

12.

Matrix factorization theorem (spectral decomposition)

Twierdzenie 1 Each positive definite and symmetrical matrix \mathbf{M} can be expressed in the form

$$\mathbf{M} = \mathbf{P}\mathbf{P}^T$$

where \mathbf{P} is nonsingular matrix (called "root of \mathbf{M} ").

Proof

let

$$\begin{aligned} w_i &- \text{eigenvalues of } M, & i &= 1, 2, \dots, s \\ \lambda_i &- \text{eigenvectors of } M, & i &= 1, 2, \dots, s \end{aligned}$$

obviously

$$Mw_i = \lambda_i w_i$$

introducing

$$W = [w_1, w_2, \dots, w_s] \text{ and } \Lambda = \text{diag}(\lambda_i) = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots \\ \dots & 0 & \dots & 0 \\ 0 & \dots & 0 & \lambda_s \end{bmatrix}$$

since M is symmetrical

$$\lambda_i - \text{real}$$

$$w_i - \text{mutually orthogonal}$$

and since $M > 0$

$$\lambda_i > 0 \text{ for each } i = 1, 2, \dots, s$$

conclusion

$$\begin{aligned} W &- \text{orthogonal} \\ W^T W &= I \implies W^{-1} = W^T \end{aligned}$$

hence

$$\begin{aligned} MW &= W\Lambda \\ M &= W\Lambda W^{-1} = W\Lambda W^T \text{ (spectral decomposition)} \end{aligned}$$

introducing

$$\Lambda = \Lambda^{1/2}\Lambda^{1/2}, \text{ where } \Lambda^{1/2} = \text{diag}(\sqrt{\lambda_i}) = \begin{bmatrix} \sqrt{\lambda_1} & 0 & \dots & 0 \\ 0 & \sqrt{\lambda_2} & 0 & \dots \\ \dots & 0 & \dots & 0 \\ 0 & \dots & 0 & \sqrt{\lambda_s} \end{bmatrix}$$

we get

$$W\Lambda W^T = W\Lambda^{1/2} \left(W\Lambda^{1/2}\right)^T$$

stad

$$M = PP^T, \text{ where } P = W\Lambda^{1/2} \blacksquare$$

1. Cholesky LU decomposition

$$Ma = b$$

M – positive ($M > 0$), symmetrical ($M^T = M$)

Cholesky

$$M = LDL^T$$

where

L – lower triangular matrix, i.e. for $i < j$ it hold that $L[i, j] = 0$

D – diagonal, $D = \text{diag}(d_i), d_i > 0$

hence

$$M = LD^{1/2}D^{1/2}L^T$$

where $D^{1/2} = \text{diag}(\sqrt{d_i})$

for

$$\bar{L} = LD^{1/2}$$

we get

$$M = \bar{L}\bar{L}^T$$

\bar{L} – lower triangular

\bar{L}^T – upper triangular

$$\det M = \det \bar{L}\bar{L}^T = \det \bar{L} \det \bar{L}^T > 0$$

Procedure

$$\bar{L}\bar{L}^T a = b$$

Stage 1.

for $\alpha = \bar{L}^T a$ solve "external" problem

$$\bar{L}\alpha = b$$

since $\det \bar{L} > 0$ diagonal elements of \bar{L} are nonzero, the solution exists and is unique

Stage 2.

α_N – given (computed in Stage 1)

solve "internal" problem

$$\bar{L}^T a = \alpha_N$$

since $\det \bar{L}^T > 0$

Advantages

1) reduced complexity

2) better accuracy

$$\begin{aligned}\det M &= (\det \bar{L})^2 \\ \det \bar{L} &= \sqrt{\det M}\end{aligned}$$

hence for $0 < \det M < 1$

$$\det \bar{L} = \det \bar{L}^T > \det M$$

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2. Householder reflections

Definicja 1 *Macierzą odbicia Householdera nazywamy macierz postaci*

$$P = I - 2ww^T$$

gdzie $\|w\| = 1$, tj. $w^T w = 1$.

