Componentes Principales 2 Carlos David Contreras Chacon 2022-10-19 Parte A X = read.csv("paises_mundo.csv") cat("Matrices de varianza-covarianza S\n") ## Matrices de varianza-covarianza S S = cov(X)S CrecPobl MortInf PorcMujeres ## CrecPobl 1.538298e+00 2.195026e+01 -6.078026e+00 -8.933379e+04 ## MortInf 2.195026e+01 1.032859e+03 -9.249342e+00 -2.269332e+06 ## PorcMujeres -6.078026e+00 -9.249342e+00 7.698322e+01 2.813114e+05 ## PNB95 -8.933379e+04 -2.269332e+06 2.813114e+05 4.999786e+10 ## ProdElec -4.973964e+04 -1.043435e+06 2.260248e+05 2.247791e+10 ## LinTelf -1.369079e+02 -4.381366e+03 4.499750e+02 2.039550e+07 ## ConsAgua -4.827092e+01 -1.288211e+03 -1.568313e+03 1.097481e+07 ## PropBosq -3.887018e+00 -1.466316e+01 6.517895e+01 2.474311e+05 ## PropDefor 3.361974e-01 1.276296e+01 2.680592e-01 -5.806203e+04 ## ConsEner -8.384169e+02 -4.442568e+04 2.855207e+02 1.415628e+08 -1.137877e+00 -9.485500e+01 -2.150132e+00 2.501673e+05 ## EmisCO2 ## ProdElec LinTelf ConsAgua PropBosq ## CrecPobl -4.973964e+04 -1.369079e+02 -4.827092e+01 -3.887018 -1.043435e+06 -4.381366e+03 -1.288211e+03 -14.663158 ## MortInf ## PorcMujeres 2.260248e+05 4.499750e+02 -1.568313e+03 65.178947 2.247791e+10 2.039550e+07 1.097481e+07 247431.122807 ## PNB95 ## ProdElec 1.821909e+10 7.583050e+06 1.399817e+07 70359.785965 ## LinTelf 7.583050e+06 3.841247e+04 1.193110e+04 248.715789 1.399817e+07 1.193110e+04 3.301981e+05 -2220.757895 ## ConsAgua ## PropBosq 7.035979e+04 2.487158e+02 -2.220758e+03 401.003509 ## PropDefor -3.180340e+04 -9.940461e+01 -6.743793e+01 2.625263 ## ConsEner 6.801296e+07 3.426262e+05 2.092242e+05 -5153.438596 ## EmisCO2 1.392779e+05 6.385700e+02 4.869328e+02 -12.897193 ## PropDefor ConsEner EmisCO2 ## CrecPobl 3.361974e-01 -8.384169e+02 -1.137877 ## MortInf 1.276296e+01 -4.442568e+04 -94.855000 ## PorcMujeres 2.680592e-01 2.855207e+02 -2.150132 ## PNB95 -5.806203e+04 1.415628e+08 250167.323509 ## ProdElec -3.180340e+04 6.801296e+07 139277.888640 ## LinTelf -9.940461e+01 3.426262e+05 638.570000 ## ConsAgua -6.743793e+01 2.092242e+05 486.932763 ## PropBosq 2.625263e+00 -5.153439e+03 -12.897193 ## PropDefor 1.817253e+00 -1.051522e+03 -2.632487 ## ConsEner -1.051522e+03 5.014395e+06 10286.159781 -2.632487e+00 1.028616e+04 ## EmisCO2 27.268614 cat("Matriz de correlaciones\n") ## Matriz de correlaciones R = cor(X)CrecPobl MortInf PorcMujeres PNB95 ProdElec 1.00000000 0.55067948 -0.55852711 -0.32212154 -0.29711119 ## CrecPobl ## MortInf ## PorcMujeres -0.55852711 -0.03280139 1.00000000 0.14338826 0.19085114 ## PNB95 -0.32212154 -0.31579250 0.14338826 1.00000000 0.74476081 ## ProdElec -0.29711119 -0.24053689 0.19085114 0.74476081 1.00000000 ## LinTelf -0.56321228 -0.69558922 0.26167018 0.46539599 0.28664508 ## ConsAgua -0.06772953 -0.06975563 -0.31106243 0.08541500 0.18047653 ## PropBosq -0.15650281 -0.02278415 0.37096694 0.05525919 0.02603078 ## PropDefor 0.20107881 0.29459348 0.02266339 -0.19262327 -0.17478434 ## ConsEner -0.30187731 -0.61731132 0.01453216 0.28272492 0.22501894 ## EmisCO2 -0.17568860 -0.56520778 -0.04692837 0.21425123 0.19760017 LinTelf ConsAgua PropBosq PropDefor ConsEner ## CrecPobl -0.56321228 -0.06772953 -0.15650281 0.20107881 -0.30187731 ## MortInf -0.69558922 -0.06975563 -0.02278415 0.29459348 -0.61731132 ## PorcMujeres 0.26167018 -0.31106243 0.37096694 0.02266339 0.01453216 ## PNB95 0.46539599 0.08541500 0.05525919 -0.19262327 0.28272492 ## ProdElec 0.28664508 0.18047653 0.02603078 -0.17478434 0.22501894 ## LinTelf 1.00000000 0.10593934 0.06337138 -0.37623801 0.78068385 ## ConsAgua 0.10593934 1.000000000 -0.19299225 -0.08705811 0.16259804 ## PropBosq 0.06337138 -0.19299225 1.00000000 0.09725032 -0.11492480 ## PropDefor -0.37623801 -0.08705811 0.09725032 1.000000000 -0.34833836 ## ConsEner 0.78068385 0.16259804 -0.11492480 -0.34833836 1.00000000 ## EmisCO2 0.62393719 0.16227447 -0.12333592 -0.37396154 0.87965517 ## CrecPobl -0.17568860 ## MortInf -0.56520778 ## PorcMujeres -0.04692837 ## PNB95 0.21425123 ## ProdElec 0.19760017 ## LinTelf 0.62393719 ## ConsAgua 0.16227447 ## PropBosq -0.12333592 ## PropDefor -0.37396154 0.87965517 ## ConsEner ## EmisCO2 1.00000000 eigenS = eigen(S)eigenR = eigen(R)lambdas = eigenS\$values totvar = sum(diag(S))cat("Proporción de varianza explicada por cada componente\n") ## Proporción de varianza explicada por cada componente varXcomp = lambdas / totvar varXcomp ## [1] 9.034543e-01 9.647298e-02 6.795804e-05 4.554567e-06 1.782429e-07 ## [6] 7.530917e-09 5.317738e-09 6.657763e-10 8.502887e-11 2.107843e-11 ## [11] 6.989035e-12 cumsum(varXcomp) ## [1] 0.9034543 0.9999273 0.99999953 0.9999998 1.00000000 1.00000000 1.00000000 ## [8] 1.0000000 1.0000000 1.0000000 1.0000000 La primera componente corresponde en gran medida a la varianza total explicada, la segunda tambien aporta una parte pero a partir de ahi es minima la varianza que van aportando cada componente. lambdas = eigenR\$values totvar = sum(diag(R))cat("Proporción de varianza explicada por cada componente\n") ## Proporción de varianza explicada por cada componente varXcomp = lambdas / totvar varXcomp [1] 0.366352638 0.175453813 0.124582832 0.078592361 0.072194597 0.066290906 ## [7] 0.051936828 0.029709178 0.015278951 0.013302563 0.006305332 cumsum(varXcomp) ## [1] 0.3663526 0.5418065 0.6663893 0.7449816 0.8171762 0.8834671 0.9354040 ## [8] 0.9651132 0.9803921 0.9936947 1.00000000 En este caso el aumento va siendo mucho mas gradual, con mas componentes que aportan partes significantes para la varianza total de R. Parte B library(stats) library(factoextra) ## Loading required package: ggplot2 ## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa library(ggplot2) datos=S cpS=princomp(datos,cor=FALSE) cpaS=as.matrix(datos)%*%cpS\$loadings plot(cpaS[,1:2], type="p", main = "Título") text(cpaS[,1],cpaS[,2],1:nrow(cpaS)) **Título** 4 0e+00 -6e+09 -4e+09 5 -3e+10 -5e+10 -4e+10 -2e+10 -1e+10 0e+00 Comp.1 biplot(cpS) ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length = ## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length = ## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length = ## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =## arrow.len): zero-length arrow is of indeterminate angle and so skipped ## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =## arrow.len): zero-length arrow is of indeterminate angle and so skipped -4e+10 -2e+10 0e+00 2e+10 NB95 2e+10 0.4 0.2 00+00 NB95 < 0.0 ProdElec -0.2 -2e+10 -0.4 -0.6 -4e+10 -0.8 **ProdElec** -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 Comp.1 datos=R cpS=princomp(datos,cor=FALSE) cpaS=as.matrix(datos)%*%cpS\$loadings plot(cpaS[,1:2], type="p", main = "Título") text(cpaS[,1],cpaS[,2],1:nrow(cpaS)) **Título** Ø 0.5 171 100 0.0 2 **4** 5 5 Ò 8 -1.0 3 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 Comp.1 biplot(cpS) -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 0.4 CrecPobl ConsAgua EmisCO2 ConsEner CrecPobl 0.2 0.5 0.0 0.0 >M8rtl inTel**t∜nTelf** Proposer r PNBUSec -0.5 -0.2 **P**ropBosq -1.0 PorcMujeres PropBosq -0.4 -1.5 9.0-PorcMujeres -0.6 -0.4 -0.2 0.0 0.2 0.4 Comp.1 Parte C library(FactoMineR) datos=S cp3 = PCA(datos)## Warning: ggrepel: 9 unlabeled data points (too many overlaps). Consider ## increasing max.overlaps PCA graph of individuals Dim 2 (3.23%) ProdElec PNB95 Dim 1 (96.75%) PCA graph of variables Dim 2 (3.23%) 0.5 ConsAgua> Prodlec PorcMujeres MørtInf <u>PropDefor</u> ConsEner EmisC 02 CrecPobl PropB/osq -0.5 -1.0 -1.0 -0.5 0.0 Dim 1 (96.75%) fviz_pca_ind(cp3, col.ind = "blue", addEllipses = TRUE, repel = TRUE) ## Warning: ggrepel: 9 unlabeled data points (too many overlaps). Consider ## increasing max.overlaps Individuals - PCA ProdElec Dim2 (3.2%) -1 **-**-5 -10 10 Dim1 (96.7%) fviz_screeplot(cp3) Scree plot 100 -Percentage of explained variances Dimensions fviz_contrib(cp3, choice = c("var")) Contribution of variables to Dim-1 7.5 **-**Contributions (%) 2.5 -0.0 -Prophetor Finiscoly Coustles Worthy Swappe Tilley Sold Mister Stockhold Coustains datos=R cp3 = PCA(datos)PCA graph of individuals Dim 2 (21.97%) PorcMujeres PropBosq PNB95 • PropDefor Lin**T**elf MortInf ConsEner EmisCO2 • CrecPobl ConsAgua 2.5 **Dim 1 (57.21%)** -5.0 -2.5 0.0 PCA graph of variables Dim 2 (21.97%) PorcMujeres PropB**o**sq 0.5 PNB9 PropDefor <MortInf 0.0 ConsEner EmisCQ2 CrecPobl -0.5 ConsAgua -1.0 -0.5 0.0 -1.0 1.0 Dim 1 (57.21%) fviz_pca_ind(cp3, col.ind = "blue", addEllipses = TRUE, repel = TRUE) Individuals - PCA 5.0 -PorcMujeres 2.5 -PropBosq PropDefor Dim2 (22%) PNB95 LinTelf MortInf ConsEner EmisCO2 CrecPobl -2.5 **-**ConsAgua -5.0 **-**Dim1 (57.2%) fviz_screeplot(cp3) Scree plot 60 **-**3 Dimensions fviz_contrib(cp3, choice = c("var")) Contribution of variables to Dim-1 15 - Contributions (%)

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