

S.3. KKT & CONSTRAINED GEOMETRIC PROGRAMMING

THEOREM 24 (2.4.1) A-L INEQUALITY

LET $x_1, \dots, x_m \in \mathbb{R}^+$, $\delta_1, \dots, \delta_m \in (0, 1)$

ST $\sum_{i=1}^m \delta_i = 1$ THEN

$$\text{IF } \sum_{i=1}^m x_i \delta_i \leq \sum_{i=1}^m \delta_i x_i$$

WITH EQUALITY IFF $x_1 = x_2 = \dots = x_m$

THEOREM 53 (S.31, EXTENDED A-L INEQ).

LET $x_1, \dots, x_m \in \mathbb{R}^+$. LET $\delta_1, \dots, \delta_n$ BE

ALL POSITIVE OR ALL ZERO.

IF $\lambda = \delta_1 + \dots + \delta_m$ THEN

$$\left(\sum_{i=1}^m x_i \right)^\lambda \geq \lambda \left(\prod_{i=1}^m \left(\frac{x_i}{\delta_i} \right)^{\delta_i} \right)$$

WHERE $\delta^0 = 1$ AND $(x_i/0)^0 = 1$

WITH EQUALITY IFF $\delta_1 = \delta_2 = \dots = \delta_m = 0$

OR IF: $x_i = \frac{\delta_i}{\lambda} \left(\sum_{j=1}^m x_j \right)$

PROOF - USE (AG)

• $\forall_i \delta_i > 0$

$$\frac{\delta_1}{\lambda} + \frac{\delta_2}{\lambda} + \frac{\delta_3}{\lambda} + \dots + \frac{\delta_n}{\lambda} = 1$$

$$\sum_{i=1}^m x_i = \sum_{i=1}^m \left(\frac{\delta_i}{\lambda}\right) \left(\frac{\lambda x_i}{\delta_i}\right) \stackrel{(A-G)}{\geq} \prod_{i=1}^m \left(\frac{\lambda x_i}{\delta_i}\right)^{\frac{\delta_i}{\lambda}}$$

(A-G) WITH EQUALITY IFF

$$\frac{\lambda x_1}{\delta_1} = \frac{\lambda x_2}{\delta_2} = \dots = \frac{\lambda x_m}{\delta_m} = M$$

$$\left(\sum_{i=1}^m x_i\right) \stackrel{M}{=} \prod_{i=1}^m \left(\frac{\lambda x_i}{\delta_i}\right)^{\delta_i} = M \prod_{i=1}^m \left(\frac{x_i}{\delta_i}\right)^{\delta_i}$$

$$\sum x_i = \sum \frac{M \delta_i}{\lambda} = \frac{M}{\lambda} \sum \delta_i = M$$

$$\frac{\lambda x_i}{\delta_i} = x_i \Rightarrow x_i = \frac{\delta_i}{\lambda} (\sum x_i)$$

• $\forall_i \delta_i = 0$ TRIVIAL $i = 1$

D

DEF - GEOMETRIC PROGRAM

LET $g_0(t), g_1(t), \dots, g_k(t)$ BE POLYNOMIALS
IN m REAL VARIABLES $t = (t_1, \dots, t_m)$. THEN
THE PROGRAM

