , 88229 ) ;
%sel ( 107 , 15576 ) ; %sel ( 108 , 83319 ) ;
%sel ( 109 , 90206 ) ; %sel ( 110 , 25221 ) ;
%sel ( 111 , 11345 ) ; %sel ( 112 , 35133 ) ;
%sel ( 113 , 57730 ) ; %sel ( 114 , 86329 ) ;
%sel ( 115 , 57344 ) ; %sel ( 116 , 87914 ) ;
%sel ( 117 , 21095 ) ; %sel ( 118 , 91279 ) ;
%sel ( 119 , 69902 ) ; %sel ( 120 , 21540 ) ;
%sel ( 121 , 63059 ) ; %sel ( 122 , 22079 ) ;
%sel ( 123 , 51355 ) ; %sel ( 124 , 75010 ) ;
%sel ( 125 , 85215 ) ; %sel ( 126 , 10590 ) ;
%sel ( 127 , 18821 ) ; %sel ( 128 , 60801 ) ;
%sel ( 129 , 45243 ) ; %sel ( 130 , 50163 ) ;
%sel ( 131 , 71707 ) ; %sel ( 132 , 46959 ) ;
%sel ( 133 , 99526 ) ; %sel ( 134 , 60669 ) ;
%sel ( 135 , 43433 ) ; %sel ( 136 , 69226 ) ;
%sel ( 137 , 83372 ) ; %sel ( 138 , 15105 ) ;
%sel ( 139 , 66160 ) ; %sel ( 140 , 68847 ) ;
%sel ( 141 , 61005 ) ; %sel ( 142 , 45044 ) ;
%sel ( 143 , 92248 ) ; %sel ( 144 , 86235 ) ;
%sel ( 145 , 28617 ) ; %sel ( 146 , 68294 ) ;
%sel ( 147 , 35647 ) ; %sel ( 148 , 2813 ) ;
%sel ( 149 , 39138 ) ; %sel ( 150 , 93455 ) ;
%sel ( 151 , 40786 ) ; %sel ( 152 , 73002 ) ;
%sel ( 153 , 71325 ) ; %sel ( 154 , 97954 ) ;
%sel ( 155 , 20338 ) ; %sel ( 156 , 57102 ) ;
%sel ( 157 , 10793 ) ; %sel ( 158 , 12959 ) ;
%sel ( 159 , 86666 ) ; %sel ( 160 , 61205 ) ;
%sel ( 161 , 13028 ) ; %sel ( 162 , 63430 ) ;
%sel ( 163 , 59274 ) ; %sel ( 164 , 44090 ) ;
%sel ( 165 , 28600 ) ; %sel ( 166 , 95663 ) ;
%sel ( 167 , 22556 ) ; %sel ( 168 , 40501 ) ;
%sel ( 169 , 72080 ) ; %sel ( 170 , 11798 ) ;
%sel ( 171 , 83621 ) ; %sel ( 172 , 62901 ) ;
%sel ( 173 , 70967 ) ; %sel ( 174 , 80505 ) ;
%sel ( 175 , 56130 ) ; %sel ( 176 , 66572 ) ;
%sel ( 177 , 20738 ) ; %sel ( 178 , 83718 ) ;
%sel ( 179 , 14546 ) ; %sel ( 180 , 75444 ) ;
%sel ( 181 , 48545 ) ; %sel ( 182 , 74441 ) ;

```

```

%sel ( 183 , 2142 ); %sel ( 184 , 51781 );
%sel ( 185 , 54372 ); %sel ( 186 , 6390 );
%sel ( 187 , 73222 ); %sel ( 188 , 28451 );
%sel ( 189 , 31665 ); %sel ( 190 , 93921 );
%sel ( 191 , 13335 ); %sel ( 192 , 75270 );
%sel ( 193 , 39017 ); %sel ( 194 , 97305 );
%sel ( 195 , 46917 ); %sel ( 196 , 86436 );
%sel ( 197 , 4297 ); %sel ( 198 , 13145 );
%sel ( 199 , 11559 ); %sel ( 200 , 40844 );

```

```

data in2.boot;

```

```

set

```

```

amostra1 amostra101
amostra2 amostra102
amostra3 amostra103
amostra4 amostra104
amostra5 amostra105
amostra6 amostra106
amostra7 amostra107
amostra8 amostra108
amostra9 amostra109
amostra10 amostra110
amostra11 amostra111
amostra12 amostra112
amostra13 amostra113
amostra14 amostra114
amostra15 amostra115
amostra16 amostra116
amostra17 amostra117
amostra18 amostra118
amostra19 amostra119
amostra20 amostra120
amostra21 amostra121
amostra22 amostra122
amostra23 amostra123
amostra24 amostra124
amostra25 amostra125
amostra26 amostra126
amostra27 amostra127
amostra28 amostra128

```

amostra29	amostra129
amostra30	amostra130
amostra31	amostra131
amostra32	amostra132
amostra33	amostra133
amostra34	amostra134
amostra35	amostra135
amostra36	amostra136
amostra37	amostra137
amostra38	amostra138
amostra39	amostra139
amostra40	amostra140
amostra41	amostra141
amostra42	amostra142
amostra43	amostra143
amostra44	amostra144
amostra45	amostra145
amostra46	amostra146
amostra47	amostra147
amostra48	amostra148
amostra49	amostra149
amostra50	amostra150
amostra51	amostra151
amostra52	amostra152
amostra53	amostra153
amostra54	amostra154
amostra55	amostra155
amostra56	amostra156
amostra57	amostra157
amostra58	amostra158
amostra59	amostra159
amostra60	amostra160
amostra61	amostra161
amostra62	amostra162
amostra63	amostra163
amostra64	amostra164
amostra65	amostra165
amostra66	amostra166
amostra67	amostra167
amostra68	amostra168

```
amostra69      amostral69
amostra70      amostral70
amostra71      amostral71
amostra72      amostral72
amostra73      amostral73
amostra74      amostral74
amostra75      amostral75
amostra76      amostral76
amostra77      amostral77
amostra78      amostral78
amostra79      amostral79
amostra80      amostral80
amostra81      amostral81
amostra82      amostral82
amostra83      amostral83
amostra84      amostral84
amostra85      amostral85
amostra86      amostral86
amostra87      amostral87
amostra88      amostral88
amostra89      amostral89
amostra90      amostral90
amostra91      amostral91
amostra92      amostral92
amostra93      amostral93
amostra94      amostral94
amostra95      amostral95
amostra96      amostral96
amostra97      amostral97
amostra98      amostral98
amostra99      amostral99
amostra100     amostra200;
run;
```

```
proc delete data=amostral1;
proc delete data=amostra2;
proc delete data=amostra3;
proc delete data=amostra4;
proc delete data=amostra5;
proc delete data=amostra6;
```

