

# Computational Optimization

## Constrained Optimization



# Easiest Problem

- Linear equality constraints


$$\min f(x) \quad f \in R^n$$

$$s.t. \quad Ax = b \quad A \in R^{m \times n}, \quad b \in R^m$$





# Null Space Representation

- Let  $x^*$  be a feasible point,  $Ax^*=b$ .
  - Any other feasible point can be written as  $x=x^*+p$  where  $Ap=0$
  - The feasible region  
 $\{x : x^*+p \quad p \in N(A)\}$   
where  $N(A)$  is null space of  $A$
- 



# Example

$$\min \frac{1}{2} (x_1^2 + x_2^2 + x_2^3)$$


$$s.t. \quad x_1 + 3x_2 + 4x_3 = 4$$

• Solve by substitution

$$\min \frac{1}{2} (x_1^2 + x_2^2 + x_2^3)$$

$$s.t. \quad x_1 = 4 - 3x_2 - 4x_3$$

becomes

$$\min \frac{1}{2} \left( (4 - 3x_2 - 4x_3)^2 + x_2^2 + x_2^3 \right)$$


# Null Space Method

$$\min \frac{1}{2} (x_1^2 + x_2^2 + x_3^2)$$

$$s.t. \quad x_1 + 3x_2 + 4x_3 = 4$$

$$x^* = [4 \ 0 \ 0]'$$

$$Z = \begin{bmatrix} -3 & -4 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$x = x^* + v \quad \begin{bmatrix} 4 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} -3 & -4 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 4 - 3v_1 - 4v_2 \\ v_1 \\ v_2 \end{bmatrix}$$

becomes  $\min \frac{1}{2} \left( (4 - 3v_1 - 4v_2)^2 + v_1^2 + v_2^2 \right)$

# Variable Reduction Method

- Let  $A=[B \ N]$  for  $x^*$  (a basic feasible solution with at most  $m$  nonzero variables corresponding to columns of

$$B) \quad A \in R^{m \times n} \quad B \in R^{m \times m} \quad N \in R^{m \times (n-m)} \quad I \in R^{(n-m) \times (n-m)}$$

$$Z = \begin{bmatrix} -B^{-1}N \\ I \end{bmatrix} \quad \text{assumes } m < n$$

is a basis matrix for null space of  $A$

$$A_r = \begin{bmatrix} B^{-1} \\ 0 \end{bmatrix} \Rightarrow AA_r = [B \ N] \begin{bmatrix} -B^{-1} \\ 0 \end{bmatrix} = I + 0$$

# Where did Z come from?

$$A=[1 \ 3 \ 4] \quad x^* = [4 \ 0 \ 0]$$

$$A=[B \ N]$$

$$B=[1] \quad N = [3 \ 4]$$

$$Z = \begin{bmatrix} -B^{-1}N \\ I \end{bmatrix} = \begin{bmatrix} -1*[3 \ 4] \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} -3 & -4 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$



# General Method


- There exists a Null Space Matrix

$$Z \in R^{n \times r} \quad r \geq n - m$$

- The feasible region is:

$$\{x \mid x^* + Zv\}$$

- Equivalent “Reduced” Problem


$$\min_v f(x^* + Zv)$$




# Practice Problem

$$\min \frac{1}{2} (x_1^2 + x_2^2 + x_3^2)$$

$$s.t. \quad x_1 + 3x_2 + 4x_3 = 4$$

$$x_2 - x_3 = 0$$


# Optimality Conditions

- Assume feasible point and convert to null space formulation

$$g(v) = f(x^* + Zv)$$

$$\nabla g(v) = Z' \nabla f(x^* + Zv) = Z' \nabla f(y) = 0 \quad \text{where } z = x^* + Zv$$

$$\nabla^2 g(v) = Z' \nabla^2 f(x^* + Zv) Z = Z' \nabla^2 f(y) Z$$

# Lemma 14.1 Necessary Conditions (Nash + Sofer)

- If  $x^*$  is a local min of  $f$  over  $\{x|Ax=b\}$ , and  $Z$  is a null matrix


