

① ~~Minimize~~ Find point x satisfying $Ax = b$ & closest to origin. A is $m \times n$ with $\text{rank}(A) = m$.

Using Lagrange multipliers

we want to minimize $\|x\|^2$ given $Ax = b$

$$\text{So } \nabla f = 2\vec{x} = A^T \lambda \quad \text{so } x = \frac{1}{2} A^T \lambda$$

$$Ax = b$$

$$\& A \frac{1}{2} A^T \lambda = b \quad \& \Rightarrow AA^T \lambda = 2b$$

Now since $\text{rank}(A) = m = \text{rank}(A^T)$ then $\ker(AA^T) = \ker(A^T) = \{0\}$

$$\text{So } AA^T \text{ is invertible } \& \lambda = (AA^T)^{-1} 2b$$

$$\& x = \frac{1}{2} A^T (AA^T)^{-1} 2b = \boxed{A^T (AA^T)^{-1} b}$$

Also note $\nabla^2 f = I_d$ so $Z^T \nabla^2 f Z = Z^T Z$ which is positive definite

② $\bar{x} \in \mathbb{R}^n$ Z is $n \times r$ matrix $f: \mathbb{R}^n \rightarrow \mathbb{R}$

$$\text{Let } \phi: \mathbb{R}^r \rightarrow \mathbb{R} \quad \phi(v) = f(\bar{x} + Zv)$$

find $\nabla \phi$ & $\nabla^2 \phi$

$$\text{Using chain Rule } \nabla \phi = \nabla [f(\bar{x} + Zv)] = \nabla_x f \quad (\nabla f)(\bar{x} + Zv)$$

$$\nabla(\bar{x} + Zv)$$

where ~~$\nabla_x f(\bar{x} + Zv) = \nabla f$~~

$$\text{if } x = \bar{x} + Zv \quad \nabla x = (\nabla x_1, \dots, \nabla x_n)$$

$$\text{Now } x_i = (\bar{x})_i + Z_i v \quad \text{so } \nabla x_i = Z_i^T \& \nabla x = Z^T$$

$$\text{So } \nabla \bar{x} + Zv = Z^T \nabla f(\bar{x} + Zv)$$

$$\text{So } \boxed{\nabla \phi = Z^T (\nabla f)(\bar{x} + Zv)}$$

$$\nabla^2 \phi = \nabla^2 [f(\bar{x} + Zv)] = \nabla^2(\bar{x} + Zv) (\nabla f(\bar{x} + Zv) + \nabla_x f) (\nabla^2 f)(\bar{x} + Zv) \nabla x(t)^T$$

where ~~$\nabla^2(\bar{x} + Zv) = \nabla^2 \bar{x}$~~

$$\nabla^2(\bar{x} + Zv) (\nabla f)(\bar{x} + Zv) = \nabla^2(x) \nabla f(x) = \sum_{i=1}^n \nabla^2 x_i \cdot (\nabla f(x))_i$$

$$\text{but } \nabla^2 x_i = (0) \text{ since } \frac{\partial \bar{x} + Zv}{\partial v_i} = 0 \& \nabla x(t) = Z^T \text{ so } \boxed{\nabla^2 \phi = Z^T \nabla^2 f(\bar{x} + Zv) Z}$$

(3) $x \in \mathbb{R}^n$
 Z $n \times r$ basis for Nullspace of $m \times n$ Matrix A

Show $Z^T x = 0 \iff$ existence of $\lambda \in \mathbb{R}^m$ s.t. $x = A^T \lambda$

\Leftarrow : if $x = A^T \lambda$ then $Z^T x = Z^T A^T \lambda = (AZ)^T \lambda$

now $AZ = 0$ since columns of Z are a basis for $\ker A$

$$\Rightarrow Z^T x = 0$$

$\Rightarrow Z^T x = 0$ note rows of Z^T are in $\ker A$

so $x \cdot \ker A = 0 \Rightarrow x \in (\ker A)^\perp$

so we need to show $(\ker A)^\perp \subseteq \text{Im}(A^T)$

In fact we'll show $(\ker A)^\perp = \text{Im}(A^T)$

this is the same as $\ker A = \text{Im}(A^T)^\perp$

$$\text{Im}(A^T)^\perp = \{x : x \cdot A^T \lambda = 0 \text{ for } \forall \lambda\}$$

$$= \{x : Ax \cdot \lambda = 0 \text{ for } \forall \lambda\}$$

$$= \{x : Ax = 0\} = \ker A$$

for this step take $\lambda = Ax$

so their equal $\&$ $(\ker A)^\perp = \text{Im}(A^T) \Rightarrow x \in \text{Im}(A^T)$

so $x = A^T \lambda$ for some λ .

