

1.1:  $f(x) = \|x\|_2^2$ ,  $h(x) = \sum_{i=1}^n x_i - 1$ .

Want to find coupled equations for Lagrangian to be stationary w.r.t  $x$  and  $\lambda$ .

$$L(x, \lambda) = \|x\|_2^2 + \lambda(\sum_{i=1}^n x_i - 1)$$

$$\nabla_L = \frac{\partial L}{\partial x} = 2\|x\|_2 + \lambda(\sum_{i=1}^n x_i - 1)$$

$$= 2\|x\|_2 + \lambda n$$

$$= 2\|x\|_2 + \lambda n = 0$$

$$\lambda = -\frac{2\|x\|_2}{n}$$

$$\frac{\partial L}{\partial \lambda} = \|x\|_2^2 + \lambda \sum_{i=1}^n x_i$$

$$= \sum_{i=1}^n x_i$$

$$= x_n$$

$$x_n = 0$$

1.2:  $f(x) = \sum_{i=1}^n x_i$ ,  $h(x) = \|x\|_2^2 - 1$ .

$$L(x, \lambda) = \sum_{i=1}^n x_i + \lambda(\|x\|_2^2 - 1)$$

$$\nabla_L = \frac{\partial L}{\partial x} = n + 2\|x\|_2 \lambda = 0$$

$$\lambda = -\frac{n}{2\|x\|_2}$$

$$\frac{\partial L}{\partial \lambda} = \|x\|_2^2$$

$$x_2 = 0$$

1.3:  $f(x) = \|x\|_2^2$ ,  $h(x) = x^T Q x - 1$ .

$$L(x, \lambda) = \|x\|_2^2 + \lambda(x^T Q x - 1)$$

$$\nabla_L = \frac{\partial L}{\partial x} = 2\|x\|_2 + \lambda(x^T Q + x^T Q) = 0$$

$$\lambda = -\frac{2\|x\|_2}{(x^T Q + x^T Q)}$$

$$\frac{\partial L}{\partial \lambda} = x^T Q x = 0$$

$$x = 0$$

2:  $L(x_1, \dots, x_n, \lambda) = f(x_1, \dots, x_n) - \lambda(g(x_1, \dots, x_n) - 1)$

$$\frac{\partial L}{\partial x_i} = a_i x_i^{(a_i-1)} + \lambda \prod_{j \neq i} x_j^{a_j} = 0$$

$$\frac{\partial L}{\partial \lambda} = \prod_{i=1}^n x_i^{(a_i)-1} = 0$$

After using the constraint, we find that  $\lambda = -\sum_{i=1}^n \frac{a_i}{x_i}$

We then substitute this back into the first equation and in both sides to get:

$$\sum_{i=1}^n a_i \ln(x_i) \leq \ln\left(\sum_{i=1}^n a_i x_i\right)$$

This proves the inequality due to natural log properties.

3:  $\min_{x,y} f(x,y) = \frac{1}{2}((x-x_0)^2 + (y-y_0)^2)$  with  $(x_0, y_0)$  fixed and subject to  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$ .

Lagrangian:  $L(x,y,\lambda) = f(x,y) + \lambda(g(x,y) - c)$

$$L(x,y,\lambda) = \frac{1}{2}((x-x_0)^2 + (y-y_0)^2) + \lambda\left(\frac{x^2}{a^2} + \frac{y^2}{b^2} - 1\right)$$

Solving for  $x, y$ :  $\frac{\partial L}{\partial x} = (x-x_0) + \frac{2\lambda x}{a^2} = 0$

$$\frac{a^2 x + 2\lambda x}{a^2} = x_0$$

$$\frac{x(a^2 + 2\lambda)}{a^2} = x_0$$

$$x = \frac{x_0 a^2}{(a^2 + 2\lambda)}$$

$$\frac{\partial L}{\partial y} = (y-y_0) + \frac{2\lambda y}{b^2} = 0$$

$$y = \frac{y_0 b^2}{(b^2 + 2\lambda)}$$

Rewriting Ellipse constraint in terms of Lagrange parameter:  $\frac{\partial L}{\partial \lambda} = \frac{x^2}{a^2} + \frac{y^2}{b^2} - 1 = 0$

$$1 = \frac{\left(\frac{x_0 a^2}{(a^2 + 2\lambda)}\right)^2}{a^2} + \frac{\left(\frac{y_0 b^2}{(b^2 + 2\lambda)}\right)^2}{b^2} \quad \text{---}$$

$$1 = \frac{a^2 + 2\lambda}{a^2(x_0 a^2)} + \frac{b^2 + 2\lambda}{b^2(y_0 b^2)}$$

$$g(\lambda) = \frac{a^2 + 2\lambda}{a^4 x_0} + \frac{b^2 + 2\lambda}{b^4 y_0} - 1 = 0$$

Show it leads to quartic, can general quartic be solved with radicals?

$$(a^2 + 2\lambda)b^4 y_0 + (b^2 + 2\lambda)a^4 x_0 - a^4 x_0 b^4 y_0 = 0$$

Expand and simplify:  $\lambda^4(2a^2 b^2 x_0 y_0) + \lambda^3(a^4 y_0^2 + b^4 x_0^2 - a^4 x_0 b^4) + \lambda^2(a^2 b^4 y_0 + b^2 a^4 x_0 - a^4 x_0 b^4) + 2\lambda(a^4 x_0 b^4) = 0$

A general quartic can't be solved by radicals, but in some cases when the quartic is symmetric, it can. (Normally you can find an approx. solution via special methods)

4:  $E(w) = \|y - \sum_{k=1}^K w_k x_k\|_2^2$  where  $w \in \mathbb{R}^K$  and  $x_i, y \in \mathbb{R}^D$

Make Lagrangian with constraint  $\sum_{k=1}^K w_k = 1$ :  $L(w, \lambda) = \|y - \sum_{k=1}^K w_k x_k\|_2^2 + \lambda (\sum_{k=1}^K w_k - 1)$

Solve for Lagrange parameter. Explain the steps: Take partial derivatives to  $w_k$  and set equal to 0:

$$\frac{\partial L}{\partial w_i} = -2(x_i^T y) + 2(x_i^T w [x_1, x_2, \dots, x_K]) + \lambda = 0$$

represent as  $X$

Note:  $E(w) = (y - \sum_{k=1}^K w_k x_k)^T (y - \sum_{k=1}^K w_k x_k)$

↓ Transpose

Rewrite to matrix form as:  $-2X^T y + 2X^T X w + \lambda \mathbf{1}_K = 0$

Solve for  $\lambda$ :  $\lambda = \frac{2(X^T y - X^T X w)}{K}$

We are dividing element by element due to the constraint that the weights sum to 1.

Solve for  $w$  while satisfying constraint: Substitute  $\lambda$  into Lagrangian:

$$L(w, \lambda) = \|y - Xw\|_2^2 + \lambda (1 - \mathbf{1}_K^T w)$$

Take the gradient of  $L(w, \lambda)$  w.r.t  $w$ :

$$\nabla L(w, \lambda) = -2X^T y + 2X^T X w - \lambda \mathbf{1}_K = 0$$

Note:  $\mathbf{1}_K^T w = 1$

We use the condition and multiply both sides by  $X^T$ , then invert  $X^T X$ .

$$w = (X^T X)^{-1} X^T (y - \frac{\lambda}{K} \mathbf{1}_K)$$

(Assuming  $X^T X$  is invertible)

5. <https://colab.research.google.com/drive/1yUwtI75cuRYvn-xzslW-cJ9x6Njm5y4x?usp=sharing>