$$\Rightarrow Z' \nabla f(x^*) = 0$$

*and  $Z' \nabla^2 f(x^*) Z$  is p.s.d.*

- Or equivalently use KKT Conditions

$$\left. \begin{aligned} \Rightarrow \nabla f(x^*) - A' \lambda &= 0 \\ Ax^* &= b \end{aligned} \right\} \text{has a solution}$$

$$\Rightarrow Z' \nabla^2 f(x^*) Z \text{ is p.s.d.}$$



# Lemma 14.2 Sufficient Conditions (Nash + Sofer)

- If  $x^*$  satisfies (where  $Z$  is a basis matrix for  $\text{Null}(A)$ )

$$Ax^* = b$$

$$Z' \nabla f(x^*) = 0$$

$$Z' \nabla^2 f(x^*) Z \text{ is } p.d.$$

then  $x^*$  is a strict local minimizer



# Lemma 14.2 Sufficient Conditions (KKT form)

- If  $(x^*, \lambda^*)$  satisfies (where  $Z$  is a basis matrix for  $\text{Null}(A)$ )

$$Ax^* = b$$

$$\nabla f(x^*) - A' \lambda = 0$$

$$Z' \nabla^2 f(x^*) Z \text{ is } p.d.$$

then  $x^*$  is a strict local minimizer

# Lagrangian Multiplier

- $\lambda^*$  is called the Lagrangian Multiplier
- It represents the sensitivity of solution to small perturbations of constraints

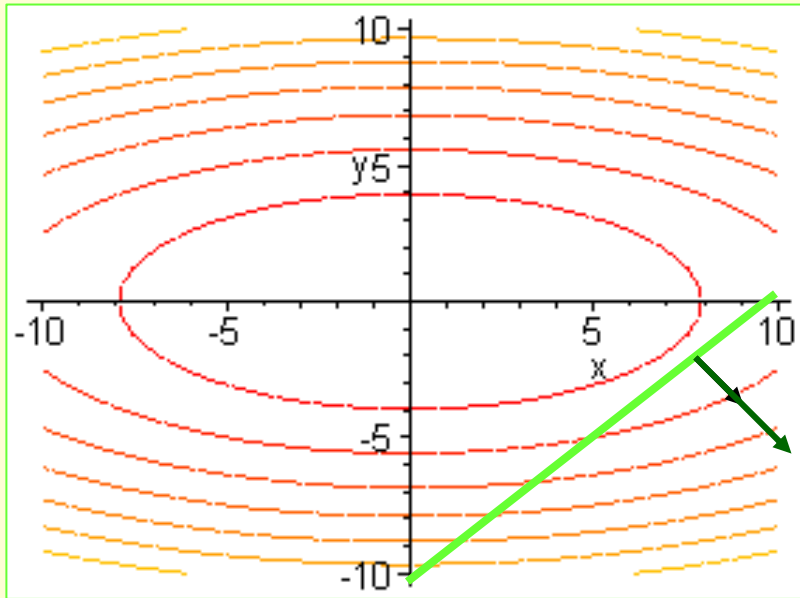
$$\begin{aligned} f(\hat{x}) &\approx f(x^*) + (\hat{x} - x^*)' \nabla f(x^*) \\ &= f(x^*) + (\hat{x} - x^*)' A' \lambda^* \quad \text{by KKT OC} \end{aligned}$$

Now let  $A\hat{x} = b + \delta$

$$= f(x^*) + \delta' \lambda^* = f(x^*) + \sum_{i=1}^m \delta_i \lambda_i^*$$

# Optimality conditions

- Consider  $\min (x^2+4y^2)/2$  s.t.  $x-y=10$



$$\nabla f(x) - A' \lambda = 0$$

$$Ax = b$$

$\Leftrightarrow$

$$\begin{bmatrix} x \\ 4y \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \lambda$$

$$x - y = 10$$

$\Leftrightarrow$

$$x^* = \lambda^* = 8, y^* = 2,$$

# Optimality conditions

## Find KKT point

$$\nabla f(x) - A' \lambda = 0$$

$$Ax = b$$

$\Leftrightarrow$

$$\begin{bmatrix} x \\ 4y \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \lambda$$

