# 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:

$$\min R(u)$$
 such that  $F(u) < \epsilon$ 

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

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

### 3 Compute Kernels

#### 3.1 Computed Tomography

From [?]

#### Compiling MEX Libraries

mex -L"/usr/local/cuda/lib64" -lcudart -I"./" Ax\_fan\_mf.cpp Ax\_fan\_mf\_cpu\_siddon.cpp
Ax\_fan\_mf\_cpu\_new.cpp Ax\_fan\_mf\_cpu\_new\_fb.cpp Ax\_fan\_mf\_gpu\_siddon.cu
Ax\_fan\_mf\_gpu\_new.cu Ax\_fan\_mf\_gpu\_new\_fb.cu find\_area.cpp sort\_alpha.cpp

Possible error message about invalid conversion from int to mxComplexity: change

plhs[0]=mxCreateNumericMatrix(nx\*ny\*nt,1,mxSINGLE\_CLASS,0);

to

plhs[0] = mxCreateNumericMatrix(nx\*ny\*nt,1,mxSINGLE\_CLASS,mxREAL);

in any mex interface files

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

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

The approach is to consider the Augmented Lagrangian:

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

We then consider the saddle point problem

$$(x^*, y^*, \lambda^*)_{\rho} = \operatorname*{arg\,min}_{(x,y)} \operatorname*{arg\,max}_{\lambda} \mathscr{L}_{\rho}(x, y, \lambda) \tag{1}$$

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

For notational convenience, we define

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

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

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

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

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

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

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

### 4 Another Section