

# FrameLab: Development Guide

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## 1 Overall Design

FrameLab 1.0 is designed to have an object-oriented, user friendly scripting interface with compute intensive routines written in compiled languages such as C and CUDA/C. The current scripting language is Matlab, using MEX as an interface mechanism to pull in compiled libraries. In the future we plan to implement Python/iPython as an alternative to Matlab, to keep the entire code open-source.

## 2 Matlab OOP System

The primary goal of Framelab is to solve general linear inverse problems of the sort

$$Au = f_0 + \eta$$

FrameLab supports models of the type:

$$\min R(u) \quad \text{such that} \quad F(u) < \epsilon$$

where  $R(u)$  is a generic regularization term, typically of the form

$$R(u) = \|Wu\|_1$$

## 3 Compute Kernels

**Alternating Direction Method of Multipliers** Recall that ADMM is designed to solve problems of the sort

$$(x^*, y^*) := \arg \min_{x, y} F(x) + G(y) \quad \text{s.t.} \quad Ax + By = b \quad (\mathcal{P})$$

The approach is to consider the Augmented Lagrangian:

$$\mathcal{L}_\rho(x, y, \lambda) := F(x) + G(y) + \langle \lambda, Ax + By - b \rangle + \frac{\rho}{2} \|Ax + By - b\|_2^2$$

We then consider the saddle point problem

$$(x^*, y^*, \lambda^*)_\rho = \arg \min_{(x,y)} \arg \max_\lambda \mathcal{L}_\rho(x, y, \lambda) \quad (1)$$

Since we are interested in the saddle point itself and not the value of the functionals, we may complete the square in the definition of  $\mathcal{L}_\rho$  to obtain

$$(1) = \arg \min_{(x,y)} \arg \max_\lambda F(x) + G(y) + \frac{\rho}{2} \|Ax + By - (b - \lambda/\rho)\|_2^2$$

For notational convenience, we define

$$\mathcal{L}_\rho^*(x, y, \lambda) := F(x) + G(y) + \frac{\rho}{2} \|Ax + By - (b - \lambda/\rho)\|_2^2$$

If we then perform coordinate descent/ascent, we arrive at the 3-step ADMM scheme:

$$\begin{cases} x^{(k+1)} &= \arg \min_x \mathcal{L}_\rho^*(x, y^{(k)}, \lambda^{(k)}) \\ y^{(k+1)} &= \arg \min_y \mathcal{L}_\rho^*(x^{(k+1)}, y, \lambda^{(k)}) \\ \lambda^{(k+1)} &= \arg \max_\lambda \mathcal{L}_\rho^*(x^{(k+1)}, y^{(k+1)}, \lambda) \end{cases}$$

The method can be generalized in the particular case that  $F(x) + G(y)$  is further separable, e.g.  $F(x) + G(y) = F_1(x_1) + F_2(x_2) + \dots + F_n(x_n)$ , with  $A\mathbf{x} = \mathbf{b}$ , where  $\mathbf{x} = (x_1, \dots, x_n)^T$ . The augmented Lagrangian then takes the form

$$\mathcal{L}_\rho(\mathbf{x}, \Lambda) := \sum F_i(x_i) + \langle \Lambda, A\mathbf{x} - \mathbf{b} \rangle + \frac{\rho}{2} \|A\mathbf{x} - \mathbf{b}\|_2^2$$

where  $\Lambda = (\lambda_1, \dots, \lambda_n)^T$ .

In FrameLab, we have implemented an ADMM object designed to solve  $\mathcal{P}$ .

## 4 Another Section