$$x - y = 10$$

$\Leftrightarrow$

$$x^* = \lambda^* = 8, y^* = -2,$$

## Check SOSOC

$$Z' = [1 \ 1]$$

$$\nabla^2 f(x) = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}$$

$Z' \nabla^2 f(x) Z$  is *p.d.*

So SOSOC satisfied

Or we could just observe that it is a convex program so FONC are sufficient

# Linear Equality Constraints - I

$$\min \frac{1}{2}(x_1^2 + 4x_2^2)$$

$$\text{s.t. } x_1 - x_2 = 10$$

$$\nabla f(x) = A^T \lambda \quad Ax = b$$

$$\nabla f(x) = \begin{bmatrix} x_1 \\ 4x_2 \end{bmatrix} \quad A = [1 \quad -1]$$

$\therefore$

$$\begin{bmatrix} x_1 \\ 4x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \lambda$$

$$x_1 - x_2 = 10$$

# Linear Equality Constraints - II

Solve:

$$x_1 = \lambda \Rightarrow 4x_2 = -x_1 \Rightarrow x_1 = -4x_2$$

$$-4x_2 - x_2 = 10 \Rightarrow$$

$$-5x_2 = 10$$

$$x_1 = 8$$

$$x_2 = -2$$

$$\lambda = 8$$

$$x^* = \begin{bmatrix} 8 \\ -2 \end{bmatrix}, \quad \lambda^* = 8 \quad \leftarrow \text{KKT point}$$

# Linear Equality Constraints - III

$$\text{SOSC} \quad A = [1 \quad -1] \quad Z = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\nabla^2 f(x) = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}$$


$$Z^T \nabla^2 f(x) Z = [1 \quad 1] \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} > 0$$

so SOSC satisfied, and  $x^*$  is a strict local minimum

Objective is convex, so KKT conditions are sufficient.



# Handy ways to compute Null Space

- Variable Reduction Method
  - Orthogonal Projection Matrix
  - QR factorization (best numerically)
  - $Z = \text{Null}(A)$  in matlab
- 

# Orthogonal Projection Method

- Use optimization. Minimize distance between given point  $c$  and null space of  $A$ .

$$\min_p \frac{1}{2} \|p - c\|^2$$

$$s.t. \quad Ap = 0$$

$$\nabla f(p^*) = A' \lambda$$

$$Ap^* = 0$$

*or equivalently*

$$(p^* - c) = A' \lambda$$

$$Ap^* = 0$$



# Orthogonal Projection Method

Optimality conditions give us the solution

*FONC is*


$$(p^* - c) = A' \lambda$$

$$Ap^* = 0$$

$$\Rightarrow Ap^* - Ac = AA' \lambda$$

$$\Rightarrow \lambda = -(AA')^{-1} Ac$$

$$\Rightarrow p^* = A' \lambda + c = -A'(AA')^{-1} Ac + c$$

$$= (I - A'(AA')^{-1} A)c$$





# Orthogonal Projection Method

Final result is:

$$(I - A'(AA')^{-1}A) \in \text{Null Matrices of } A$$

Note null space matrix is not unique

Try it in Matlab for  $A = [1 \ 3 \ 5; 2 \ 4 \ -1]$   
Compare with `Null(A)` `Null('A',r)`



# Get Lagrangian Multipliers for free!

- The matrix

$$A_r = A'(AA')^{-1} \text{ where } AA_r = AA'(AA')^{-1} = I$$

is the right inverse matrix for A.