MINIMIZE $g_0(t)$

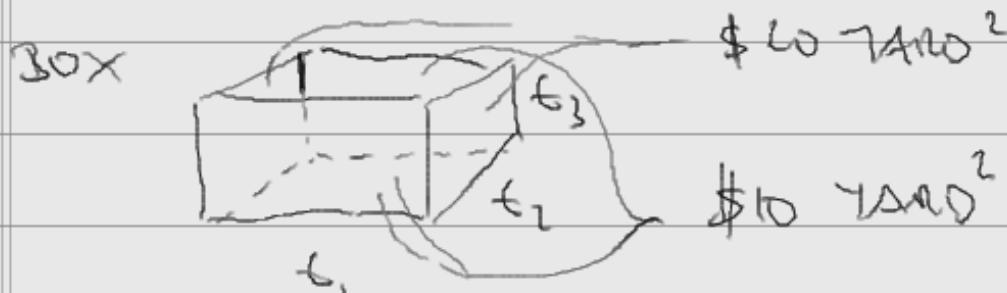
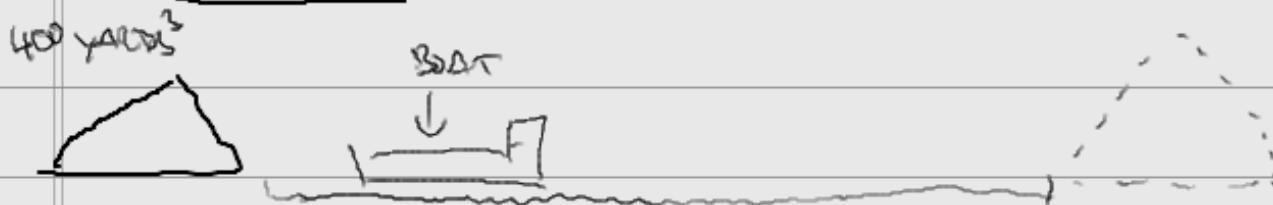
S.T. $g_1(t) \leq 1, \dots, g_k(t) \leq 1$

WHERE $t_1 > 0, t_2 > 0, \dots, t_m > 0$

IS CALLED CONSTRAINED GEOMETRIC
PROGRAM.

LE DEF POLYNOMIAL $\sum c_i t_i^{d_i}$

EXAMPLE



1 LOAD \$0.1

MINIMIZES COST

NO VALUE AFTER 2

GRAVEL TRANSPORTED

MINIMIZE $0.1 \frac{400}{e_1 t_1 t_3} + 2.20 e_1 t_3 + 2.10 e_1 t_3 + 10 e_1 t_2$

WHERE $e_i > 0, t_i > 0, t_j > 0$

& NO CONSTRAINTS.

MINIMIZE USING (AG)

$$8\left(\frac{1}{5} \frac{20}{e_1 t_1 t_3} + \frac{1}{5} \frac{20}{e_1 t_3} + \frac{1}{5} 400 e_1 t_3 + \frac{1}{5} 2.20 e_1 t_3 + \frac{1}{5} 10 e_1 t_2\right) \geq$$

$$5 \left(20.20. 40, 20.20 \right)^{\frac{1}{5}} = 5.20 = 100$$

$$e_1 t_3 = \frac{1}{2}$$

$$t_1 = 2$$

$$e_1 t_3 = 1$$

$$t_3 = 1/2$$

$$e_1 t_2 = 2$$

$$t_2 = 1$$

$$e_1 t_1 t_2 = 1$$

$$\Rightarrow \text{cost} = 40 + 80 + 2.20 + 20 < 100$$

LET'S CHANGE THE SETTING

WE HAVE 4 YARDS² OF THE SCRAP

MAYBE ONE FOR FREE BUT NOT MORE

$$\text{M.W. } \frac{40}{\epsilon_1 t_1 t_3} + 40 t_2 t_3$$

$$\text{s.t. } 2t_1 t_3 + t_2 t_3 \leq 4$$

where $\epsilon_i > 0$ $t_1 > 0$ $t_3 > 0$ $(t = t_1 + 2t_3)$

$$g_0(t) = \frac{40}{\epsilon_1 t_1 t_3} + 40 t_2 t_3$$

$$g_1(t) = \frac{t_1 t_3}{2} + \frac{\epsilon_1 t_1}{4}$$

$$\text{If } \lambda > 0, [g_1(t)]^\lambda \leq 1 \quad (\text{by AM-GM})$$

$$g_0(t) \geq g_0(\bar{t}) \quad [g_1(t)]^\lambda = \left(\frac{40}{\epsilon_1 t_1 t_3} + 40 t_2 t_3 \right) (g_1(t))^\lambda \\ = \left(\delta_1 \left(\frac{40}{\delta_1 t_1 t_3} \right) + \delta_2 \left(\frac{40 t_2 t_3}{\delta_2} \right) \right) (g_1(t))^\lambda \geq$$

