## Course A5: Convex Analysis / Optimization and Applications

Charles Dossal - dossal@insa-toulouse.fr
François Malgouyres - francois.malgouyres@math.univ-toulouse.fr
Aude Rondepierre - aude.rondepierre@insa-toulouse.fr

## Abstract:

This course is meant to provide a broad introduction to Convex analysis; numerical optimization; and to study the properties of optimization models commonly used in image processing. The content of the course will include:

- 1. Elements of convex analysis: Convexity (strict/strong), continuity, lower semi-continuity, functions with a lipschitz gradient, sub-differential. Coercivity. Existence of a minimizer. Optimality conditions.
- 2. Constrained optimization: Lagrange multipliers, Karush-Kuhn-Tucker Conditions.
- 3. Duality: Legendre-Fenchel transform. Fenchel-Rockafellar duality.
- 4. Numerical optimization: gradient descent, Newton and quasi-Newton algorithms. Non-smooth optimization: proximal gradient algorithm, Accelerated proximal gradient algorithm, projected gradient algorithm.
- 5. Applications in image processing:
  - Compressed sensing: Sparse modeling, algorithms and their guarantees ( $\ell^0$  and  $\ell^1$  minimization under Restricted Isometry Property)
  - Dictionary learning: Model, algorithms and their guarantees.

## Prerequisites:

Differential calculus, linear algebra, Functional analysis.

## References:

- 1. R.T. Rockafellar, Convex analysis. Princeton University Press, 1970.
- 2. Y. Nesterov, Introductory lectures on convex optimization: A basic course Kluwer Academic Publishers, 2004.
  - 3. D. P. Bertsekas, Nonlinear Programming Athena Scientific, 2003.
- 4. M. Elad, Sparse and Redundant Representations: From Theory to Applications in Signal and Image Processing Springer, 2010.