

Bilinear and quadratic forms

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Bilinear form

$h : V \times V \rightarrow K$ where V is a vector space and h is bilinear is called bilinear form.

As usually, bilinear form is characterized by $a_{i,j} = h(e_i, e_j)$ which is called the matrix of h in basis e . Here is how to evaluate h through A :

$$h(u, v) = [u]_e^T A [v]_e$$

Use cases

- o Dot product: $x, y \mapsto \sum x_i y_i$
- o Convolution?: $f, g \mapsto \int_0^1 f(x)g(x)w(x) dx$

Symmetry

$\forall v, u : h(u, v) = h(v, u) \Rightarrow h$ — symmetric. It's equivalent to A being symmetric.

Examples of asymmetry:

- o Something with \mathbb{C} : $x, y \mapsto \bar{x}y$
- o Another convolution (it's usually antisymmetric): $f, g \mapsto \int_0^1 f'(x)g(x) dx$

Positive definite form

$$\forall x \neq 0 : h(x, x) > 0$$

Use cases

It allows decomposition into a direct sum by taking orthogonal vectors.

Def. $x \perp y \Leftrightarrow h(x, y) = 0$

Def (Orthogonal complement). $U \leq V, U^\perp = \{x \in V \mid \forall y \in U, x \perp y\}$

Th. 1. $h > 0$ — symmetric bilinear, $U \leq V$ then $V = U \oplus U^\perp$

Proof. 1. Suppose $0 \neq x \in U \cap U^\perp$. Then $x \neq 0 \Rightarrow h(x, x) > 0$, but $x \in U \cap U^\perp \Rightarrow h(x, x) = 0$.

2. Now let's show $\dim U + \dim U^\perp = \dim V$. Consider what equations define U^\perp :

$$\forall y \in U : h(x, e_i) = 0$$

Here e is the basis of U , and it's enough to check only with basis elements. So we have not more (they might be dependent, though they aren't) than $\dim U$ equations. So $\dim U^\perp \geq \dim V - \dim U$. \square

Quadratic form

Def. $q : V \rightarrow K$ which is a homogenous quadratic polynomial in every basis.

For $q(x) = \sum_{i \leq j} b_{i,j} x_i x_j$ let's define A this way:

$$a_{i,j} = \begin{cases} \frac{b_{i,j}}{2}, i < j \\ \frac{b_{j,i}}{2}, i > j \\ b_{i,j}, i = j \end{cases}$$

Now, both symmetric bilinear forms and quadratic forms both correspond to symmetric matrices, so they correspond to each other, and mapping is very simple: $h(x, y) = \frac{q(x+y) - q(x) - q(y)}{2}$.

Change of basis

Suppose h is $x^T A x$ in e and $y^T B y$ in f , then:

$$x^T A x = (C y)^T A C y = y^T (C^T A C) y, C = [id]_e^f$$

Note that if we find an orthogonal basis than A will be a diagonal matrix.

Th. 2. $q : V \rightarrow K, \exists e$ in which A (the matrix of q) will be diagonal.

Proof. Here $a_{ij} = a_{ji}$:

$$q(x) = \sum a_{ij} x_i x_j = a_{11} (x_1 + \sum_{j \geq 2} \frac{a_{1j}}{a_{11}} x_j)^2 + q(x_{2...n}) - p(x_{2...n})$$

Where:

$$p(x_{2...n}) = \sum_{j \geq 2} \frac{a_{1j}^2}{a_{11}}$$

Then we continue to extract squares from $q - p$ on $x_{2\dots n}$ and the first basis element is:

$$e_1 = x_1 + \sum \frac{a_{1j}}{a_{11}} x_j$$

Final transition matrix will be upper-triangular with ones on the main diagonal, so it will be nondegenerate and give a basis. If $a_{11} = 0$ we can swap this row with another. If all $a_{ii} = 0$, take some $i, j : a_{ij} \neq 0$ and go to $x'_i, x'_j = x_i + x_j, x_i - x_j$, there will appear a difference of squares. \square

We want to compute the main diagonal of matrix after this process.