$$\delta_1, \delta_2 > 0, \delta_1 + \delta_2 = 1$$

$$\geq \left(\left(\frac{40}{\delta_1 t_1 t_3} \right)^{\delta_1} \left(\frac{40 t_2 t_3}{\delta_2} \right)^{\delta_2} \right) (g_1(t))^\lambda$$

$$= \left(\left(\frac{40}{\delta_1} \right)^{\delta_1} \cdot \left(\frac{40}{\delta_2} \right)^{\delta_2} \left(\bar{t}_1 \right)^{-\delta_1} \left(\bar{t}_3 \right)^{-\delta_1 + \delta_2} \left(\bar{t}_2 \right)^{\delta_1 + \delta_2} \right) (g_1(t))^\lambda$$

$$\begin{aligned}
 (g_1(t))^{\lambda} &= \left(\frac{\epsilon_1 t_3}{2} + \frac{\epsilon_1 t_1}{4} \right)^{\lambda} \\
 &\geq \lambda \left(\left(\frac{\epsilon_1 t_3}{2\delta_3} \right)^{\delta_3} \left(\frac{\epsilon_1 t_1}{4\delta_4} \right)^{\delta_4} \right) \\
 &= \lambda \left(\left(\frac{1}{2\delta_3} \right)^{\delta_3} \left(\frac{1}{4\delta_4} \right)^{\delta_4} \right) t_1^{\delta_3} t_2^{\delta_4} t_3^{\delta_3}
 \end{aligned}$$

where $\lambda = (\delta_3 + \delta_4)$

$$g_0(t) \geq \left(\left(\frac{40}{\delta_1} \right)^{\delta_1} \left(\frac{40}{\delta_2} \right)^{\delta_2} \left(\frac{1}{2\delta_3} \right)^{\delta_3} \left(\frac{1}{4\delta_4} \right)^{\delta_4} \right) \cdot (\delta_3 + \delta_4)^{\delta_3 + \delta_4}$$

$\cdot t_1^{\delta_1} t_2^{\delta_2} t_3^{\delta_3}$

CANCELLING $\epsilon_1, \epsilon_2, t_3$:

$$\delta_1 + \delta_2 = 1$$

$$\delta_3 + \delta_4 = \lambda$$

$$-\delta_1 + \delta_3 + \delta_4 = 0$$

$$-\delta_1 + \delta_2 + \delta_4 = 0$$

$$-\delta_1 + \delta_2 + \delta_3 = 0$$

$$\delta_1 > 0, \delta_2 > 0, \delta_3 \geq 0, \delta_4 \geq 0$$

SOLUTION IS

$$\delta_1 = \frac{2}{3} \quad \delta_2 = \frac{1}{3} \quad \delta_3 = \frac{1}{3} \quad \delta_4 = \frac{1}{3}$$

$$\Rightarrow g_0(t) = 60$$

WE WANT EQUALITIES IN BOTH (A-H) MEQ.

$$\left| \frac{t_1^+ t_3^-}{\delta_3} = \frac{t_1^+ t_2^+}{\delta_3} = K \right.$$

$$1 = \frac{t_1^+ t_3^-}{2} + \frac{t_1^+ t_2^+}{4} = \delta_3 K + \delta_4 K = \frac{3}{3} K$$

$$\Rightarrow K = \frac{3}{2}$$

$$\left| \begin{array}{l} t_1^+ t_3^- = 1 \\ t_1^+ t_2^+ = 2 \end{array} \right| \quad (E(A-H))$$