```
proc delete data=amostra7;  
proc delete data=amostra8;  
proc delete data=amostra9;  
proc delete data=amostra10;  
proc delete data=amostra11;  
proc delete data=amostra12;  
proc delete data=amostra13;  
proc delete data=amostra14;  
proc delete data=amostra15;  
proc delete data=amostra16;  
proc delete data=amostra17;  
proc delete data=amostra18;  
proc delete data=amostra19;  
proc delete data=amostra20;  
proc delete data=amostra21;  
proc delete data=amostra22;  
proc delete data=amostra23;  
proc delete data=amostra24;  
proc delete data=amostra25;  
proc delete data=amostra26;  
proc delete data=amostra27;  
proc delete data=amostra28;  
proc delete data=amostra29;  
proc delete data=amostra30;  
proc delete data=amostra31;  
proc delete data=amostra32;  
proc delete data=amostra33;  
proc delete data=amostra34;  
proc delete data=amostra35;  
proc delete data=amostra36;  
proc delete data=amostra37;  
proc delete data=amostra38;  
proc delete data=amostra39;  
proc delete data=amostra40;  
proc delete data=amostra41;  
proc delete data=amostra42;  
proc delete data=amostra43;  
proc delete data=amostra44;  
proc delete data=amostra45;  
proc delete data=amostra46;
```

```
proc delete data=amostra47;  
proc delete data=amostra48;  
proc delete data=amostra49;  
proc delete data=amostra50;  
proc delete data=amostra51;  
proc delete data=amostra52;  
proc delete data=amostra53;  
proc delete data=amostra54;  
proc delete data=amostra55;  
proc delete data=amostra56;  
proc delete data=amostra57;  
proc delete data=amostra58;  
proc delete data=amostra59;  
proc delete data=amostra60;  
proc delete data=amostra61;  
proc delete data=amostra62;  
proc delete data=amostra63;  
proc delete data=amostra64;  
proc delete data=amostra65;  
proc delete data=amostra66;  
proc delete data=amostra67;  
proc delete data=amostra68;  
proc delete data=amostra69;  
proc delete data=amostra70;  
proc delete data=amostra71;  
proc delete data=amostra72;  
proc delete data=amostra73;  
proc delete data=amostra74;  
proc delete data=amostra75;  
proc delete data=amostra76;  
proc delete data=amostra77;  
proc delete data=amostra78;  
proc delete data=amostra79;  
proc delete data=amostra80;  
proc delete data=amostra81;  
proc delete data=amostra82;  
proc delete data=amostra83;  
proc delete data=amostra84;  
proc delete data=amostra85;  
proc delete data=amostra86;
```

```
proc delete data=amostra87;  
proc delete data=amostra88;  
proc delete data=amostra89;  
proc delete data=amostra90;  
proc delete data=amostra91;  
proc delete data=amostra92;  
proc delete data=amostra93;  
proc delete data=amostra94;  
proc delete data=amostra95;  
proc delete data=amostra96;  
proc delete data=amostra97;  
proc delete data=amostra98;  
proc delete data=amostra99;  
proc delete data=amostra100;  
proc delete data=amostra101;  
proc delete data=amostra102;  
proc delete data=amostra103;  
proc delete data=amostra104;  
proc delete data=amostra105;  
proc delete data=amostra106;  
proc delete data=amostra107;  
proc delete data=amostra108;  
proc delete data=amostra109;  
proc delete data=amostra110;  
proc delete data=amostra111;  
proc delete data=amostra112;  
proc delete data=amostra113;  
proc delete data=amostra114;  
proc delete data=amostra115;  
proc delete data=amostra116;  
proc delete data=amostra117;  
proc delete data=amostra118;  
proc delete data=amostra119;  
proc delete data=amostra120;  
proc delete data=amostra121;  
proc delete data=amostra122;  
proc delete data=amostra123;  
proc delete data=amostra124;  
proc delete data=amostra125;  
proc delete data=amostra126;
```

```
proc delete data=amostra127;  
proc delete data=amostra128;  
proc delete data=amostra129;  
proc delete data=amostra130;  
proc delete data=amostra131;  
proc delete data=amostra132;  
proc delete data=amostra133;  
proc delete data=amostra134;  
proc delete data=amostra135;  
proc delete data=amostra136;  
proc delete data=amostra137;  
proc delete data=amostra138;  
proc delete data=amostra139;  
proc delete data=amostra140;  
proc delete data=amostra141;  
proc delete data=amostra142;  
proc delete data=amostra143;  
proc delete data=amostra144;  
proc delete data=amostra145;  
proc delete data=amostra146;  
proc delete data=amostra147;  
proc delete data=amostra148;  
proc delete data=amostra149;  
proc delete data=amostra150;  
proc delete data=amostra151;  
proc delete data=amostra152;  
proc delete data=amostra153;  
proc delete data=amostra154;  
proc delete data=amostra155;  
proc delete data=amostra156;  
proc delete data=amostra157;  
proc delete data=amostra158;  
proc delete data=amostra159;  
proc delete data=amostra160;  
proc delete data=amostra161;  
proc delete data=amostra162;  
proc delete data=amostra163;  
proc delete data=amostra164;  
proc delete data=amostra165;  
proc delete data=amostra166;
```



```
proc delete data=amostra167;
proc delete data=amostra168;
proc delete data=amostra169;
proc delete data=amostra170;
proc delete data=amostra171;
proc delete data=amostra172;
proc delete data=amostra173;
proc delete data=amostra174;
proc delete data=amostra175;
proc delete data=amostra176;
proc delete data=amostra177;
proc delete data=amostra178;
proc delete data=amostra179;
proc delete data=amostra180;
proc delete data=amostra181;
proc delete data=amostra182;
proc delete data=amostra183;
proc delete data=amostra184;
proc delete data=amostra185;
proc delete data=amostra186;
proc delete data=amostra187;
proc delete data=amostra188;
proc delete data=amostra189;
proc delete data=amostra190;
proc delete data=amostra191;
proc delete data=amostra192;
proc delete data=amostra193;
proc delete data=amostra194;
proc delete data=amostra195;
proc delete data=amostra196;
proc delete data=amostra197;
proc delete data=amostra198;
proc delete data=amostra199;
proc delete data=amostra200;

proc export data= in2.boot
            outfile= "boot"
            dbms=excel2000 replace;

run;
```

```

proc sort data=in2.boot;
  by amostra;
run;

proc means data=in2.boot mean var max noprint;
  var amostra
  f1_v01  f1_v02  f1_v03  f1_v04  f1_v05
  f1_v06
  f2_v01  f2_v02  f2_v03  f2_v04  f2_v05
  f2_v06
;
  output out=saida(drop=nada1-nada2)
         mean=nada1
  m_f1_v1  m_f1_v2  m_f1_v3  m_f1_v4  m_f1_v5
  m_f1_v6
  m_f2_v1  m_f2_v2  m_f2_v3  m_f2_v4  m_f2_v5
  m_f2_v6
         var =nada2
  v_f1_v1  v_f1_v2  v_f1_v3  v_f1_v4  v_f1_v5
  v_f1_v6
  v_f2_v1  v_f2_v2  v_f2_v3  v_f2_v4  v_f2_v5
  v_f2_v6
         max =B;
  title'Media, Variancia e Tamanho da Amostra - Topicos 2
e 4';
run;

