

Rectangular matrices with O.N. cols

want to solve $Qx=b$

$${}^n \begin{bmatrix} - & q_1^T & - \\ & \vdots & \\ - & q_n^T & - \end{bmatrix} \begin{bmatrix} | & & | \\ q_1 & \dots & q_n \\ | & & | \end{bmatrix} = {}^n \begin{bmatrix} - & & - \\ & I & \\ - & & - \end{bmatrix}$$

Here Q^T is only a left inv of Q .

O.N. matrices are crucial for numerical LA as they don't change lengths and thus keep round off error under control.

To solve $Qx=b$ we use least squares (it is simpler than usual)

$$Q^T Q \bar{x} = Q^T b$$

$$\bar{x} = Q^T b$$

$$p = Q \bar{x} = Q Q^T b$$

proj of b onto cols: $p = P b = q_1^T b \hat{q}_1 + \dots + q_n^T b \hat{q}_n$

$$= Q Q^T b$$

$$= [\hat{q}_1 \hat{q}_1^T + \hat{q}_2 \hat{q}_2^T + \dots + \hat{q}_n \hat{q}_n^T] b$$

Proj Matrix

$$P = A (A^T A)^{-1} A^T$$

$$= Q (Q^T Q)^{-1} Q^T = Q Q^T \text{ simple form for } P$$

Gram-Schmidt (orthonormalization process)

Take an arb set of vectors $\{a_1, \dots, a_n\}$ and produce an O.N. set $\{\hat{q}_1, \dots, \hat{q}_n\}$ with same span.

Let $\hat{q}_1 := \frac{1}{\|a_1\|} a_1$

$$a_2' := \bar{a}_2 - P_{\hat{q}_1}(a_2) = \bar{a}_2 - (q_1^T a_2) \hat{q}_1$$

$$\hat{q}_2 := \frac{1}{\|a_2'\|} a_2'$$

$$a_3' := \bar{a}_3 - (q_1^T a_3) \hat{q}_1 - (q_2^T a_3) \hat{q}_2$$

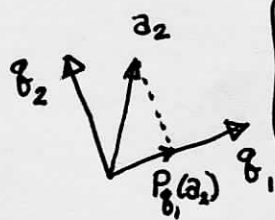
$$\hat{q}_3 := \frac{1}{\|a_3'\|} a_3'$$

etc...

- Key Idea:
- Take the next vector a_j
 - O.N. Proj it onto the subsp spanned by prev $\hat{q}_1, \dots, \hat{q}_{j-1}$
 - Subtract that off and get a new vector a_j' to that subsp, namely a_j'
 - Normalize a_j' to become \hat{q}_j .

$$a_j' := a_j - [\langle a_j, \hat{q}_1 \rangle \hat{q}_1 + \dots + \langle a_j, \hat{q}_{j-1} \rangle \hat{q}_{j-1}]$$

$$q_j := \frac{1}{\|a_j'\|} a_j'$$



Actually if $\{a_i\}$ is not LI, then we end up with fewer than n q_i 's

All main diag elts pos because $q_i^T a_i = \|a_i'\|$ $\Rightarrow R$ is invertible. P. 181 #3.4.19 says it is more stable numerically to subtract these one at a time.

A=QR factorization

The above eq gives us: $\|a_j'\| q_j = a_j - \sum_{i=1}^{j-1} \langle a_j, q_i \rangle q_i$

Apply q_j^T to both sides: $\|a_j'\| 1 = q_j^T a_j$

$$\Rightarrow a_j = \sum \langle a_j, q_i \rangle q_i$$

In matrix form:

Take $n=3$

Really $A = \begin{bmatrix} | & & | \\ & & \\ | & & | \end{bmatrix}$

$${}^n \begin{bmatrix} | & & | \\ a_1 & a_2 & a_3 \\ | & & | \end{bmatrix} = {}^n \begin{bmatrix} | & & | \\ q_1 & q_2 & q_3 \\ | & & | \end{bmatrix} \begin{bmatrix} q_1^T a_1 & q_1^T a_2 & q_1^T a_3 \\ 0 & q_2^T a_2 & q_2^T a_3 \\ 0 & 0 & q_3^T a_3 \end{bmatrix}$$

The main point of orthogonalization is to simplify the least sq problem:

$$A^T A = R^T Q^T Q R = R^T R$$

$$A^T A \bar{x} = A^T b \text{ becomes } R^T R \bar{x} = R^T b$$

$$\text{or just } R \bar{x} = Q^T b$$

since R invertible and can be cancelled.