(A-H):

$$\frac{40}{\frac{2}{3} t_1^+ t_3^-} = \frac{40 t_1^+ t_2^+}{\frac{1}{3}} = 2 = 60$$

$$\frac{2}{3} Z + \frac{1}{3} Z = 60 \Rightarrow Z = 60$$

$$\Rightarrow \epsilon_1^r \epsilon_L^t \epsilon_3^t = 1$$

$$2\epsilon_2^t \epsilon_3^t = 1$$

$$\left| \begin{array}{l} \epsilon_1^r \epsilon_3^t = 1 \\ \underline{\epsilon_1^r \epsilon_2^t = 2} \end{array} \right| \leftarrow \text{FROM Q1D}$$

log A_{II}

$$\log \epsilon_1^r + \log \epsilon_1^t + \log \epsilon_3^t = 0$$

$$\log \epsilon_1^r + \log \epsilon_3^t = -\log 2$$

$$\log \epsilon_1^r + \log \epsilon_3^t = 0$$

$$\log \epsilon_1^r + \log \epsilon_2^t = \log 2$$

$$\log \epsilon_1^r = \log 2 = \epsilon_1^r = 2$$

$$\log \epsilon_3^t = 0 \Rightarrow \epsilon_3^t = 1$$

$$\log \epsilon_2^t = \log 2^{-1} \Rightarrow \epsilon_2^t = -\frac{1}{2}$$

DOMS

\rightarrow Steuerkurve A-L in NST und D

FORMALIZE PREVIOUS EXAMPLES

$$M_j(t) = C_j t_1^{d_{j1}} t_2^{d_{j2}} t_3^{d_{j3}} \cdots t_m^{d_{jm}} \quad (C_j > 0)$$

CONSTRAINED (CP)

$$\left. \begin{array}{l} \text{MINIMIZE } g_0(t) = M_1(t) + \dots + M_{m_0}(t) \\ \text{SUBJECT TO} \end{array} \right\}$$

$$(CP) \quad g_1(t) = M_{m_0+1}(t) + \dots + M_{m_1}(t) \leq 1$$

;

$$g_K(t) = M_{m_{K-1}+1}(t) + \dots + M_{m_K}(t) \leq 1$$

WHERE $t \geq 0$

$$\text{SO } m_K = n$$

AK:

$$\begin{aligned} g_\sigma(t) &\geq g_0(t) \cdot g_1(t)^{\gamma_1} g_2(t)^{\gamma_2} \cdots g_K(t)^{\gamma_K} \geq \\ &\geq \left(\prod_{i=1}^K \frac{c_i}{\delta_i} \right)^{\delta_i} \prod_{i=1}^K \lambda_i^{\gamma_i} (t^{\alpha_1} t^{\alpha_2} \cdots t^{\alpha_n}) \end{aligned}$$

$$\left. \begin{array}{l} \text{MAXIMIZE } \Lambda(\sigma) = \left(\prod_{i=1}^K \frac{c_i}{\delta_i} \right)^{\delta_i} \prod_{i=1}^K \lambda_i^{\gamma_i} \\ \text{ST. } \delta_1 + \dots + \delta_{m_0} = 1 \quad (\text{just AK}) \end{array} \right\}$$

DP

$$\alpha_1 \delta_1 + \alpha_2 \delta_2 + \dots + \alpha_n \delta_n = 0$$

;

$$\alpha_{1m} \delta_1 + \dots + \alpha_{nm} \delta_n = 0$$

} WHERE $s_i > 0$ FOR $1 \leq i \leq m_1$
 } IF $s_i > 0$ FOR $m_{k-1} + 1 \leq i \leq m_k$ OR
 } $s_i = 0$ FOR $m_{k-1} \leq i \leq m_k$
 $\lambda_i = s_{m_{k-1}+1} + \dots + s_{m_k}$

