## OM-S20-12: Ax = b to Constrained Optimization

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# Lagrangian and Constrained Optimization

Max 
$$f(x)$$
 subject to: $g(x) = 0$ 

Lagrangian function:

$$L(x,\lambda) = f(x) - \lambda g(x)$$

Maximize 
$$f(x, y)$$
 Subject to:  $g(x, y) = c$   
 $L(x, y, \lambda) = f(x, y) - \lambda(g(x, y) - c)$ 

Eg.

Maximimize 
$$x^2y$$
 Subject to: $x^2 + y^2 = 1$ 

$$\frac{\partial L}{\partial x}=0$$
  $\frac{\partial L}{\partial y}=0$  and  $\frac{\partial L}{\partial \lambda}=0$  (Optimally conditions) leads to

$$2xy = 2x$$
;  $x^2 = 2\lambda y$  and  $x^2 + y^2 = 1$ 

Ans: 
$$(\pm \sqrt{\frac{2}{3}}, \pm \sqrt{\frac{1}{3}})$$

Maximimize 
$$xy$$
 Subject to: $x + y = 10$ 

Ans: 25

## Example

Find the points on the circle  $x^2 + y^2 = 80$  that is closes and farthest from (1,2)

maximize 
$$f(x, y) = (x - 1)^2 + (y - 2)^2$$

subject to:

$$x^2 + y^2 = 80$$

$$\frac{\partial L}{\partial x}=0$$
  $\frac{\partial L}{\partial y}=0$  and  $\frac{\partial L}{\partial \lambda}=0$ 

$$\frac{(x-1)}{x} = \frac{(y-1)}{y} = \lambda$$

or 
$$y = 2x$$

$$x^2 + 4x^2 = 5x^2 = 80$$

 $x = \pm 4$  and corresponding points are (4,8) and (-4,-8)

### Two Examples

#### 1. Least Norm

Minimize  $x^T x$  subject to Ax = b

$$L(x,\lambda) = x^T x + \lambda^T [Ax - b]$$

$$2x + A^T \lambda = 0 \text{ and } Ax - b = 0$$

$$x = \frac{-A^T \lambda}{2} \text{ and } \frac{A(-A^T \lambda)}{2} - b = 0$$

$$\lambda = -2(AA^T)^{-1}b \text{ and } x = A^T (AA^T)^{-1}b$$

### 2. Solution as Eigen Vector

Minimize 
$$||Ax||^2$$
 Subject to: $||x||^2 = 1$   
 $x^T A^T A x - \lambda x^T x$ 

or

$$A^T A x = \lambda x$$

## 3. Homogeneous Equations: Ax = 0

$$Ax = 0$$

Trivial solution has the smallest norm and of no interest.

Minimize
$$||Ax||^2$$
 Subject to: $||x||^2 = 1$ 

x is an eigen vactor of  $A^TA$ . Which one?

$$x^T A^T A x = \lambda^2$$

Eigen vector corresponding to the which eigen value?

Solution using SVD: Which column of U/V?

### Comments

#### **SVD**

- Pseudo inverse:  $D_{ii}^{-1} = \frac{1}{D_{ii}}$  if  $> \theta$  else zero.
- Rank of a matrix is equal to the number of nonzero singular values
- Low rank approximation using SVD.

#### **Problem of Interest**

- **1** Solve Ax = b when A is square and full rank.
- 2 Least square solution for Ax = b
- **3** Least norm solution for Ax = b
- **1** Non-trivial solution to Ax = 0