R - upper triang and invertible, if A is. Only pos elts on main diag $\Rightarrow \det R > 0$

Ch 6 (Symmetric) Pos Def Matrices

①

For any $n \times n$ matrix A , define quadratic form $f: \mathbb{R}^n \rightarrow \mathbb{R}$
 $x \mapsto x^T A x = \sum_{i=1}^n \sum_{j=1}^n a_{ij} x_i x_j$

If A is anti-symm ($A^T = -A$), this reduces to pure quadratic
 $a_{11}x_1^2 + a_{22}x_2^2 + \dots + a_{nn}x_n^2$
 $[x \ y \ z] \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = ax^2 + (b+d)xy + ey^2 + (c+g)xz + (f+g)yz + iz^2$
 mixed terms vanish if A anti-symm.

Motivation

Consider a smooth fun $F: \mathbb{R}^n \rightarrow \mathbb{R}$. Taylor expand: $F(x) = F(0) + DF_0(x) + \frac{1}{2!} D^2 F_0(x,x) + o(\|x\|^3)$
 F has a local minima at 0 if $DF_0 = 0$ and $D^2 F_0(x,x) > 0 \ \forall x \neq 0$
 Let $A := D^2 F_0$. Then $DF_0(x,x) = x^T A x$ and $x^T A x > 0 \ \forall x \neq 0$ if A pos def.

Def A is pos def if $x^T A x > 0$ for all $x, x \neq 0$

Strang only considers Symm pos def matrices, so that the following then holds.

It certainly is possible for a non-symm matrix to satisfy $x^T A x > 0$

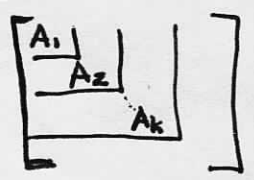
For example $A = \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix}$ $[x \ y] \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 2x^2 + 1xy + 3y^2 > 0$ for $x, y \neq 0$
 because $xy < 2x^2 + 3y^2$

To see that $xy < 2x^2 + 3y^2$: For pos numbers $a, b, a \neq b$ $0 < (b-a)^2 = b^2 - 2ab + a^2$
 $2ab < a^2 + b^2$
 Thus $xy \leq |x||y| < 2|x||y| < x^2 + y^2 < 2x^2 + 3y^2$ (if $x=y$ the result is even easier).

Another way: $f(x,y) := 2x^2 + xy + 3y^2$ $D^2 f_x = \begin{bmatrix} 4 & 1 \\ 1 & 6 \end{bmatrix}$ This is not A , but it is symm pos def.
 $Df_x = [4x+y, x+6y]$
 $Df_0 = [0 \ 0]$ critical pt.
 Thus, from calculus we would know $f(0,0) = 0$ is a local min (in fact global) and thus $f(x,y) > 0 \ \forall x, y \neq 0 \ \square$

A symm pos def matrix

- Thm T.F.A.E. (I) $x^T A x > 0 \ \forall x \neq 0$
- \Leftrightarrow (II) All EWs λ of A satisfy $\lambda > 0$
 - \Leftrightarrow (III) All upper left submatrices A_k have $\det(A_k) > 0$
 - \Leftrightarrow (IV) All pivots satisfy $d_i > 0$ (with no row exchanges)
 - \Leftrightarrow (V) \exists matrix R with LI cols $\ni A = R^T R$



Pf (I) \Rightarrow (II) Let $Ax = \lambda x$ where $\|x\| = 1$
 $0 < x^T A x = x^T \lambda x = \lambda \ \square$

(I) \Leftarrow (II) Since A symm, Spectral Thm says A has full set of O.N. EVs $\{x_1, \dots, x_n\}$ (basis of span)
 For any vector u , $u = \sum_{i=1}^n c_i x_i$
 $Ax = c_1 A x_1 + \dots + c_n A x_n = c_1 \lambda_1 x_1 + \dots + c_n \lambda_n x_n$ O.N.
 $x^T A x = (c_1 x_1^T + \dots + c_n x_n^T)(c_1 \lambda_1 x_1 + \dots + c_n \lambda_n x_n) = c_1^2 \lambda_1 + \dots + c_n^2 \lambda_n > 0$

(I) \Rightarrow (III) $x^T A x > 0 \ \forall x$, in particular if we fix $k \in \{1, \dots, n-1\}$ let $x_k^T = [x_k^T \ 0 \ \dots \ 0]$
 $\Rightarrow [x_k^T \ 0] \begin{bmatrix} A_k & * \\ * & * \end{bmatrix} \begin{bmatrix} x_k \\ 0 \end{bmatrix} = x_k^T A_k x_k > 0 \Rightarrow$ All EWs $\lambda_i^{(k)}$ of A_k are pos by II (last $n-k$ components)
 $\Rightarrow \det A_k = \lambda_1^{(k)} \dots \lambda_k^{(k)} > 0 \ \square$
 Thus a nec cond for A is each $a_{ii} > 0$.
 Not suff.
 There is nothing special about using upper left submatrices. Any chain of principal submats would work, starting from a diag elt a_{ii} .
 Principal submatrix A_k is starting with a_{ii} and adding new col and row pair each time.