Interpretacja

$$Pw = (I - 2ww^T)w = w - 2ww^T w = w - 2w = -w$$

Własności:

(i) $P = P^T$ (symetria)

(ii) $P^T P = P P^T = I$ (ortogonalność)

(iii) dla każdego wektora x istnieje macierz P_j , taka że

$$P_j x = \pm \|x\| e_j, \text{ gdzie } e_j \text{ jest } j\text{-tym wersorem } e_j = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

(iiii) dla ciągu macierzy Householdera P_1, P_2, \dots, P_s przekształcenie złożenia

$$\Psi = P_s \cdot P_{s-1} \cdot \dots \cdot P_2 \cdot P_1$$

jest macierzą ortogonalną, tj.

$$\Psi^T \Psi = I$$

Dowody

(i) oczywiste – różnica macierzy symetrycznych jest symetryczna

(ii)

$$PP^T = (I - 2ww^T)(I - 2ww^T) = I - 2ww^T - 2ww^T + 4ww^Tww^T = I - 4ww^T + 4ww^T = I$$

(iii)

x – dowolny wektor

$$u_j = x \pm \|x\| e_j$$

$$P_j = I - \frac{u_j u_j^T}{H_j}, \text{ gdzie } H_j = \frac{1}{2} \|u_j\|^2 = \frac{1}{2} u_j^T u_j$$

macierz P_j jest macierzą Householdera, ponieważ

$$P_j = I - 2 \frac{u_j}{\|u_j\|} \frac{u_j^T}{\|u_j\|} \text{ (wektory o długości 1)}$$

$$\begin{aligned} P_j x &= \left(I - \frac{u_j u_j^T}{H_j} \right) x = x - u_j \frac{2u_j^T x}{u_j^T u_j} = x - u_j \frac{2(x \pm \|x\| e_j)^T x}{(x \pm \|x\| e_j)^T (x \pm \|x\| e_j)} = \\ &= \dots = x - u_j = x - x \pm \|x\| e_j = \pm \|x\| e_j \end{aligned}$$

(iii)

$$\Psi^T \Psi = P_1^T P_2^T \dots P_{s-1}^T \underbrace{P_s^T P_s}_{=I} P_{s-1} \dots P_2 P_1 = I$$

Lemat 1 Dla każdej macierzy $X_{N \times s}$ istnieje taki ciąg macierzy Householdera $\{\tilde{P}_1, \tilde{P}_2, \dots, \tilde{P}_s\}$, że przekształcenie

$$\Psi = \tilde{P}_s \tilde{P}_{s-1}, \dots, \tilde{P}_2 \tilde{P}_1$$

ma własność

$$\Psi X_{N \times s} = \begin{bmatrix} R \\ 0 \end{bmatrix}$$

gdzie R jest macierzą trójkątną górną o rozmiarach $s \times s$.

Twierdzenie 2 Oszacowanie NK jest równoważne rozwiązaniu układu równań

$$Ra = \eta_R$$

gdzie η_R jest wektorem zawierającym s pierwszych elementów wektora ΨY_N .

Rozwiązanie tego układu równań istnieje i jest jednoznaczne. Pytanie dotyczy samego oszacowania NK.

Dowód

$$\begin{aligned} Q(a) &= \|X_N a - Y_N\|_e^2 \rightarrow \min_a \\ Q(a) &= (X_N a - Y_N)^T (X_N a - Y_N) = \\ &= (X_N a - Y_N)^T \Psi^T \Psi (X_N a - Y_N) = \\ &= [\Psi (X_N a - Y_N)]^T [\Psi (X_N a - Y_N)] = \\ &= [\Psi X_N a - \Psi Y_N]^T [\Psi X_N a - \Psi Y_N] = \\ &= \left[\begin{bmatrix} R \\ 0 \end{bmatrix} a - \begin{bmatrix} \eta_R \\ \eta_z \end{bmatrix} \right]^T \left[\begin{bmatrix} R \\ 0 \end{bmatrix} a - \begin{bmatrix} \eta_R \\ \eta_z \end{bmatrix} \right] = \\ &= \begin{bmatrix} Ra - \eta_R \\ -\eta_z \end{bmatrix}^T \begin{bmatrix} Ra - \eta_R \\ -\eta_z \end{bmatrix} = (Ra - \eta_R)^T (Ra - \eta_R) + \eta_z^T \eta_z = \\ &= \|Ra - \eta_R\|_e^2 + \|\eta_z\|_e^2 \rightarrow \min_a \end{aligned}$$

zatem

$$Q(a) \rightarrow \min_a \Rightarrow Ra = \eta_R$$