Th. 3. $q(x) = x^T A x, d_i \neq 0$ — *principal minors (i.e. $\det A[:, i]$)*.

The main diagonal will be $\frac{d_i}{d_{i-1}}$, where $d_0 \stackrel{\text{def}}{=} 0$ (i.e. they won't change).

Proof. It's enough to proof that multiplying $A \in M_n(K)$ by $C \in UT_n(K)$ with ones on the main diagonal doesn't change its principal minors.

Actually, we consider only top-left $k \times k$ blocks and find the determinant of their product. But the determinant of UT_n with ones on the diagonal is 1, so the product is equal to the original determinant.

Now we need to show, that on each iteration $a_{ii} \neq 0$ so that only the main branch of the algorithms works, otherwise transition matrices wouldn't have the right form.

Consider the k -th step of the algorithm:

$$A^{(k)} = \begin{pmatrix} c_1 & & & 0 & 0 \\ & \dots & & 0 & 0 \\ & & c_k & 0 & 0 \\ 0 & 0 & 0 & x & \dots \\ 0 & 0 & 0 & \dots & \dots \end{pmatrix}$$

Here $x = a_{k+1,k+1} \neq 0$, otherwise the $(k+1)$ -th minor would be zero. More precisely, $x = \frac{d_{k+1}}{d_k}$. \square

Th. 4 (LU decomposition from the diagonalized form). *After the process from the previous theorem we get $A = C^T D C$, where $C \in UT_n(K)$ with ones on the diagonal, D — diagonal.*

Real case

By scaling the basis elements we can get $q(x) = \sum_{i=1}^k x_i^2 - \sum_{k+1}^l x_i^2$.

Def. (k, l) or simply $k - l$ is called signature

Th. 5. $k = \max\{\dim U \mid U \leq V, q|_U > 0\}$

Proof. If $q(x) = \sum_{i=1}^k x_i^2 - \sum_{i=k+1}^l x_i^2$ is true for some (k, l) , then $\max\{\dim U \mid U \leq V, q|_U > 0\} \geq k$, because we can take $\langle e_1 \dots e_k \rangle$.

Now suppose we found $U \leq V : \dim U \geq k+1$ and $q|_U > 0$. Also consider $W = \langle e_{k+1} \dots e_n \rangle$. We know $q|_W \leq 0$ and $U \cap W \neq \{0\}$ (from Grassman's formula). So we have a nonzero vector lying in the both forms, which is a contradiction. \square

Th. 6 (Stronger version of Sylvester criterion). $q(x) = x^T A x, A^T = A, d_i \neq 0, c_i = \frac{d_i}{d_{i-1}} \Rightarrow l = \text{number of sign changes in } 1, d_1, d_2 \dots d_n$, because each negative element on the diagonal (in canonical form) changes the sign of this and next minors.

Th. 7. Following are equivalent:

1. $\forall i : d_i > 0$
2. $\exists C \in UT_n(\mathbb{R}), \det C \neq 0 : A = C^T C$
3. $\exists C, \det C \neq 0 : A = C^T C$
4. $q > 0$

Proof. $1 \Rightarrow 2$: We can find $A = C^T D C$, then $A = (C^T \sqrt{D}^T)(\sqrt{D} C)$ as they are positive.

$2 \Rightarrow 3$: Trivial.

$3 \Rightarrow 4$: $x^T C^T C x > 0$ because with $y = Cx$, $y^T y \geq 0$ and $y \neq 0$ because $x \neq 0, \det C \neq 0$.

$4 \Rightarrow 1$: Positive definite form being restricted on anything is still positive (semi)definite. And k -th principal minor is a restriction on $\langle e_1 \dots e_k \rangle$. And it's determinant is nonzero because $A = C^T C$, where $|C| \neq 0$, so $|A| = |C|^2 > 0$. \square