(III) \Rightarrow (IV) From ch 4 section 4 we know $\det A_k = d_1 \cdot d_2 \cdots d_k$ a priori a pivot is not zero.
 Then $d_k = \frac{d_1 \cdot d_2 \cdots d_{k-1} d_k}{d_1 \cdots d_{k-1}} = \frac{\det A_k}{\det A_{k-1}} > 0$

(IV) \Rightarrow (I) Since A is symm, $A = LDL^T$ where D is diag matrix of pivots.
 $x^T A x = x^T (LDL^T) x = (x^T L) D (L^T x)$
 $= d_1 ((L^T x)^{(1)})^2 + d_2 ((L^T x)^{(2)})^2 + \dots + d_n ((L^T x)^{(n)})^2$

(I) \Leftarrow (V) $A = R^T R$ where R has n cols. This is always pos if each $d_i > 0$ \square

(I) \Rightarrow (V) There are many valid choices for R . $x^T A x = x^T R^T R x = \|R x\|_2^2 > 0$ if $x \neq 0$ and $R x \neq 0$ if $x \neq 0$ because R has n cols $\Rightarrow \ker(R) = \{0\}$
 $A = LDL^T \Rightarrow \begin{matrix} (L^T D) & (D^T L^T) \\ R^T & R \end{matrix}$ or $A = Q \Lambda Q^T$ spectral thm
 $= (Q \sqrt{\Lambda}) (\sqrt{\Lambda} Q^T)$
 $= R^T R$

And R can be rectangular if we like.
 Take any valid sq R and a matrix $Q = \begin{bmatrix} & \\ & \\ & \end{bmatrix}$ with O.N. cols.
 Rectangular $R := QR$

Then $R^T R = R^T [Q^T] \begin{bmatrix} Q \\ R \end{bmatrix} R = R^T R = A$ \square

In least squares, the normal eqs are $A^T A \bar{x} = A^T b$

This is now known to be Symm pos def!

Ellipsoids in n -dim

$x^T A x = 1$ defines an ellipsoid for ~~any~~ A (exclude extreme degenerate cases).
 A symm pos def (or else we might get hyperboloids, etc...)

$$\sum \sum a_{ij} x_i x_j = 1$$

Since A is symm, pos def $A = Q \Lambda Q^T$ where Q is O.N. and Λ has only pos elts.

Thm \exists rotation $Q \ni$ with the new co-ords $y = Q^T x$ is aligned with the co-ord axes. Thus the EVs of A are the ~~co-ord~~ ^{ellipsoid} ~~natural~~ axes of the ellipsoid and the length of axis i is $\frac{1}{\sqrt{\lambda_i}}$

Pf We already found $Q \ni A = Q \Lambda Q^T$ from spectral thm

If $\det Q = +1$, Q is rotation. We know $\det Q = \pm 1$, if $\det Q = -1$ replace Q with

$$1 = x^T A x = x^T Q \Lambda Q^T = y^T \Lambda y = \lambda_1 y_1^2 + \dots + \lambda_n y_n^2$$

For axis i , plug in $y_j = 0 \forall j \neq i \Rightarrow 1 = \lambda_i y_i^2 \Rightarrow y_i = \pm \frac{1}{\sqrt{\lambda_i}}$ \square

each $\lambda_i > 0$ $\left[-\frac{1}{\sqrt{\lambda_1}} \quad \frac{1}{\sqrt{\lambda_2}} \quad \dots \quad \frac{1}{\sqrt{\lambda_n}} \right]$ still EVs with same EWs.

Say we have $q(x) = x^T A x$ for A symm, pos def
 we want to make a linear COV $x = C y$ for C using matrix

$$q(x) = x^T A x = y^T C^T A C y = \tilde{q}(y)$$

$B = C^T A C$ is still symm pos def!

$B^T = (C^T A C)^T = C^T A^T C = C^T A C = B$ \checkmark
 For any $y \neq 0$, $y^T B y = y^T C^T A C y = (C y)^T A C y > 0$
 since $C y \neq 0$ because C nsing. \square