%macro alfa(alfa);
data saidaK(keep=LI LS _type_);
  set saida;
  LI=ROUND(B*&alfa/2); *limite inferior do intervalo
Bootstrap*;
  LS=ROUND((B-LI)+1); *limite superior do intervalo
Bootstrap*;
run;
%mend alfa;
%alfa(0.05); *confiança*;

%macro valor_limite(k,j);
proc sort data=in2.boot;

```

```

by f&k._&j;

data boot&k(keep=n _type_ f&k._&j);
  set in2.boot;
  _type_=0;
  n=_n_;
run;

data limite;
  merge saidaK boot&k;
  by _type_;
  if LI=n then LI_Boot_F&k._&j=f&k._&j;
  else LI_Boot_F&k._&j=.;
  if LS=n then LS_Boot_F&k._&j=f&k._&j;
  else LS_Boot_F&k._&j=.;
run;

proc means data=limite max noprint;
var LI_Boot_F&k._&j LS_Boot_F&k._&j;
  output out=limites_f&k._&j
  max=LI_Boot LS_Boot;
run;
proc delete data=limite;
proc delete data=boot&k;
run;
%mend valor_limite;
%valor_limite(1,v01 );
%valor_limite(1,v02 );
%valor_limite(1,v03 );
%valor_limite(1,v04 );
%valor_limite(1,v05 );
%valor_limite(1,v06 );
%valor_limite(2,v01 );
%valor_limite(2,v02 );
%valor_limite(2,v03 );
%valor_limite(2,v04 );
%valor_limite(2,v05 );
%valor_limite(2,v06 );

data limites;

```

```

set
  limites_f1_v01  limites_f1_v02  limites_f1_v03
limites_f1_v04  limites_f1_v05
  limites_f1_v06
  limites_f2_v01  limites_f2_v02  limites_f2_v03
limites_f2_v04  limites_f2_v05
  limites_f2_v06;

      if _n_=1 then _name_='f1_v01';
else if _n_=2 then _name_='f1_v02';
else if _n_=3 then _name_='f1_v03';
else if _n_=4 then _name_='f1_v04';
else if _n_=5 then _name_='f1_v05';
else if _n_=6 then _name_='f1_v06';
else if _n_=7 then _name_='f2_v01';
else if _n_=8 then _name_='f2_v02';
else if _n_=9 then _name_='f2_v03';
else if _n_=10 then _name_='f2_v04';
else if _n_=11 then _name_='f2_v05';
else if _n_=12 then _name_='f2_v06';

run;

proc delete data=limites_f1_v01;
proc delete data=limites_f1_v02;
proc delete data=limites_f1_v03;
proc delete data=limites_f1_v04;
proc delete data=limites_f1_v05;
proc delete data=limites_f1_v06;
proc delete data= limites_f2_v01;
proc delete data=limites_f2_v02;
proc delete data=limites_f2_v03;
proc delete data=limites_f2_v04;
proc delete data=limites_f2_v05;
proc delete data= limites_f2_v06;
run;

proc factor data=base
           method=principal
           rotate=Varimax

```

```

                outstat=perfil
                nfactores=2 noprint;
var V1-V6;
run;

data perfil_f1(keep=i f1_v01-f1_v06)
    perfil_f2(keep=i f2_v01-f2_v06);
set perfil;
if _type_='UNROTATE' and _name_ in ('Factor1'
'Factor2');

if _name_='Factor1' then do;
    f1_v01= v1 ;    f1_v02= v2 ;    f1_v03= v3 ;    f1_v04=
v4 ;    f1_v05= v5 ;
    f1_v06= v6 ;
end;
else if _name_='Factor2' then do;
    f2_v01= v1 ;    f2_v02= v2 ;    f2_v03= v3 ;    f2_v04=
v4 ;    f2_v05= v5 ;
    f2_v06= v6 ;
end;
    i=1;

    if _name_='Factor1' then output perfil_f1;
else if _name_='Factor2' then output perfil_f2;
run;

data Total(drop=i);
merge perfil_f1 perfil_f2;
by i;
run;

proc delete data=perfil;
proc delete data=perfil_f1;
proc delete data=perfil_f2;
run;

proc transpose data=total out=fator;
var f1_v01-f1_v06 f2_v01-f2_v06;
run;

```

```

data fator;
  set fator;
  rename coll=cargas;
run;

proc transpose data=saida out=medias;
  var m_f1_v1-m_f1_v6 m_f2_v1-m_f2_v6;
  id _type_;
run;

proc transpose data=saida out=variâncias;
  var v_f1_v1-v_f1_v6 v_f2_v1-v_f2_v6;
  id _type_;
run;

data medias;
  set medias;
  rename _0=media;

  if _name_='m_f1_v1' then _name_='f1_v01';
  if _name_='m_f1_v2' then _name_='f1_v02';
  if _name_='m_f1_v3' then _name_='f1_v03';
  if _name_='m_f1_v4' then _name_='f1_v04';
  if _name_='m_f1_v5' then _name_='f1_v05';
  if _name_='m_f1_v6' then _name_='f1_v06';
  if _name_='m_f2_v1' then _name_='f2_v01';
  if _name_='m_f2_v2' then _name_='f2_v02';
  if _name_='m_f2_v3' then _name_='f2_v03';
  if _name_='m_f2_v4' then _name_='f2_v04';
  if _name_='m_f2_v5' then _name_='f2_v05';
  if _name_='m_f2_v6' then _name_='f2_v06';
run;

data variâncias;
  set variâncias;
  rename _0=variância;
  if _name_='v_f1_v1' then _name_='f1_v01';
  if _name_='v_f1_v2' then _name_='f1_v02';
  if _name_='v_f1_v3' then _name_='f1_v03';

```

```

if _name_='v_f1_v4' then _name_='f1_v04';
if _name_='v_f1_v5' then _name_='f1_v05';
if _name_='v_f1_v6' then _name_='f1_v06';
if _name_='v_f2_v1 ' then _name_='f2_v01';
if _name_='v_f2_v2 ' then _name_='f2_v02';
if _name_='v_f2_v3 ' then _name_='f2_v03';
if _name_='v_f2_v4 ' then _name_='f2_v04';
if _name_='v_f2_v5 ' then _name_='f2_v05';
if _name_='v_f2_v6 ' then _name_='f2_v06';
run;
proc sort data=limites;
  by _name_;
proc sort data=fator;
  by _name_;
proc sort data=medias;
  by _name_;
proc sort data=variâncias;
  by _name_;

data total_5_6(keep=fator_variavel cargas media variancia
                LI_Boot LS_Boot sig_ic vies
erro_medio g);
merge fator medias variâncias limites;
  by _name_;
  rename _name_=fator_variavel;
  vies=cargas-media;
  erro_medio=variância+vies**2;
  zero=0;
  if LI_Boot<=zero<=LS_Boot then sig_ic='Nao_Signif';
  else sig_ic='Signif      ';
  g=cargas/(variância**(1/2));

run;

proc transpose data=total_5_6
out=g_total_5_6(drop=_name_);
  var g;
  id fator_variavel;
run;

```

