

# SIMULATION AND OPTIMIZATION

## Lab 1

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# *Project assignment*

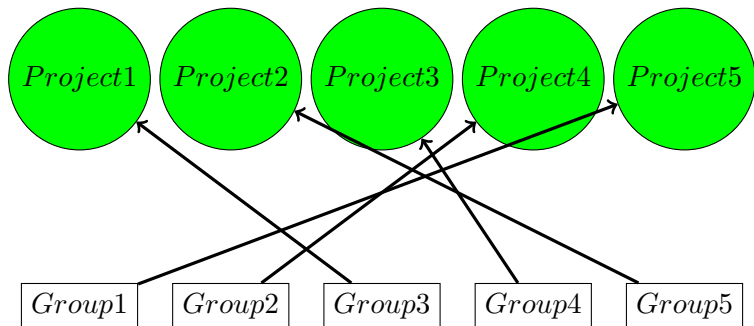
# Groups

Group	Members
1	Flavio Finger, Jonas Gros, Evanthia Kazagli
2	Jessie Madrazo, Florian Lucker, Julien Monge
3	Marija Nikolic, Tomas Robenek
4	Paul Anderson, Franz Zeimetz, Alberto Mian
5	Sofia Samoili, Marco Vocialta, Seyedeh Taheri

# Projects

Project	Optimization algorithm Matlab coding
1	Preconditioned Projected Gradient Method
2	Interior Point Method
3	Augmented Lagrangian Method
4	Local Sequential Quadratic Programming Method
5	Global Sequential Quadratic Programming Method

# Assignment



# *Some tips for Matlab coding*

## Some tips

- Matrix multiplication:  $y = (AB)x = A(Bx)$ , which way is better?
- To solve the linear equation system  $Ax = b$  where  $A$  is square, in Matlab,  $x = A \setminus b$  is more efficient than  $x = inv(A) * b$
- Matlab discriminates in favor of upper-right triangular matrices for inversion! If your matrix is lower-left triangular, first transpose it, invert the result, and transport back.

# *Exercises today*



# Line search

- Objective: find a step  $\alpha^*$  such that both Wolfe's conditions are verified.
- Input:
  - 1 Function  $f : \mathcal{R}^n \rightarrow \mathcal{R}$ , continuously differentiable
  - 2 Gradient  $\nabla f : \mathcal{R}^n \rightarrow \mathcal{R}^n$
  - 3 Vector  $x \in \mathcal{R}^n$
  - 4 Descent direction  $d$  such that  $\nabla f(x)^T d < 0$
  - 5 First approximation of the solution  $\alpha_0 > 0$
  - 6 Parameters  $\beta_1$  and  $\beta_2$  such that  $0 < \beta_1 < \beta_2 < 1$  (e.g.,  $\beta_1 = 10^{-4}$  and  $\beta_2 = 0.99$ )
  - 7 Parameter  $\lambda > 1$

# Line search

- Initialization:  $i = 0, \alpha_l = 0, \alpha_r = +\infty$
- Iterations:
  - ① If  $\alpha_i$  verify both conditions, then  $\alpha^* = \alpha_i$ . STOP.
  - ② If  $\alpha_i$  violates Wolfe 1, then the step is too long and

$$\alpha_r = \alpha_i$$

$$\alpha_{i+1} = \frac{\alpha_l + \alpha_r}{2}$$

- ③ If  $\alpha_i$  verifies Wolfe 1 and violate Wolfe 2, then the step is too short and

$$\alpha_l = \alpha_i$$

$$\alpha_{i+1} = \begin{cases} \frac{\alpha_l + \alpha_r}{2}, & \text{if } \alpha_r < +\infty; \\ \lambda \alpha_i, & \text{otherwise.} \end{cases}$$

- ④  $i = i + 1$

# Line search

Try your code for the following example:

$$f(x_1, x_2) = x_1^2 + x_2^2$$

- Starting point: (1,1)
- $d = (-0.5, -1)$

# Modified Cholesky Factorization

- Objective: Modify a matrix in order to make it positive-definite
- Input: Symmetric matrix  $A \in \mathcal{R}^{n \times n}$
- Output: A lower triangular matrix  $L$  and  $\tau \geq 0$  such that  $A + \tau I = LL^T$  is positive-definite

# Modified Cholesky Factorization

Frobenius Norm of a matrix:

$$\|A\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2}$$

- Initialization:  $k = 0$ ; if  $\min_i a_{ii} > 0$ , then  $\tau_k = 0$ , otherwise  $\tau_k = \frac{1}{2}\|A\|_F$
- Iterations:
  - ① Compute the Cholesky factorization  $LL^T$  of  $A + \tau_k I$
  - ② If the factorization is successful, STOP.
  - ③ Else  $\tau_{k+1} = \max\{2\tau_k, \frac{1}{2}\|A\|_F\}$
  - ④  $k = k + 1$

Matlab Hint:

- `norm(A, 'fro')`
- `chol`
- `try...catch...end`

# Modified Cholesky Factorization

Try your Matlab code on the following matrix:

$$A = \begin{bmatrix} 6 & 3 & 4 & 8 \\ 3 & 6 & 5 & 1 \\ 4 & 5 & 10 & 7 \\ 8 & 1 & 7 & 5 \end{bmatrix}$$