



Master in
Computer Vision
Barcelona



T6: Background on vector calculus and Sylvester's criterion

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Optimization and inference techniques for Computer Vision

Material on multivariate calculus

If you have gaps in your background about multivariate calculus, Khan academy has some nice online course. These are the links to the course and the most relevant sections for our class.

Multivariate calculus at Khan Academy

Section on multivariable derivatives

Section on applications of multivariable derivatives

In particular, the section on applications of derivatives contains:

- The Hessian matrix
- First and second order Taylor approximations
- Optimization without constraints
- Optimization with equality constraints using the method of Lagrange multipliers

Sylvester's criterion

Remember James Joseph Sylvester, one of the inventors of the SVD? It turns out that he also came up with a criterion to check if a symmetric $n \times n$ matrix \mathbf{A} is positive definite using the principal minors.

\mathbf{A} is positive definite \iff its n leading principal minors are positive.

The **leading** principal minors are the determinants of all the top-left (square) sub-matrices of \mathbf{A} . Thus, if we compute the Hessian matrix of a function at a point \mathbf{x} and we find that its n leading principal minors are positive for any \mathbf{x} , then the function is **strictly convex** (because the eigenvalues of the Hessian are always strictly positive).

There is also a condition for checking if \mathbf{A} is positive semi-definite:

\mathbf{A} is positive semi-definite \iff **all** principal minors are non-negative.

This requires much more computations, because the total number of principal minors for an $n \times n$ matrix is $2^n - 1$. Whereas the number of **leading** principal minors is n . Therefore Sylvester's criterion for checking semi-definiteness is mainly useful for small matrices.