```

data g_total_5_6;
  set g_total_5_6;
  _type_=0;
  rename f1_v01=g_f1_v01;
  rename f1_v02=g_f1_v02;
  rename f1_v03=g_f1_v03;
  rename f1_v04=g_f1_v04;
  rename f1_v05=g_f1_v05;
  rename f1_v06=g_f1_v06;
  rename f2_v01=g_f2_v01;
  rename f2_v02=g_f2_v02;
  rename f2_v03=g_f2_v03;
  rename f2_v04=g_f2_v04;
  rename f2_v05=g_f2_v05;
  rename f2_v06=g_f2_v06;
run;

data boot;
  set in2.boot;
  _type_=0;
run;

data boot_norm(keep=norm_f1_v01-norm_f1_v06
                 norm_f2_v01-norm_f2_v06
                 g_f1_v01-g_f1_v06
                 g_f2_v01-g_f2_v06);
merge boot saida g_total_5_6;
  by _type_;

norm_f1_v01=(f1_v01-m_f1_v1)/(v_f1_v1**(1/2));
norm_f1_v02=(f1_v02-m_f1_v2)/(v_f1_v2**(1/2));
norm_f1_v03=(f1_v03-m_f1_v3)/(v_f1_v3**(1/2));
norm_f1_v04=(f1_v04-m_f1_v4)/(v_f1_v4**(1/2));
norm_f1_v05=(f1_v05-m_f1_v5)/(v_f1_v5**(1/2));
norm_f1_v06=(f1_v06-m_f1_v6)/(v_f1_v6**(1/2));
norm_f2_v01=(f2_v01-m_f2_v1)/(v_f2_v1**(1/2));
norm_f2_v02=(f2_v02-m_f2_v2)/(v_f2_v2**(1/2));
norm_f2_v03=(f2_v03-m_f2_v3)/(v_f2_v3**(1/2));
norm_f2_v04=(f2_v04-m_f2_v4)/(v_f2_v4**(1/2));
norm_f2_v05=(f2_v05-m_f2_v5)/(v_f2_v5**(1/2));

```



```
norm_f2_v06=(f2_v06-m_f2_v6)/(v_f2_v6**(1/2));
```

```
run;
```

```
%macro pvalor(p);
```

```
data s_&p(keep=s_&p);
```

```
set boot_norm;
```

```
if (g_&p>0 and norm_&p>g_&p) or
```

```
(g_&p<0 and norm_&p<g_&p) then s_&p=1;
```

```
else s_&p=0;
```

```
run;
```

```
proc means data=s_&p sum noprint;
```

```
var s_&p;
```

```
output out=saida_&p
```

```
sum=;
```

```
run;
```

```
proc delete data=s_&p;
```

```
data saida_&p(keep=fator_variavel p_valor);
```

```
set saida_&p;
```

```
fator_variavel="&p";
```

```
p_valor=s_&p/_freq;
```

```
run;
```

```
%mend pvalor;
```

```
%pvalor(f1_v01);
```

```
%pvalor(f1_v02);
```

```
%pvalor(f1_v03);
```

```
%pvalor(f1_v04);
```

```
%pvalor(f1_v05);
```

```
%pvalor(f1_v06);
```

```
%pvalor(f2_v01);
```

```
%pvalor(f2_v02);
```

```
%pvalor(f2_v03);
```

```
%pvalor(f2_v04);
```

```
%pvalor(f2_v05);
```

```
%pvalor(f2_v06);
```

```

data p_valor;
  set saida_f1_v01 saida_f1_v02 saida_f1_v03 saida_f1_v04
saida_f1_v05
      saida_f1_v06 saida_f2_v01 saida_f2_v02 saida_f2_v03
saida_f2_v04
      saida_f2_v05 saida_f2_v06;
  if p_valor>0.05 then sig_vp='Nao_Signif';
  else
      sig_vp='Signif      ';

run;

proc delete data=saida_f1_v01;
proc delete data=saida_f1_v02;
proc delete data=saida_f1_v03;
proc delete data=saida_f1_v04;
proc delete data=saida_f1_v05;
proc delete data=saida_f1_v06;
proc delete data=saida_f2_v01;
proc delete data=saida_f2_v02;
proc delete data=saida_f2_v03;
proc delete data=saida_f2_v04;
proc delete data=saida_f2_v05;
proc delete data=saida_f2_v06;
run;

data total_5_6(drop=g);
  merge total_5_6 p_valor;
  by fator_variavel;
run;

proc print data=total_5_6;
  title'Analise Fatorial - Bootstrap';
run;

```

Anexo 3

O Programa Jackknife em SAS V.8

O programa abaixo referê-se ao usado com a base de dados “Creme Dental”.

```
libname in2 'C:\Tese Giovani\Terceira Fase';
options nodate ls=max ps=max;
proc import out= in2.basecremedental
            datafile=
            "C:\Tese Giovani\Terceira Fase\basecremedental.xls"
            dbms=excel2000 replace;
            getnames=yes;
run;

data base;
  set in2.basecremedental;
run;
%macro sel(i);
data base&i;
  set base;
  if _n_=&i then delete;
run;

proc factor data=base&i
            method=principal
            rotate=Varimax
            outstat=perfil&i
            nfactors=2 noprint;

  var V1-V6;
run;

data perfil_f1_&i(keep=i f1_v01-f1_v06)
  perfil_f2_&i(keep=i f2_v01-f2_v06);
  set perfil&i;
  if _type_='UNROTATE' and _name_ in ('Factor1'
  'Factor2');

  if _name_='Factor1' then do;
```

```

        f1_v01= v1 ;    f1_v02= v2 ;    f1_v03= v3 ;    f1_v04=
v4 ;    f1_v05= v5 ;    f1_v06= v6 ;
    end;
    else if _name_='Factor2' then do;
        f2_v01= v1 ;    f2_v02= v2 ;    f2_v03= v3 ;    f2_v04=
v4 ;    f2_v05= v5 ;    f2_v06= v6 ;
    end;
        i=1;

        if _name_='Factor1' then output perfil_f1_&i;
    else if _name_='Factor2' then output perfil_f2_&i;
run;

data amostra&i(drop=i);
    merge perfil_f1_&i perfil_f2_&i;
    by i;
    amostra=&i;
run;

proc delete data=perfil&i;
proc delete data=perfil_f1_&i;
proc delete data=perfil_f2_&i;
proc delete data=base&i;
run;

%mend sel;
%sel (1 ) ; %sel (2 ) ;
%sel (3 ) ; %sel (4 ) ;
%sel (5 ) ; %sel (6 ) ;
%sel (7 ) ; %sel (8 ) ;
%sel (9 ) ; %sel (10 ) ;
%sel (11 ) ; %sel (12 ) ;
%sel (13 ) ; %sel (14 ) ;
%sel (15 ) ; %sel (16 ) ;
%sel (17 ) ; %sel (18 ) ;
%sel (19 ) ; %sel (20 ) ;
%sel (21 ) ; %sel (22 ) ;
%sel (23 ) ; %sel (24 ) ;
%sel (25 ) ; %sel (26 ) ;
%sel (27 ) ; %sel (28 ) ;

```