(4) Maximize $x_1 \dots x_n$

Subject to $\frac{x_1}{a_1} + \dots + \frac{x_n}{a_n} = 1$ $a_i > 0$

Setup Lagrange Multiplier $\lambda = (\lambda_1, \dots, \lambda_n)$

$$\frac{\partial f}{\partial x_i} = \frac{x_1 \dots x_n}{x_i} \quad \text{For } x_i \neq 0$$

~~Assume~~ $x_i = 0$ $x_1 \dots x_n = 0$, won't be a minimum

Assuming $x_i \neq 0$ then $x_i \frac{\partial f}{\partial x_i} = x_1 \dots x_n = f$

$$\nabla f = A^T \lambda = \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{pmatrix} \quad \text{so} \quad \begin{pmatrix} f \\ \vdots \\ f \end{pmatrix} = \begin{pmatrix} \frac{x_1 \lambda_1}{a_1} \\ \vdots \\ \frac{x_n \lambda_n}{a_n} \end{pmatrix}$$

so $\frac{x_i \lambda_i}{a_i}$ is constant for all i & since $\sum_{i=1}^n \frac{x_i}{a_i} = 1$

$$\Rightarrow \frac{x_i}{a_i} = \frac{1}{n} \quad \text{so} \quad x_i = \frac{a_i}{n}$$

lets compute $\nabla^2 f = \frac{\partial^2 f}{\partial x_i \partial x_j} = \begin{cases} 0 & i=j \\ \frac{x_1 \dots x_n}{x_i x_j} & i \neq j \end{cases}$

$$\text{so } \nabla^2 f \left(\frac{a_1}{n}, \dots, \frac{a_n}{n} \right) = \begin{pmatrix} 0 & \dots & \frac{a_1 \dots a_n}{n^n a_1 a_n} \\ \frac{a_1 \dots a_n}{n^n a_1 a_2} & \dots & 0 \\ \vdots & \dots & \vdots \\ \frac{a_1 \dots a_n}{n^n a_1 a_n} & \dots & 0 \end{pmatrix} = \frac{a_1 \dots a_n}{n^n} \begin{pmatrix} 0 & \dots & \frac{1}{a_1 a_2 \dots a_n} \\ \frac{1}{a_1 a_2} & \dots & 0 \\ \vdots & \dots & \vdots \\ \frac{1}{a_1 a_n} & \dots & 0 \end{pmatrix}$$

also we can check $Z = \begin{pmatrix} a_1 & \dots & a_n & 0 \\ -a_2 & & & 0 \\ & -a_3 & & 0 \\ 0 & & & -a_n \end{pmatrix}$

$$\begin{aligned} Z^T \nabla^2 f Z &= \frac{a_1 \dots a_n}{n^n} \begin{pmatrix} a_1 & -a_2 & & 0 \\ & & & 0 \\ & & & 0 \\ a_n & 0 & & -a_n \end{pmatrix} \begin{pmatrix} 0 & \frac{1}{a_1 a_2} & \dots & \frac{1}{a_1 a_n} \\ \frac{1}{a_1 a_2} & \dots & & 0 \\ \vdots & \dots & \dots & \vdots \\ \frac{1}{a_1 a_n} & \dots & & 0 \end{pmatrix} \begin{pmatrix} a_1 & \dots & a_n \\ -a_2 & & 0 \\ & 0 & 0 \\ & & 0 & -a_n \end{pmatrix} \\ &= \begin{pmatrix} a_1 & -a_2 & & 0 \\ & & & 0 \\ & & & 0 \\ a_n & 0 & & -a_n \end{pmatrix} \begin{pmatrix} -\frac{1}{a_1} & \dots & -\frac{1}{a_n} \\ \frac{1}{a_2} & & 0 \\ & \dots & 0 \\ 0 & & \frac{1}{a_n} \end{pmatrix} = \begin{pmatrix} -2 & & -1 \\ & \dots & \\ -1 & & -2 \end{pmatrix} \end{aligned}$$

So we just need to show $M = \begin{pmatrix} -2 & & -1 \\ & \dots & \\ -1 & & -2 \end{pmatrix}$ is negative definite

Now nullity $(A - (-1)I) =$ nullity $\begin{pmatrix} -1 & \dots & -1 \\ \vdots & & \vdots \\ -1 & \dots & -1 \end{pmatrix} = n-1$ since rank $\begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ -1 & \dots & -1 \end{pmatrix} = 1$

~~From~~ So -1 is an eigenvalue multiplicity $n-1$
we have one more eigenvalue λ multiplicity 1

$$\& (-1)(n-1) + \lambda = \text{tr } M = -2n \Rightarrow \lambda = -n-1$$

So we have eigenvalues -1 & $-n-1 < 0$ so M

is ~~is~~ negative definite $\Rightarrow x_i = \frac{a_i}{n}$ is a local Maximum \square