

s je sumacioni vektor, ta da je $s^t s = ?$

$$s^t s = [1 \ 1 \ 1 \ 1] \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = 1 * 1 + 1 * 1 + 1 * 1 + 1 * 1 = 4$$

$s^t R s =$

$$s^t R s = [1 \ 1 \ 1 \ 1] \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 3 \\ 0 & 2 & 0 & 3 \\ 1 & 2 & 4 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$s^t R s = [2 \ 7 \ 9 \ 15] \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = 33$$

Projektor

A razapinje nejeki prostor

$$P = PP$$

$$\begin{aligned} P_A &= A(A^t A)^{-1} A^t \\ P_A P_A &= A(A^t A)^{-1} (A^t A) (A^t A)^{-1} A^t \\ &= A(A^t A)^{-1} I A^t \\ &= A(A^t A)^{-1} A^t \\ &= P_A \end{aligned}$$

$P_A x = x'$ slika vektora, koordinate vektora u prostoru kolona matrice A
 $x - P_A x = (I - P_A) x = x''$ deo vektora x koji štrči iz prostora

$$\begin{aligned} P_{A'} &= (I - P_A) \\ P_{A'} P_{A'} &= (I - P_A)(I - P_A) \\ &= I I - I P_A - P_A I + P_A P_A \\ &= I - P_A - P_A + P_A \\ &= I - P_A \end{aligned}$$

Izvodi

Projektor od centroidnog (sumacionog) vektora

$$\begin{aligned} P_s &= s(s^t s)^{-1} s^t \\ &= s n^{-1} s^t \\ &= \end{aligned}$$

$$ss^t n^{-1} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} [1 \ 1 \ 1 \ 1] 4^{-1} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} 4^{-1}$$

$$ss^t n^{-1} = \begin{bmatrix} 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \end{bmatrix}$$

$$P_s x =$$

$$ss^t n^{-1} x = \begin{bmatrix} 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 4 \\ 4 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \\ 3 \end{bmatrix}$$

$$(I - P_s) = I - s(s^t s)^{-1} s^t$$

$$x' = (I - P_s)x = x - P_s x$$

$$\begin{bmatrix} 1 \\ 3 \\ 4 \\ 4 \end{bmatrix} - \begin{bmatrix} 3 \\ 3 \\ 3 \\ 3 \end{bmatrix} = \begin{bmatrix} -2 \\ 0 \\ 1 \\ 1 \end{bmatrix}$$

$$x^t x, n^{-1} = \begin{bmatrix} -2 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} -2 \\ 0 \\ 1 \\ 1 \end{bmatrix} 4^{-1} = 6/4 = 1.5 = \text{Varijansa}$$

Regresija

$$Zx = k' = k - e$$

$$f(e) = \text{najmanji kvadrat} = e^t e$$

$$e = k - Zx$$

$$\begin{aligned} f(x) &= e^t e = (k - Zx)^t (k - Zx) \\ &= (k^t - x^t Z^t) (k - Zx) \\ &= k^t k - k^t Zx - x^t Z^t k + x^t Z^t Zx \\ &= k^t k - 2 x^t Z^t k + x^t Z^t Zx = \text{minimum (najmanji kvadrat)} \end{aligned}$$

$$f(x)' = \frac{\partial f(x)}{\partial x} = 0 - 2 Z^t k + 2 Z^t Zx$$

$$-2 Z^t k + 2 Z^t Zx = 0$$

$$-Z^t k + Z^t Zx = 0$$

$$Z^t Zx = Z^t k$$

$$(Z^t Z)^{-1} Z^t Zx = (Z^t Z)^{-1} Z^t k$$

$$x = (Z^t Z)^{-1} Z^t k$$

predvidjeni skorovi

$$k' = Zx = Z(Z^t Z)^{-1} Z^t k = P_z k$$

$$e = k - k' = k - Zx = k - Z(Z^t Z)^{-1} Z^t k = k - P_z k = (I - P_z) k$$

Domaci

Treba napraviti program za racunanje multiple regresije, u matricnom obliku. U SPSS-u napravite standradizovane 3 koolone sa 5 isptiaanika i onda te vrednosti prenesetu u R i proverite vas racun i SPSS.