```
%sel (29 ) ; %sel (30 ) ;  
%sel (31 ) ; %sel (32 ) ;  
%sel (33 ) ; %sel (34 ) ;  
%sel (35 ) ; %sel (36 ) ;  
%sel (37 ) ; %sel (38 ) ;  
%sel (39 ) ; %sel (40 ) ;
```

```
data in2.jack;
```

```
set
```

```
amostra1  
amostra2  
amostra3  
amostra4  
amostra5  
amostra6  
amostra7  
amostra8  
amostra9  
amostra10  
amostra11  
amostra12  
amostra13  
amostra14  
amostra15  
amostra16  
amostra17  
amostra18  
amostra19  
amostra20  
amostra21  
amostra22  
amostra23  
amostra24  
amostra25  
amostra26  
amostra27  
amostra28  
amostra29  
amostra30  
amostra31
```

```
amostra32  
amostra33  
amostra34  
amostra35  
amostra36  
amostra37  
amostra38  
amostra39  
amostra40;  
run;
```

```
proc delete data=amostra1;  
proc delete data=amostra2;  
proc delete data=amostra3;  
proc delete data=amostra4;  
proc delete data=amostra5;  
proc delete data=amostra6;  
proc delete data=amostra7;  
proc delete data=amostra8;  
proc delete data=amostra9;  
proc delete data=amostra10;  
proc delete data=amostra11;  
proc delete data=amostra12;  
proc delete data=amostra13;  
proc delete data=amostra14;  
proc delete data=amostra15;  
proc delete data=amostra16;  
proc delete data=amostra17;  
proc delete data=amostra18;  
proc delete data=amostra19;  
proc delete data=amostra20;  
proc delete data=amostra21;  
proc delete data=amostra22;  
proc delete data=amostra23;  
proc delete data=amostra24;  
proc delete data=amostra25;  
proc delete data=amostra26;  
proc delete data=amostra27;  
proc delete data=amostra28;  
proc delete data=amostra29;
```

```

proc delete data=amostra30;
proc delete data=amostra31;
proc delete data=amostra32;
proc delete data=amostra33;
proc delete data=amostra34;
proc delete data=amostra35;
proc delete data=amostra36;
proc delete data=amostra37;
proc delete data=amostra38;
proc delete data=amostra39;
proc delete data=amostra40;

proc export data= in2.jack
            outfile= "jack"
            dbms=excel2000 replace;

run;

proc sort data=in2.jack;
  by amostra;
run;

proc means data=in2.jack mean var max noprint;
  var amostra
  f1_v01  f1_v02  f1_v03  f1_v04  f1_v05  f1_v06
  f2_v01  f2_v02  f2_v03  f2_v04  f2_v05  f2_v06;
  output out=saida(drop=nada1-nada2)
         mean=nada1
  m_f1_v1  m_f1_v2  m_f1_v3  m_f1_v4  m_f1_v5  m_f1_v6
  m_f2_v1  m_f2_v2  m_f2_v3  m_f2_v4  m_f2_v5  m_f2_v6

         var =nada2
  v_f1_v1  v_f1_v2  v_f1_v3  v_f1_v4  v_f1_v5  v_f1_v6
  v_f2_v1  v_f2_v2  v_f2_v3  v_f2_v4  v_f2_v5  v_f2_v6
  max =B;
  title'Media, Variancia e Tamanho da Amostra - Topicos 2
e 4';
run;

%macro alfa(alfa);
data saidaK(keep=LI LS _type_);

```

```

set saida;
  LI=ROUND(B*&alfa/2); *limite inferior do intervalo
Bootstrap*;
  LS=ROUND((B-LI)+1); *limite superior do intervalo
Bootstrap*;
run;
%mend alfa;
%alfa(0.05); *confiança*;

%macro valor_limite(k,j);
proc sort data=in2.jack;
  by f&k._&j;

data boot&k(keep=n _type_ f&k._&j);
  set in2.jack;
  _type_=0;
  n=_n_;
run;

data limite;
  merge saidaK boot&k;
  by _type_;
  if LI=n then LI_Jack_F&k._&j=f&k._&j;
  else      LI_Jack_F&k._&j=.;
  if LS=n then LS_Jack_F&k._&j=f&k._&j;
  else      LS_Jack_F&k._&j=.;
run;

proc means data=limite max noprint;
var LI_Jack_F&k._&j LS_Jack_F&k._&j;
  output out=limites_f&k._&j
         max=LI_Jack LS_Jack;
run;
proc delete data=limite;
proc delete data=boot&k;
run;
%mend valor_limite;
%valor_limite(1,v01 );
%valor_limite(1,v02 );
%valor_limite(1,v03 );

```



```

%valor_limite(1,v04 );
%valor_limite(1,v05 );
%valor_limite(1,v06 );
%valor_limite(2,v01 );
%valor_limite(2,v02 );
%valor_limite(2,v03 );
%valor_limite(2,v04 );
%valor_limite(2,v05 );
%valor_limite(2,v06 );

data limites;
  set
    limites_f1_v01  limites_f1_v02  limites_f1_v03
    limites_f1_v04  limites_f1_v05
    limites_f1_v06  limites_f2_v01  limites_f2_v02
    limites_f2_v03  limites_f2_v04  limites_f2_v05
    limites_f2_v06;

    if _n_=1 then _name_='f1_v01';
else if _n_=2 then _name_='f1_v02';
else if _n_=3 then _name_='f1_v03';
else if _n_=4 then _name_='f1_v04';
else if _n_=5 then _name_='f1_v05';
else if _n_=6 then _name_='f1_v06';
else if _n_=7 then _name_='f2_v01';
else if _n_=8 then _name_='f2_v02';
else if _n_=9 then _name_='f2_v03';
else if _n_=10 then _name_='f2_v04';
else if _n_=11 then _name_='f2_v05';
else if _n_=12 then _name_='f2_v06';
run;

proc delete data=limites_f1_v01;
proc delete data=limites_f1_v02;
proc delete data=limites_f1_v03;
proc delete data=limites_f1_v04;
proc delete data=limites_f1_v05;
proc delete data=limites_f1_v06;
proc delete data= limites_f2_v01;
proc delete data=limites_f2_v02;

```