- For general problems

$$\min f(x) \text{ s.t. } Ax = b$$

$$\lambda^* = A_r' \nabla f(x^*)$$

# Let's try it

• For  $\min f(x) = \frac{1}{2} [x_1^2 + x_2^2 + x_3^2 + x_4^2]$   
*s.t.*  $x_1 + x_2 + x_3 + x_4 = 1$

• Projection matrix

$$Z = I - A'(AA')^{-1}A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{3}{4} & -\frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{3}{4} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{3}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} & \frac{3}{4} \end{bmatrix}$$

# Solve FONC for Optimal Point

## ● FONC

$$\nabla f(x) - A' \lambda = \begin{bmatrix} x_1 \\ x_2 \\ x \\ x_4 \end{bmatrix} = \lambda \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$x_1 + x_2 + x_3 + x_4 = 1$$

# Check Optimality Conditions

For

$$\begin{aligned}x^* &= [1111]/4 \\ \nabla f(x^*) &= [1111]/4 \\ Ax^* &= b\end{aligned}$$
$$Z\nabla f(x^*) = \begin{bmatrix} \frac{3}{4} & -\frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & \frac{3}{4} & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{3}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} & \frac{3}{4} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} * \frac{1}{4} = 0$$

Using Lagrangian

$$A_r = A'(AA')^{-1} = [1111]' \frac{1}{4}$$

$$\lambda = A_r \nabla f(x^*) = 1/4$$

*Clearly*

$$\nabla f(x^*) = A' \lambda$$

# You try it

$$\min f(x) = \frac{1}{2} x' C x$$

$$s.t. \quad Ax = b$$

For

$$C = \begin{bmatrix} 0 & -13 & -6 & -3 \\ -13 & 23 & -9 & 3 \\ -6 & -9 & -12 & 1 \\ -3 & 3 & 1 & 3 \end{bmatrix} \quad A = \begin{bmatrix} 2 & 1 & 2 & 1 \\ 1 & 1 & 3 & -1 \end{bmatrix} \quad b = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

Find projection matrix

Confirm optimality conds are  $Z' C x^* = 0$ ,

$$Ax^* = b$$

Find  $x^*$

Compute Lagrangian multipliers

Check Lagrangian form of the multipliers.



# Variable Reduction Method

- Let  $A = [B \ N]$   $A$  is  $m$  by  $n$   $B$  is  $m$  by  $m$   
assume  $m < n$

$$Z = \begin{bmatrix} -B^{-1}N \\ I \end{bmatrix}$$

is a basis matrix for null space of  $A$

$$A_r = \begin{bmatrix} B^{-1} \\ 0 \end{bmatrix} \Rightarrow AA_r = [B \ N] \begin{bmatrix} B^{-1} \\ 0 \end{bmatrix} = I + 0$$



# Try on our example

- Take for example first two columns for B

$$A = \begin{bmatrix} 2 & 1 & 2 & 1 \\ 1 & 1 & 3 & -1 \end{bmatrix} = [B \quad N] = \left[ \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 2 & 1 \\ 3 & -1 \end{bmatrix} \right]$$

- Then

$$Z = \begin{bmatrix} 1 & -2 \\ -4 & 3 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \quad A_r = \begin{bmatrix} 1 & -1 \\ -1 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$

- Condition number of  $Z' CZ = 158$   
better but not great

# QR Factorization

- Use Gram-Schmidt algorithm to make orthogonal factorize  $A' = QR$  with  $Q$  orthogonal and  $R$  upper triangular

$$A' = QR = [Q_1 \quad Q_2] \begin{bmatrix} R_1 \\ 0 \end{bmatrix}$$

where  $A \in m \times n$ ,  $Q_1 \in n \times m$ ,  $Q_2 \in n \times (n - m)$ ,  $R_1 \in m \times m$

$$Z = Q_2 \quad A_r = Q_1 R_1^{-T}$$



# QR on problem

- Use matlab command QR

$$[Q \ R] = \text{qr}(A')$$

$$Q2 = Q(:,3:4)$$

$$\text{Cond}(Q2'*C*Q2) = 9.79$$




# In Class Practice

- Find optimal solution and verify FONC and SOSC of the following:
- Let the perimeter of a rectangle be fixed to 4. Find the shape of the rectangle with largest area.
- Solve the problem

$$\max x_1x_2+x_2x_3+x_1x_3$$

$$\text{s.t. } x_1+x_2+x_3=3$$
