

## MORE DETAILS ON GETTING YOUR AUGMENTED LAGRANGIAN CODE WORKING.

Remember that the goal is to formulate "stand-alone" code that can solve any appropriately formulated optimization problem.

This means that for:  $\min f(x)$  s.t.  $C_E(x)=0$ ,  $C_I(x) \geq 0$  where

$$C_E(x)^T = [\dots C_i(x) \dots] \quad i \in \mathcal{E} \quad \text{and} \quad C_I(x)^T = [\dots C_i(x) \dots] \quad i \in \mathcal{I}.$$

a problem is defined by a structure variable / dictionary / class:

- pr. obj  $\rightarrow$  function handle to  $f$  and  $\nabla f$  computation
- pr. eom  $\rightarrow$  function handle to  $C_E$  and  $h_E = \nabla C_E$  computation
- pr. iom  $\rightarrow$  function handle to  $C_I$  and  $h_I = \nabla C_I$  computation
- pr. x0  $\rightarrow$  initial point in  $\mathbb{R}^n$ .
- pr. par  $\rightarrow$  parameters necessary for computing  $f, g, C_E$  etc.  
(and hyperparameters)

This is all  
that the  
user  
supplies!

The problem description can be complicated because there are two distinct (but related) problems to keep track of.

PR  $\rightarrow$  The user-defined problem

SPR  $\rightarrow$  the internally-defined unconstrained subproblem

Subproblem SPR will call the Augmented Lagrangian objective function to compute  $F := L_A(w, \lambda, \mu)$  and  $G := \nabla_x L_A(w, \lambda, \mu)$  where  $w^T = [x^T y^T]$  and the  $y$ 's are slack variables used to convert the inequality constraints into equality constraints. ( $C_I(x) \geq 0 \rightarrow C_I(x) - y^2 = 0$ ).

Not performed by the user

So, ...

SPR.obj = @ALOBJ

SPR.par = PR.par

SPR.par.obj = PR.obj

SPR.par.icon = PR.icon

SPR.par.econ = PR.econ

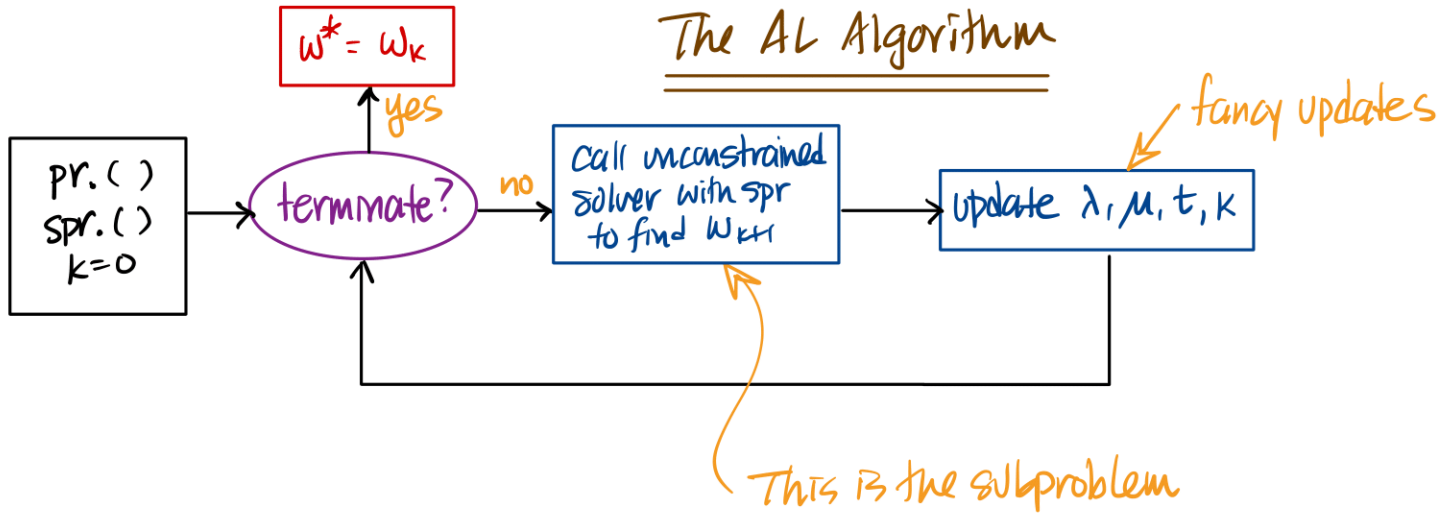
SPR.par.mu = mu

SPR.par.lambda = lambda

$\leftarrow$  internally use this function to compute objective and gradient of  $L_A$

with parameters it needs.

## The AL Algorithm



Define  $C(w) = C(x, y) = \begin{bmatrix} C_E(x) \\ C_I(x) - y^2 \end{bmatrix}$

$$F := L_A(w; \lambda; \mu) = f(x) - \lambda^T C(w) + \frac{1}{2} \mu C(w)^T C(w)$$

$$\begin{aligned} G := \nabla_w L_A(w; \lambda; \mu) &= \begin{bmatrix} \nabla_x f(x) \\ \nabla_y f(x) \end{bmatrix} - \lambda^T \begin{bmatrix} C_E(x) \\ C_I(x) - y^2 \end{bmatrix} + \frac{1}{2} \mu \begin{bmatrix} C_E(x) \\ C_I(x) - y^2 \end{bmatrix}^T \begin{bmatrix} C_E(x) \\ C_I(x) - y^2 \end{bmatrix} \\ &= \begin{bmatrix} \nabla_x f(x) \\ 0 \end{bmatrix} - \begin{bmatrix} \nabla_x C_E(x) & \nabla_x C_I(x) \\ 0 & -2y \end{bmatrix} \left( \lambda - \mu \begin{bmatrix} C_E(x) \\ C_I(x) - y^2 \end{bmatrix} \right) \end{aligned}$$

Skeleton of ALOBJ:

function [F,G] = ALOBJ(w,P)

this is spr.par

n = length(w) - P.I

P.I and P.E are the number of ineq./eq. constraints

x = w(1:n); y = w(n+1:end)

[f,g] = P.obj(x,P)

[C\_E, h\_E] = P.econ(x,P)

[C\_I, h\_I] = P.icon(x,P)

} calling  
user-defined  
functions

This illustrates the case with  
both eq. and ineq. constraints  
and computing both F and G.

Y = diag(y)

C = [C\_E; C\_I - Y\*Y]

F = f - λ'\*C + (μ/2)\*C'\*C

G = [g; zeros(P.I,1)] - [C\_E C\_I; zeros(P.E,n) - 2\*Y]\*[λ - μ\*C]

return

Notice that this is a fixed problem-independent function.

The specific user-defined problem is only in the computations  
of f,g,C\_E etc and defined by the user elsewhere.