```

proc delete data=limites_f2_v03;
proc delete data=limites_f2_v04;
proc delete data=limites_f2_v05;
proc delete data= limites_f2_v06;

run;

proc factor data=base
           method=principal
           rotate=Varimax
           outstat=perfil
           nfactors=2 noprint;

var V1-V6;
run;

data perfil_f1(keep=i f1_v01-f1_v06)
  perfil_f2(keep=i f2_v01-f2_v06);
set perfil;
if _name_='Factor1' then do;
  f1_v01= v1 ;   f1_v02= v2 ;   f1_v03= v3 ;   f1_v04=
v4 ;   f1_v05= v5 ;
  f1_v06= v6 ;
end;
else if _name_='Factor2' then do;
  f2_v01= v1 ;   f2_v02= v2 ;   f2_v03= v3 ;   f2_v04=
v4 ;   f2_v05= v5 ;
  f2_v06= v6 ;
end;
i=1;

if _name_='Factor1' then output perfil_f1;
else if _name_='Factor2' then output perfil_f2;
run;

data Total(drop=i);
merge perfil_f1 perfil_f2;
by i;
run;

proc delete data=perfil;

```

```

proc delete data=perfil_f1;
proc delete data=perfil_f2;
run;

proc transpose data=total out=fator;
  var f1_v01-f1_v06 f2_v01-f2_v06;
run;

data fator;
  set fator;
  rename coll=cargas;
run;

proc transpose data=saida out=medias;
  var m_f1_v1-m_f1_v6 m_f2_v1-m_f2_v6;
  id _type_;
run;

proc transpose data=saida out=variâncias;
  var v_f1_v1-v_f1_v6 v_f2_v1-v_f2_v6;
  id _type_;
run;

data medias;
  set medias;
  rename _0=media;

  if _name_='m_f1_v1' then _name_='f1_v01';
  if _name_='m_f1_v2' then _name_='f1_v02';
  if _name_='m_f1_v3' then _name_='f1_v03';
  if _name_='m_f1_v4' then _name_='f1_v04';
  if _name_='m_f1_v5' then _name_='f1_v05';
  if _name_='m_f1_v6' then _name_='f1_v06';
  if _name_='m_f2_v1' then _name_='f2_v01';
  if _name_='m_f2_v2' then _name_='f2_v02';
  if _name_='m_f2_v3' then _name_='f2_v03';
  if _name_='m_f2_v4' then _name_='f2_v04';
  if _name_='m_f2_v5' then _name_='f2_v05';
  if _name_='m_f2_v6' then _name_='f2_v06';
run;

```

```

data variancias;
set variancias;
  rename _0=variancia;
  if _name_='v_f1_v1' then _name_='f1_v01';
  if _name_='v_f1_v2' then _name_='f1_v02';
  if _name_='v_f1_v3' then _name_='f1_v03';
  if _name_='v_f1_v4' then _name_='f1_v04';
  if _name_='v_f1_v5' then _name_='f1_v05';
  if _name_='v_f1_v6' then _name_='f1_v06';
  if _name_='v_f2_v1' then _name_='f2_v01';
  if _name_='v_f2_v2' then _name_='f2_v02';
  if _name_='v_f2_v3' then _name_='f2_v03';
  if _name_='v_f2_v4' then _name_='f2_v04';
  if _name_='v_f2_v5' then _name_='f2_v05';
  if _name_='v_f2_v6' then _name_='f2_v06';
  run;
proc sort data=limites;
  by _name_;
proc sort data=fator;
  by _name_;
proc sort data=medias;
  by _name_;
proc sort data=variancias;
  by _name_;

data total_5_6(keep=fator_variavel cargas media variancia
                LI_Jack LS_Jack sig_ic vies
erro_medio g);
merge fator medias variancias limites;
  by _name_;
  rename _name_=fator_variavel;
  vies=cargas-media;
  erro_medio=variancia+vies**2;
  zero=0;
  if LI_Jack<=zero<=LS_Jack then sig_ic='Nao_Signif';
  else sig_ic='Signif';
  g=cargas/(variancia**(1/2));

run;

```

```

proc transpose data=total_5_6
out=g_total_5_6(drop=_name_);
  var g;
  id fator_variavel;
run;

data g_total_5_6;
set g_total_5_6;
  _type_=0;
  rename f1_v01=g_f1_v01;
  rename f1_v02=g_f1_v02;
  rename f1_v03=g_f1_v03;
  rename f1_v04=g_f1_v04;
  rename f1_v05=g_f1_v05;
  rename f1_v06=g_f1_v06;
  rename f2_v01=g_f2_v01;
  rename f2_v02=g_f2_v02;
  rename f2_v03=g_f2_v03;
  rename f2_v04=g_f2_v04;
  rename f2_v05=g_f2_v05;
  rename f2_v06=g_f2_v06;
run;

data jack;
set in2.jack;
  _type_=0;
run;

data jack_norm(keep=norm_f1_v01-norm_f1_v06
                 norm_f2_v01-norm_f2_v06
                 g_f1_v01-g_f1_v06
                 g_f2_v01-g_f2_v06);
merge jack saida g_total_5_6;
  by _type_;

norm_f1_v01=(f1_v01-m_f1_v1)/(v_f1_v1**(1/2));
norm_f1_v02=(f1_v02-m_f1_v2)/(v_f1_v2**(1/2));
norm_f1_v03=(f1_v03-m_f1_v3)/(v_f1_v3**(1/2));
norm_f1_v04=(f1_v04-m_f1_v4)/(v_f1_v4**(1/2));
norm_f1_v05=(f1_v05-m_f1_v5)/(v_f1_v5**(1/2));

```

```

norm_f1_v06=(f1_v06-m_f1_v6)/(v_f1_v6**(1/2));
norm_f2_v01=(f2_v01-m_f2_v1)/(v_f2_v1**(1/2));
norm_f2_v02=(f2_v02-m_f2_v2)/(v_f2_v2**(1/2));
norm_f2_v03=(f2_v03-m_f2_v3)/(v_f2_v3**(1/2));
norm_f2_v04=(f2_v04-m_f2_v4)/(v_f2_v4**(1/2));
norm_f2_v05=(f2_v05-m_f2_v5)/(v_f2_v5**(1/2));
norm_f2_v06=(f2_v06-m_f2_v6)/(v_f2_v6**(1/2));

```

```
run;
```

```
%macro pvalor(p);
```

```

data s_&p(keep=s_&p);
  set jack_norm;
  if (g_&p>0 and norm_&p>g_&p) or
      (g_&p<0 and norm_&p<g_&p) then s_&p=1;
  else s_&p=0;
run;

```

```

proc means data=s_&p sum noprint;
  var s_&p;
  output out=saida_&p
         sum=;
run;

```

```
proc delete data=s_&p;
```

```

data saida_&p(keep=fator_variavel p_valor);
  set saida_&p;
  fator_variavel="&p";
  p_valor=s_&p/_freq_;
run;

```

```
%mend pvalor;
```

```

%pvalor(f1_v01);
%pvalor(f1_v02);
%pvalor(f1_v03);
%pvalor(f1_v04);
%pvalor(f1_v05);
%pvalor(f1_v06);

```

