

# Sum of Squares

Labs in Mathematical Optimization

# Quick reminder on PSD matrices

- A matrix  $M$  is called **Positive-Semidefinite (PSD)**, if it is:
  1. Symmetric:  $M^T = M$
  2. Non-negative: for any vector  $x$ :  $x^T M x \geq 0$

Other equivalent formulations:

1.  $M = A^T A$  for some matrix  $A$
2.  $M = \sum_{i=1}^k \lambda_i x_i x_i^T$  for some  $\lambda_i \geq 0$  and  $x_i$  orthonormal vectors (spectral decomposition)

Remember that any symmetric matrix is diagonalizable with an orthonormal basis and real eigenvalues (do you remember what that means? :)

# Sum of squares

- Definition: A polynomial  $p(x) \in R[x]$  is called a **Sum of Squares** (SOS) if it can be written as  $p(x) = \sum_j p_j^2(x)$  where  $p_j(x)$  are some polynomials. We denote it as  $p(x) \in \Sigma[x]$
- Only polynomials of even degree can be SOS. (Show it)
- Other results:
  1. The set of SOS is closed, convex (Show it) and non-empty
  2. If a polynomial is SOS, then it's non-negative. Converse is not true in general

# Excercises

- Only polynomials of even degree can be SOS

Assume, that  $p(x)$  is SOS. Then there exist polynomials  $p_j(x)$ , such that:  $p(x) = \sum_j p_j(x)^2$ . Since the coefficients are non-zero, they can't cancel out in a way that they reduce the degree. So that the degree of the polynomial  $p$  is  $\max\{2\deg(p_j) \mid j \text{ is the coefficient enumerator}\}$ .

- SOS is a convex set

Let  $p, q \in \Sigma[x]$ , and let  $r \in [0,1]$ . Then  $p(x) = \sum_j p_j(x)^2$  and  $q(x) = \sum_{j'} q_{j'}(x)^2$ . Consider the convex combination  $rp(x) + (1-r)q(x) = \sum_j [\sqrt{r}p_j(x)]^2 + \sum_{j'} [\sqrt{1-r}q_{j'}(x)]^2$  so that any convex combination is also SOS.

# Sum of squares and PSD matrices

- A polynomial  $p(x) \in \mathbb{R}[x]$  of degree  $2d$  is SOS if and only if it can be written as  $v_x^T M v_x$  where  $M$  is some PSD matrix and  $v_x = (1, x, \dots, x^d)$

$\Rightarrow$  Let us assume that  $p(x)$  is SOS. Let us first notice that one can always write  $p(x)$  as  $v_x^T M v_x$ , where  $M$  is symmetric matrix (check it, hint: if  $M$  is not symmetric then  $M' = (M + M^T)/2$  is).  $M$  has to be PSD, because for the negative eigenvalues,  $p$  is not SOS.

$$\begin{aligned} \text{Crucial part: } v_x^T M v_x &= \sum_{i,j} m_{ij} x^i x^j = \sum_{i,j} \frac{1}{2} m_{ij} x^i x^j + \sum_{i,j} \frac{1}{2} m_{ij} x^i x^j = \sum_{i,j} \frac{1}{2} m_{ij} x^i x^j + \\ &\sum_{i,j} \frac{1}{2} m_{ji} x^i x^j = \sum_{i,j} \frac{1}{2} (m_{ij} + m_{ji}) x^i x^j = v_x^T \frac{1}{2} (M + M^T) v_x \end{aligned}$$

$\Leftarrow$  Let us assume that  $p(x) = v_x^T M v_x$  for some PSD. Then by eigen decomposition:  $M = \sum_j \lambda_j z_j z_j^T$  for some non-negative  $\lambda_j$  and  $p(x) = \sum_j [\sqrt{\lambda_j} z_j^T v_x]^2$  and  $z_j^T v_j = \sum_{a=0}^d (z_j)_a x^a$ , then  $p(x)$  is SOS

## Task (obligatory)

- Check whether polynomial  $p(x) = 5x^4 - 4x^3 - x^2 + 2x + 2$  is SOS and if yes, then find the decomposition of it. Write a test, to check the result.

## More challenging task\*

- Write a function *sum\_of\_squares* that takes a list representing the coefficients of the polynomial, and then returns the list of polynomials from the SOS decomposition. Remember to handle the exceptions properly if the solution is infeasible.

# The ultimate challenge\*\*

- Write a function `sum_of_squares_2` that takes a list, representing the multivariate polynomial and then returns its decomposition into SOS
- *Hint: just figure out what should be the dimension and the form of the vectors of monomials. Conceptually, the rest is pretty much the same.*