# **1** Project Options

Solve some other related (application) optimization problem that seems useful and exciting to you. You can look deeper into something we have done, or use the class as a spring board for exploration into the "next level" of strategies: *e.g.* Constrained problems — Linear and Quadratic Programming, Interior Point methods, Penalty Methods, etc...

#### **Some Examples of Past Projects**

- Conjugate Gradient Based Methods
- Finding Steady State Vortex Solutions to the 2D Nonlinear Schrodinger Equation Using Optimization Algorithms
- Generating Time-Independent Solutions to the Gross-Pitaevskii Equation
- Image Restoration
- Influence of Step Size in Finite differencing of Hessian for Unconstrained Minimization Test Functions
- Nonlinear Parameter Fitting with Applications to Mathematical Biology
- Linear Least Squares QR Decomposition; Conjugate Gradient
- Optimization of a Traveling Numeric Soliton Like Solution to the Nonlinear Schrodinger Equation
- Optimization of Laguerre Type Orbital Basis Sets

#### Some Examples of Past Test Functions

- Branin Function
- Dixon Price Function
- Easom Function
- Extended Powell Function
- Helical Valley Function
- Powell Singular Function
- Trigonometric Function
- Wood Function

### Some Examples of Past Implementation Environments

- C++
- Fortran
- Intel Math Kernel Library
- Mathematica
- Matlab
- Python w/NumPy Library

# 2 The Math 693a "Default Project" — No Longer an Option

#### Not-an-option as of Fall 2018

The solution(s) to this project options are downloadable (with very revealing misspellings of variable names) from "Uncle Google."

#### Milestone #1



**Figure:** Code structure for the project. The top block contains all the drivers, common linear algebra solvers, stopping criteria checking, etc... The hess-fn-grad triplet defines the test function, its gradient and hessian; you will need one such "block" for each test function. The bottom four boxes define 4 different strategies.

The first milestone is to implement — from *Dennis & Schnabel* Appendix A — the driver block, the linesearch strategy, for one test function from Appendix B (see handout) — if you pick Rosenbrock then you can compare results with previous homework. This is STRONGLY recommended.

#### Milestone #2

The second milestone is to implement (at least) one additional strategy, and add (at least) one additional test function from Appendix B to your code-base. If you "only" implement one additional strategy, it is a BAD IDEA to select linesearchmod.

#### Milestone #3

Add the following to your code base:

- fdhessg Finite difference approximation to the Hessian using analytic gradient. (Executed when analgrad=TRUE, analhess=FALSE, cheapf=TRUE.)
- fdjac The core call from fdhessg, note that fvec in the pseudo-code corresponds to your analytic gradient.
- fdgrad Finite difference (forward) approximation to the gradient. (Executed when analgrad=FALSE.)
- fdhessf Finite difference approximation to the Hessian using only function values. (Executed when analgrad=FALSE, analhess=FALSE, cheapf=TRUE.)

#### Milestone #4

Add something else to your framework: *e.g.* Conjugate Gradient as an overall solution strategy, or as Linear Systems solver "only", or in the form of Steihaug's method.

#### Things to Explore (not an Exhaustive List)

- Test Functions
  - (At least) 2 Different test functions (see Appendix B, and other resources)
  - Different starting points (see Appendix B)
  - Different dimensionality (n)
- Finite Differencing
  - Analytic everything (no FD)
  - Analytic gradient (fdhessg+fdjac)
  - Finite difference everything (fdhessf+fdgrad)
  - optimal and non-optimal  $\epsilon$

## 3 "Deliverables"

- 10-15 minute presentation at the end of the semester
- Soft-copy of presentation
- Soft-copy of code

Time Management vs. Inspiration