```

%pvalor(f2_v01);
%pvalor(f2_v02);
%pvalor(f2_v03);
%pvalor(f2_v04);
%pvalor(f2_v05);
%pvalor(f2_v06);

data p_valor;
  set saida_f1_v01 saida_f1_v02 saida_f1_v03 saida_f1_v04
saida_f1_v05
      saida_f1_v06 saida_f2_v01 saida_f2_v02 saida_f2_v03
saida_f2_v04
      saida_f2_v05 saida_f2_v06;
  if p_valor>0.05 then sig_vp='Nao_Signif';
  else sig_vp='Signif';

run;

proc delete data=saida_f1_v01;
proc delete data=saida_f1_v02;
proc delete data=saida_f1_v03;
proc delete data=saida_f1_v04;
proc delete data=saida_f1_v05;
proc delete data=saida_f1_v06;
proc delete data=saida_f2_v01;
proc delete data=saida_f2_v02;
proc delete data=saida_f2_v03;
proc delete data=saida_f2_v04;
proc delete data=saida_f2_v05;
proc delete data=saida_f2_v06;
run;

data total_5_6(drop=g);
  merge total_5_6 p_valor;
  by fator_variavel;
run;

proc print data=total_5_6;
  title'Analise Fatorial - Jackknife';
run;

```

Anexo 4

Tabelas dos Testes de Convergência

Tabela 10: Caso 1: Base Busca de Fatores-EMQ - Fator 1

Variáveis	B=25	B=100	B=200	B=1000
V1	1,08E-02	1,12E-02	1,12E-02	1,16E-02
V2	8,75E-03	9,00E-03	9,00E-03	9,46E-03
V3	9,93E-03	5,15E-03	5,15E-03	5,43E-03
V4	2,95E-03	1,65E-03	1,65E-03	1,33E-03
V5	2,86E-03	2,02E-03	2,02E-03	1,91E-03
V6	8,55E-03	6,01E-03	6,01E-03	5,58E-03
V7	1,43E-03	2,04E-03	2,04E-03	2,15E-03
V8	4,14E-03	5,11E-03	5,11E-03	4,99E-03
V9	5,95E-03	5,52E-03	5,52E-03	6,72E-03
V10	8,20E-03	6,07E-03	6,07E-03	5,76E-03
V11	5,88E-03	5,45E-03	5,45E-03	5,30E-03
V12	1,23E-02	1,34E-02	1,34E-02	1,51E-02
V13	6,60E-03	7,24E-03	7,24E-03	6,05E-03
V14	9,45E-03	8,11E-03	8,11E-03	7,17E-03
V15	1,42E-03	1,20E-03	1,20E-03	1,31E-03
V16	6,06E-03	3,70E-03	3,70E-03	3,62E-03
V17	6,50E-03	3,88E-03	3,88E-03	3,77E-03
V18	3,11E-03	3,84E-03	3,84E-03	4,64E-03
V19	1,93E-03	1,81E-03	1,81E-03	1,91E-03
V20	6,16E-03	5,18E-03	5,18E-03	5,52E-03
V21	1,32E-02	1,33E-02	1,33E-02	1,61E-02
V22	1,44E-03	1,84E-03	1,84E-03	1,80E-03
V23	6,02E-03	6,25E-03	6,25E-03	1,16E-02

Tabela 11: Caso 1 : Base Busca de Fatores-EMQ - Fator 2

Variáveis	B=25	B=100	B=200	B=1000
V1	7,79E-03	8,80E-03	8,44E-03	6,93E-03
V2	8,14E-03	1,18E-02	1,28E-02	8,06E-03
V3	1,11E-02	9,81E-03	1,02E-02	1,31E-02
V4	5,75E-03	5,58E-03	5,28E-03	8,98E-03
V5	6,97E-03	7,77E-03	7,76E-03	5,87E-03
V6	1,05E-02	8,81E-03	9,18E-03	7,20E-03
V7	6,75E-03	9,41E-03	8,06E-03	9,07E-03
V8	1,44E-02	1,28E-02	1,20E-02	8,34E-03
V9	1,80E-02	1,42E-02	1,49E-02	1,11E-02
V10	8,14E-03	8,07E-03	7,28E-03	1,41E-02
V11	7,72E-03	8,38E-03	7,62E-03	6,87E-03
V12	6,53E-03	3,55E-03	3,00E-03	7,20E-03
V13	1,40E-02	9,47E-03	8,75E-03	3,30E-03
V14	9,86E-03	6,61E-03	7,13E-03	9,37E-03
V15	5,57E-03	4,05E-03	4,86E-03	7,80E-03
V16	1,13E-02	8,64E-03	7,32E-03	4,74E-03
V17	1,32E-02	1,13E-02	1,04E-02	7,70E-03
V18	7,09E-03	8,09E-03	8,04E-03	9,47E-03
V19	8,71E-03	7,49E-03	7,12E-03	8,17E-03
V20	1,20E-02	1,03E-02	9,55E-03	6,17E-03
V21	5,90E-03	3,35E-03	2,69E-03	9,67E-03
V22	1,02E-02	7,22E-03	8,07E-03	3,03E-03
V23	8,03E-03	7,49E-03	8,86E-03	8,17E-03

Tabela 12: Caso 1: Base Busca de Fatores-Intervalo de Confiança - Fator 1

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	-0,28	0,10	-0,28	0,10	-0,33	0,10	-0,31	0,10
V2	0,11	0,49	0,11	0,46	0,08	0,49	0,10	0,47
V3	0,26	0,62	0,32	0,62	0,29	0,62	0,33	0,62
V4	0,60	0,80	0,60	0,77	0,61	0,78	0,63	0,78
V5	0,58	0,78	0,59	0,76	0,58	0,76	0,59	0,76
V6	0,25	0,65	0,33	0,61	0,31	0,61	0,33	0,61
V7	0,58	0,72	0,55	0,72	0,54	0,72	0,55	0,73
V8	0,46	0,68	0,39	0,68	0,39	0,68	0,41	0,69
V9	0,29	0,57	0,29	0,57	0,25	0,57	0,27	0,58
V10	0,38	0,73	0,46	0,75	0,45	0,76	0,45	0,75
V11	0,47	0,72	0,47	0,75	0,48	0,75	0,46	0,75
V12	-0,07	0,34	-0,16	0,33	-0,18	0,32	-0,17	0,30
V13	0,29	0,59	0,23	0,57	0,21	0,57	0,25	0,55
V14	0,34	0,68	0,31	0,65	0,26	0,64	0,30	0,64
V15	0,70	0,84	0,69	0,84	0,69	0,84	0,70	0,84
V16	0,51	0,83	0,53	0,80	0,56	0,80	0,56	0,79
V17	0,48	0,76	0,52	0,76	0,53	0,77	0,53	0,77
V18	0,51	0,72	0,49	0,72	0,46	0,72	0,45	0,72
V19	0,58	0,73	0,55	0,72	0,55	0,73	0,56	0,73
V20	0,25	0,58	0,24	0,51	0,24	0,52	0,23	0,52
V21	-0,09	0,41	-0,11	0,36	-0,14	0,36	-0,13	0,36
V22	0,61	0,77	0,60	0,77	0,60	0,77	0,60	0,77
V23	0,26	0,55	0,25	0,55	0,22	0,56	0,23	0,56

Tabela 13: Caso 1: Base Busca de Fatores-Intervalo de Confiança - Fator 2