DUAL OBJECTIVE FUNCTION, DUAL CONSTRAINTS,
 DLP IS LINEAR .. WHICH = 1 SOLUTION

EXAMPLE:
 MAX $\epsilon_1^2 \epsilon_2 \epsilon_3 \dots s_1$

$$\begin{aligned}
 \text{ST. } & -2\epsilon_1 \epsilon_2 + \frac{1}{2} \epsilon_3 \leq 1 && \rightarrow \delta_1 \\
 \delta_2 & -\frac{1}{4} \epsilon_1^2 + \frac{1}{2} \epsilon_2 \epsilon_3 \leq 1 && \rightarrow \delta_2 \\
 \delta_3 & \text{ WHERE } \epsilon_i \geq 0 && \dots \delta_5
 \end{aligned}$$

$$\begin{aligned}
 \text{MAX } & \left(\frac{1}{\delta_1}\right)^{\delta_1} \left(\frac{2}{\delta_2}\right)^{\delta_2} \left(\frac{1}{2\delta_3}\right)^{\delta_3} \left(\frac{1}{4\delta_4}\right)^{\delta_4} \left(\frac{1}{2\delta_5}\right)^{\delta_5} \\
 & \lambda_1 \dots (\delta_2 + \delta_3)^{\delta_2 + \delta_3} \cdot (\delta_4 + \delta_5)^{\delta_4 + \delta_5} \dots \lambda_2
 \end{aligned}$$

SUBJECT TO

$$\begin{array}{lll}
 \text{obs. } & \delta_1 & = 1 \\
 t_1 & 2\delta_1 + \delta_2 & -2\delta_4 = 0 \\
 t_2 & \delta_1 - 2\delta_2 & + 2\delta_5 = 0 \\
 t_3 & -\delta_1 - \delta_3 & + 2\delta_5 = 0
 \end{array}$$

$$\begin{aligned}
 \delta_1 > 0, \quad \delta_2, \delta_4 > 0 \text{ or } \delta_2 = \delta_4 = 0 \\
 \delta_4, \delta_5 > 0 \text{ or } \delta_4 = \delta_5 = 0
 \end{aligned}$$

$$\delta_1 = 1$$

$$\delta_2 = \alpha \quad \alpha \geq 0$$

$$\delta_4 = 1 - \frac{\alpha}{2} \quad \alpha \leq 2$$

$$\delta_5 = -\frac{1}{2} + \alpha \quad \alpha \geq \frac{1}{2}$$

$$\delta_3 = -2 + 2\alpha \quad \alpha \geq 1$$

USING KKT FOR LP : (KKT for convex)

$$\min f(x)$$

$$\text{s.t. } g_1(x) \leq 0, g_2(x) \leq 0, \dots, g_n(x) \leq 0$$

f, g convex, x convex

Polynomial is not convex: $c \sim t^{\frac{1}{2}}$

$\exists t > 0 \Rightarrow \exists x : e^x = t$

$$t_j = e^{x_j}$$

$$g(t) = \sum_{i=1}^m c_i t_i^{\alpha_i} t_2^{\alpha_2} \dots t_m^{\alpha_m}$$

$$h(x) = \sum_{i=1}^m c_i e^{\alpha_i x_i}$$

$h(x)$ is convex

$$(GP) \begin{cases} \min g_0(t) \\ \text{s.t. } g_1(t) \leq 1, g_2(t) \leq 1, \dots, g_N(t) \leq 1 \\ \text{where } t > 0 \end{cases}$$

ASSOCIATED CONVEX PROGRAM IS

$$t_i \Rightarrow e^{x_i}$$

$$(CP) \begin{cases} \min h_0(x) \\ h_1(x) - 1 \leq 0, \dots, h_N(x) - 1 \leq 0 \\ x \in \mathbb{R}^m \end{cases}$$

e STRICTLY INCREASING \Rightarrow

$t^* = (t_1^*, t_2^*, \dots, t_m^*)$ is solution of (GP)

IFF $x^* = (x_1^*, \dots, x_m^*)$ is solution of (CP*)