Svojstvene vrednosti i svojstveni vektori – eigenvalues & eigenvectors

A – kvadratna matrica

$$A x = \lambda x$$

λ skalara, svojstvena vrednost

x je vektor svojstveni vektor

$$A x - \lambda x = 0$$

$$(A - I\lambda) x = 0$$

$$|A - I\lambda| = 0$$

dimenzije matrice A je m

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

$$X = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix}$$

$$AX = X\Lambda$$

$$AXX^{-1} = A = X\Lambda X^{-1}$$

$$X^{-1}AX = X^{-1}X\Lambda = \Lambda$$

ako je matrica A simetricna, onda je X ortogonalno $X^tX = I$

$$AX = X \Lambda$$

$$AXX^t = A = X \Lambda X^t$$

$$X^t AX = X^t X \Lambda = \Lambda$$

Lagrange multipliers

Principal components

Z – standardizovane varijable, u standardnoj normalnoj formi

$Z^t Z = R$ matrica inerkorelacija

$Zx = k$, $k^t k = \text{varijansa} = \text{zelimo maximum, pod uslovom da je } x^t x = 1$

$$f(x) = k^t k = (Zx)^t (Zx) = x^t Z^t Z x = x^t Rx = \text{maximum}$$

$$g(x) = x^t x - 1$$

$$L(x, \lambda) = f(x) - \lambda g(x)$$

$$= x^t Rx - \lambda(x^t I x - 1)$$

$$\frac{\partial L}{\partial x} = 2Rx - \lambda 2Ix = 0$$

$$\frac{\partial L}{\partial \lambda} = x^t x - 1 = 0$$

$$(R - \lambda I)x = 0$$

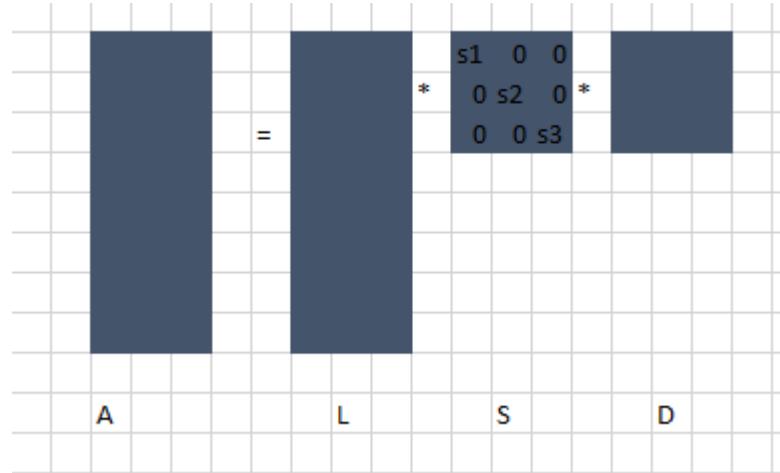
$$|R - \lambda I| = 0$$

λ – varijanse glavnih komponenti

x – koeficijenti za dovijanje glavnih komponenti

$$\mathbf{k} = \mathbf{Z}\mathbf{x}$$

Singular Value Decomposition



$$\mathbf{A} = \mathbf{L}\mathbf{S}\mathbf{D}^t$$

$\mathbf{S} = \text{diag}(\mathbf{S})$ – matrica sa singularnim vrednostima

$$\mathbf{L}^t\mathbf{L} = \mathbf{I}$$

$$\mathbf{D}^t\mathbf{D} = \mathbf{D}\mathbf{D}^t = \mathbf{I}$$

\mathbf{D} je matrica svojstvenih vektora

$$\mathbf{A}^t\mathbf{A} = (\mathbf{L}\mathbf{S}\mathbf{D}^t)^t(\mathbf{L}\mathbf{S}\mathbf{D}^t)$$

$$= (\mathbf{D}^t)^t\mathbf{S}^t\mathbf{L}^t\mathbf{L}\mathbf{S}\mathbf{D}^t$$

$$= \mathbf{D}\mathbf{S}\mathbf{L}^t\mathbf{L}\mathbf{S}\mathbf{D}^t$$

$$= \mathbf{D} \mathbf{S}^2 \mathbf{D}^t$$

$$\text{iz svojstvenih vrednosti } \mathbf{A}\mathbf{X}\mathbf{X}^t = \mathbf{A} = \mathbf{X}\Lambda\mathbf{X}^t$$

sto znači da je $\mathbf{D} = \mathbf{X}$, a $\Lambda = \mathbf{S}^2$

Generalized inverse

Je ustvari SVD

$$\mathbf{A} = \mathbf{L}\mathbf{S}\mathbf{D}^t$$

tada je generalizovani invez matrice $\mathbf{A}^* = \mathbf{L}\mathbf{S}^{-1}\mathbf{D}^t$

Optimization algorithms

Newton's method

Quasi-Newton method

Gradient descent

R

Domaci

<https://machinelearningmastery.com/logistic-regression-tutorial-for-machine-learning/>