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	0,29	0,55	0,24	0,59	0,27	0,60	0,26	0,60
V2	0,13	0,48	0,11	0,50	0,08	0,52	0,08	0,53
V3	-0,04	0,35	-0,01	0,35	-0,02	0,37	-0,02	0,35
V4	-0,13	0,15	-0,13	0,17	-0,11	0,17	-0,12	0,19
V5	-0,27	0,12	-0,22	0,12	-0,22	0,16	-0,18	0,15
V6	0,05	0,49	0,06	0,48	0,06	0,48	0,06	0,43
V7	-0,15	0,17	-0,18	0,23	-0,18	0,21	-0,16	0,22
V8	-0,08	0,40	-0,07	0,40	-0,06	0,40	-0,04	0,38
V9	-0,04	0,40	0,00	0,45	0,00	0,47	-0,01	0,43
V10	-0,74	-0,30	-0,65	-0,31	-0,65	-0,31	-0,67	-0,36
V11	-0,71	-0,29	-0,64	-0,29	-0,63	-0,29	-0,65	-0,32
V12	0,50	0,79	0,58	0,81	0,60	0,80	0,59	0,80
V13	0,08	0,57	0,12	0,50	0,14	0,51	0,12	0,49
V14	0,28	0,63	0,35	0,68	0,35	0,68	0,32	0,67
V15	-0,41	-0,17	-0,41	-0,15	-0,41	-0,12	-0,40	-0,13
V16	-0,67	-0,25	-0,62	-0,25	-0,57	-0,27	-0,59	-0,27
V17	-0,65	-0,20	-0,61	-0,18	-0,58	-0,18	-0,57	-0,20
V18	0,05	0,31	-0,03	0,31	-0,03	0,33	-0,04	0,33
V19	-0,11	0,31	0,00	0,34	0,00	0,34	0,01	0,32
V20	0,01	0,40	0,02	0,40	0,02	0,41	0,00	0,38
V21	0,52	0,81	0,61	0,84	0,62	0,82	0,60	0,82
V22	-0,23	0,23	-0,23	0,12	-0,20	0,14	-0,20	0,17
V23	0,24	0,54	0,23	0,53	0,21	0,59	0,21	0,57

Tabela 14: Caso 2: Base Creme Dental-EMQ - Fator 1

Variáveis	B=25	B=100	B=200	B=1000
V1	9,14E-05	3,65E-05	1,82E-05	3,62E-06
V2	1,23E-02	3,98E-03	1,98E-03	3,94E-04
V3	1,76E-04	4,59E-05	2,28E-05	4,55E-06
V4	1,01E-02	3,22E-03	1,60E-03	3,20E-04
V5	3,46E-04	1,20E-04	5,96E-05	1,19E-05
V6	1,16E-02	3,80E-03	1,89E-03	3,76E-04

Tabela 15: Caso 2: Base Creme Dental-EMQ - Fator 2

Variáveis	B=25	B=100	B=200	B=1000
V1	9,04E-03	2,96E-03	1,47E-03	2,93E-04
V2	2,24E-04	7,85E-05	3,91E-05	7,79E-06
V3	9,42E-03	3,05E-03	1,52E-03	3,02E-04
V4	1,05E-03	3,57E-04	1,78E-04	3,54E-05
V5	7,98E-03	2,57E-03	1,28E-03	2,55E-04
V6	4,06E-04	1,27E-04	6,30E-05	1,26E-05

Tabela 16: Caso 2: Base Creme Dental-Intervalo de Confiança - Fator 1

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	0,92	0,95	0,93	0,95	0,94	0,95	0,95	0,95
V2	-0,27	0,20	-0,19	0,15	-0,17	0,06	-0,05	-0,05
V3	0,88	0,94	0,90	0,93	0,92	0,93	0,93	0,93
V4	-0,43	0,00	-0,37	-0,10	-0,36	-0,16	-0,24	-0,24
V5	-0,93	-0,85	-0,92	-0,88	-0,92	-0,88	-0,91	-0,91
V6	-0,14	0,33	-0,09	0,26	-0,05	0,16	0,08	0,08

Tabela 17: Caso 2: Base Creme Dental-Intervalo de Confiança - Fator 2

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	-0,16	0,25	-0,08	0,22	-0,01	0,18	0,07	0,07
V2	0,79	0,85	0,81	0,85	0,81	0,84	0,84	0,84
V3	-0,30	0,13	-0,22	0,07	-0,14	0,05	-0,06	-0,06
V4	0,75	0,88	0,78	0,86	0,79	0,86	0,83	0,83
V5	-0,36	0,00	-0,30	-0,03	-0,29	-0,10	-0,18	-0,18
V6	0,80	0,88	0,83	0,88	0,85	0,87	0,87	0,87

Tabela 18: Caso 3: Base Modo de Vida-EMQ - Fator 1

Variáveis	B=25	B=100	B=200	B=1000
V1	1,18E-03	2,86E-04	1,42E-04	2,83E-05
V2	1,03E-02	2,50E-03	1,24E-03	2,48E-04
V3	1,03E-04	2,50E-05	1,24E-05	2,48E-06
V4	7,44E-03	1,80E-03	8,97E-04	1,79E-04
V5	2,36E-03	5,73E-04	2,85E-04	5,68E-05
V6	1,15E-02	2,79E-03	1,39E-03	2,76E-04
V7	1,96E-03	4,75E-04	2,36E-04	4,71E-05

Tabela 19: Caso 3 : Base Modo de Vida-EMQ - Fator 2

Variáveis	B=25	B=100	B=200	B=1000
V1	2,15E-02	5,22E-03	2,60E-03	5,18E-04
V2	9,88E-02	2,40E-02	1,19E-02	2,38E-03
V3	4,25E-03	1,03E-03	5,12E-04	1,02E-04
V4	6,05E-02	1,47E-02	7,31E-03	1,46E-03
V5	2,93E-02	7,11E-03	3,54E-03	7,05E-04
V6	7,14E-02	1,73E-02	8,63E-03	1,72E-03
V7	2,59E-02	6,28E-03	3,12E-03	6,22E-04

Tabela 20: Caso 3 : Base Modo de Vida-Intervalo de Confiança - Fator 1

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	0,75	0,89	0,76	0,87	0,78	0,85	0,82	0,82
V2	0,12	0,50	0,14	0,43	0,16	0,38	0,28	0,28
V3	0,87	0,91	0,87	0,90	0,88	0,89	0,89	0,89
V4	-0,35	0,02	-0,32	-0,03	-0,27	-0,15	-0,20	-0,20
V5	0,60	0,79	0,61	0,78	0,62	0,70	0,66	0,66
V6	-0,18	0,27	-0,09	0,22	-0,04	0,15	0,05	0,05
V7	-0,77	-0,59	-0,74	-0,59	-0,72	-0,64	-0,68	-0,68

Tabela 21: Caso 3: Base Modo de Vida-Intervalo de Confiança - Fator 2

Variáveis	B=25		B=100		B=200		B=1000	
	LI	LS	LI	LS	LI	LS	LI	LS
V1	-0,26	0,50	0,23	0,47	0,31	0,46	0,38	0,38
V2	-0,80	0,80	-0,78	-0,60	-0,74	-0,64	-0,71	-0,71
V3	-0,18	0,09	-0,13	0,08	-0,10	0,02	-0,03	-0,03
V4	-0,54	0,71	0,51	0,69	0,55	0,67	0,63	0,63
V5	-0,24	0,63	0,26	0,62	0,44	0,58	0,51	0,51
V6	-0,72	0,67	-0,69	-0,49	-0,65	-0,53	-0,60	-0,60
V7	-0,34	0,52	0,23	0,49	0,32	0,44	0,38	0,38