

8.4. Quadratic Forms

Quadratic forms generalize norm, lengths, inner-products, ..

Definition of a quadratic forms

- Sum of $a_{ij}x_ix_j$ for $i,j=1,2,\dots,n$ ($i < j$ written usually)
- Example: $f(x_1, x_2, x_3) = a_{11}x_1^2 + 2a_{12}x_1x_2 + 2a_{13}x_1x_3 + a_{22}x_2^2 + 2a_{23}x_2x_3 + a_{33}x_3^2$
- In matrix form $q(x) = x^T A x$ for A symmetric $n \times n$ -matrix.
- Note that $A_{ij} = a_{ij} = a_{ji}$... here
- If $A = I$, then $q(x) = x \cdot x = x \cdot x = \|x\|^2$.
- If $A = D$, diagonal, then $q(x) = l_1 x_1^2 + l_2 x_2^2 + \dots + l_n x_n^2$.

Change of variables in a quadratic form.

- We can use substitution $\mathbf{x} = P\mathbf{y}$ to simplify $q(\mathbf{x})$.
- This will help us solve many problems...
- Since A is symmetric, we can find P s.t. P^TAP is D .
- Then $\mathbf{x}^T A \mathbf{x} = \mathbf{y}^T P^T A P \mathbf{y} = \mathbf{y}^T D \mathbf{y}$.

Theorem 8.4.1 (The Principal Axes Theorem) *If A is a symmetric $n \times n$ matrix, then there is an orthogonal change of variable that transforms the quadratic form $\mathbf{x}^T A \mathbf{x}$ into a quadratic form $\mathbf{y}^T D \mathbf{y}$ with no cross product terms. Specifically, if P orthogonally diagonalizes A , then making the change of variable $\mathbf{x} = P\mathbf{y}$ in the quadratic form $\mathbf{x}^T A \mathbf{x}$ yields the quadratic form*

$$\mathbf{x}^T A \mathbf{x} = \mathbf{y}^T D \mathbf{y} = \lambda_1 y_1^2 + \lambda_2 y_2^2 + \cdots + \lambda_n y_n^2$$

in which $\lambda_1, \lambda_2, \dots, \lambda_n$ are the eigenvalues of A corresponding to the eigenvectors that form the successive columns of P .

- Example 2. $Q(x) = -23/25x_1^2 - 2/25x_2^2 + 72/25x_1x_2$

$$q(x) = x^T Ax = [x_1, x_2] \begin{bmatrix} -23/25 & 36/25 \\ 36/25 & -2/25 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

- We find the eigenvectors to diagonalize it.
- Then make it into an orthonormal set.
- Use $P = [v_1, v_2]$ eigenvectors. (One may need to orthogonalize it)

Quadratic forms in geometry

- $ax^2+2bxy+cy^2+dx+ey+f=0$.
- Set $d,e=0$.
- We wish to solve $ax^2+2bxy+cy^2+f =0$.
- We wish to turn it into $ax^2+cy^2+f =0$ by coordinate change.
- By dividing by $-f$, we obtain $a'x^2+b'y^2=1$.
- If $a,b>0$, then we obtain an ellipse or a circle.
- If $a>0,b<0$, or $a<0,b>0$ then we obtain a hyperbola
- If $a<0,b<0$, then an empty set.

Identifying conic sections

- We identify minor and major axis. Thus, basically, we have to rotate.
- This amounts to finding P.
- For \mathbb{R}^2 , P is always a rotation. Find the rotation angle.
- Example 3.
- Remark: $ax^2+2bxy+cy^2=k$. Rotate by the angle t s.t $\cos 2t=(a-c)/2b$.
Solution: Find eigenvectors for all a,b,c .

Positive definite quadratic forms

Definition 8.4.2 A quadratic form $\mathbf{x}^T A \mathbf{x}$ is said to be

positive definite if $\mathbf{x}^T A \mathbf{x} > 0$ for $\mathbf{x} \neq \mathbf{0}$

negative definite if $\mathbf{x}^T A \mathbf{x} < 0$ for $\mathbf{x} \neq \mathbf{0}$

indefinite if $\mathbf{x}^T A \mathbf{x}$ has both positive and negative values

Theorem 8.4.3 If A is a symmetric matrix, then:

- (a) $\mathbf{x}^T A \mathbf{x}$ is positive definite if and only if all eigenvalues of A are positive.
- (b) $\mathbf{x}^T A \mathbf{x}$ is negative definite if and only if all eigenvalues of A are negative.
- (c) $\mathbf{x}^T A \mathbf{x}$ is indefinite if and only if A has at least one positive eigenvalue and at least one negative eigenvalue.

- Positive semidefinite $x^T Ax \geq 0$ only if x is not 0.
- Negative semidefinite $x^T Ax \leq 0$ only if x is not 0.
- In higher dimensions, this is classified by the number of positive eigenvalues and negative eigenvalues and the multiplicity of 0 in the characteristic polynomial.

Classifying conics

- $x^T A x = 1$.
- A diagonalizes to $[[\lambda_1, 0], [0, \lambda_2]]$
- If $\lambda_1 > 0$ and $\lambda_2 > 0$, then ellipse.
- If $\lambda_1 < 0$ and $\lambda_2 < 0$, then no graph
- If $\lambda_1 \cdot \lambda_2 < 0$, then a hyperbola.

Theorem 8.4.4 *If A is a symmetric 2×2 matrix, then:*

- (a) $\mathbf{x}^T A \mathbf{x} = 1$ represents an ellipse if A is positive definite.*
- (b) $\mathbf{x}^T A \mathbf{x} = 1$ has no graph if A is negative definite.*
- (c) $\mathbf{x}^T A \mathbf{x} = 1$ represents a hyperbola if A is indefinite.*

- Positive semidefinite case: two lines L union $-L$
- Negative semidefinite case: empty set.

Identifying positive definite matrices.

- k-th principal submatrix of an $n \times n$ -matrix consists of the first k-rows intersected with first k-columns of A.

Theorem 8.4.5 *A symmetric matrix A is positive definite if and only if the determinant of every principal submatrix is positive.*

Theorem 8.4.6 *If A is a symmetric matrix, then the following statements are equivalent.*

- A is positive definite.*
- There is a symmetric positive definite matrix B such that $A = B^2$.*
- There is an invertible matrix C such that $A = C^T C$.*

- Proof: (a) \rightarrow (b): A is positive definite. D has only positive eigenvalues. $D=D_1^2$. $A=PD_1^2P^T = PD_1P^T PD_1P^T$.
 - Let $B=PD_1P^T$. B is symmetric.
 - Since D_1 has positive diagonals, B is positive definite.
- (b) \rightarrow (c): $A=B^2$. B symmetric positive definite. B is invertible. Take $C=B$.
- (c) \rightarrow (a): $A=C^T C$.
 - $x^T A x = x^T C^T C x = (C x)^T C x = C x \cdot C x = \|C x\|^2 > 0$ for x nonzero.
- Example 6.

Cholesky factorization

- $A = R^T R$. R is upper triangular and has positive entries in the diagonal.