

椭圆不确定集下的鲁棒对等问题推导

我们要证明在椭圆不确定集下，鲁棒对等问题可转化为如下的二阶锥规划问题：

1.

$$\min_{\tilde{v}_{i,s} \in U_v^e} \left\{ \sum_{i \in I} \sum_{s \in S_i} \tilde{v}_{i,s} y_{i,s} \right\} = \sum_{i \in I} \sum_{s \in S_i} \bar{v}_{i,s} y_{i,s} - \Omega_v \sqrt{\sum_{i \in I} \sum_{s \in S_i} y_{i,s}^2 (\sigma_{i,s}^v)^2}$$
$$\text{where } \tilde{v}_{i,s} \in U_v^e, U_v^e = \{ \tilde{v} : (\tilde{v} - \bar{v})^T \Sigma_v^{-1} (\tilde{v} - \bar{v}) \leq \Omega_v^2 \}$$

2.

$$\max_{\tilde{c}_{i,s} \in U_c^e} \left\{ \sum_{i \in I} \sum_{s \in S_i} \tilde{c}_{i,s} y_{i,s} \right\} = \sum_{i \in I} \sum_{s \in S_i} \bar{c}_{i,s} y_{i,s} + \Omega_c \sqrt{\sum_{i \in I} \sum_{s \in S_i} y_{i,s}^2 (\sigma_{i,s}^c)^2}$$
$$\text{where } \tilde{c}_{i,s} \in U_c^e, U_c^e = \{ \tilde{c} : (\tilde{c} - \bar{c})^T \Sigma_c^{-1} (\tilde{c} - \bar{c}) \leq \Omega_c^2 \}$$

Proof 1:

已知：

- 向量/矩阵求偏导：

$$\frac{\partial Ax}{\partial x} = A^T \quad \frac{\partial Ax}{\partial x^T} = A$$

- KKT条件：

$$\begin{aligned} \min \quad & f(\mathbf{x}) \\ \text{s.t.} \quad & g_j(\mathbf{x}) = 0, \quad j = 1, \dots, m, \\ & h_k(\mathbf{x}) \leq 0, \quad k = 1, \dots, p. \end{aligned}$$

定义Lagrangian 函数

$$L(\mathbf{x}, \{\lambda_j\}, \{\mu_k\}) = f(\mathbf{x}) + \sum_{j=1}^m \lambda_j g_j(\mathbf{x}) + \sum_{k=1}^p \mu_k h_k(\mathbf{x})$$

其中 λ_j 是对应 $g_j(\mathbf{x}) = 0$ 的Lagrange乘数, μ_k 是对应 $h_k(\mathbf{x}) \leq 0$ 的Lagrange乘数(或称KKT乘数)。KKT条件包括

$$\begin{aligned} \nabla_{\mathbf{x}} L &= \mathbf{0} \\ g_j(\mathbf{x}) &= 0, \quad j = 1, \dots, m, \\ h_k(\mathbf{x}) &\leq 0, \\ \mu_k &\geq 0, \\ \mu_k h_k(\mathbf{x}) &= 0, \quad k = 1, \dots, p. \end{aligned}$$

问题1的矩阵形式:

$$\begin{aligned} \min_{\tilde{v}} \quad & y_{1 \times P}^T \cdot \tilde{v}_{P \times 1} \\ \text{s.t.} \quad & (\tilde{v} - \bar{v})_{1 \times P}^T \cdot \Sigma_{P \times P}^{-1} \cdot (\tilde{v} - \bar{v})_{P \times 1} \leq \Omega^2 \end{aligned}$$

由KKT条件:

$$f(\tilde{v}, \mu) = y^T \cdot \tilde{v} + \mu \cdot [(\tilde{v} - \bar{v})^T \cdot \Sigma^{-1} \cdot (\tilde{v} - \bar{v}) - \Omega^2]$$

$$\frac{\partial f(\tilde{v}, \mu)}{\partial \tilde{v}} = y + 2\mu \Sigma^{-1}(\tilde{v} - \bar{v}) = 0 \quad (1)$$

$$\mu \cdot [(\tilde{v} - \bar{v})^T \cdot \Sigma^{-1} \cdot (\tilde{v} - \bar{v}) - \Omega^2] = 0 \quad (2)$$

由 (1) 得:

$$\tilde{v} = \bar{v} - \frac{\Sigma}{2\mu} \cdot y \quad (3)$$

带入 (2) :

$$\left(-\frac{y^T}{2\mu}\right) \cdot \Sigma^T \cdot \Sigma^{-1} \cdot \left(-\frac{\Sigma}{2\mu}\right) \cdot y = \frac{y^T \Sigma y}{4\mu^2} = \Omega^2 \quad (4)$$

由 (4) 得:

$$2\mu = \frac{\sqrt{y^T \Sigma y}}{\Omega} \quad (5)$$

把 (5) 带入 (3) :

$$\tilde{v} = \bar{v} - \frac{\Omega}{\sqrt{y^T \Sigma y}} \cdot \Sigma \cdot y \quad (*)$$

所以有：

$$\begin{aligned} \min_{\tilde{v}_{i,s} \in U_v^e} \left\{ \sum_{i \in I} \sum_{s \in S_i} \tilde{v}_{i,s} y_{i,s} \right\} &= y^T \cdot \tilde{v} = y^T \cdot \bar{v} - \Omega \cdot \frac{y^T \Sigma y}{\sqrt{y^T \Sigma y}} \\ &= \sum_{i \in I} \sum_{s \in S_i} \bar{v}_{i,s} y_{i,s} - \Omega_v \sqrt{\sum_{i \in I} \sum_{s \in S_i} y_{i,s}^2 (\sigma_{i,s}^v)^2} \end{aligned}$$

得证。

Proof 2:

和证明1基本一致，细微差异如下。

问题2的矩阵形式：

$$\begin{aligned} \min_{\tilde{c}} \quad & y_{1 \times P}^T \cdot \tilde{c}_{P \times 1} \\ \text{s.t.} \quad & (\tilde{c} - \bar{c})_{1 \times P}^T \cdot \Sigma_{P \times P}^{-1} \cdot (\tilde{c} - \bar{c})_{P \times 1} \leq \Omega^2 \end{aligned}$$

由KKT条件：

$$f(\tilde{c}, \mu) = y^T \cdot \tilde{c} + \mu \cdot [\Omega^2 - (\tilde{c} - \bar{c})^T \cdot \Sigma^{-1} \cdot (\tilde{c} - \bar{c})]$$

$$\frac{\partial f(\tilde{c}, \mu)}{\partial \tilde{c}} = y - 2\mu \Sigma^{-1} (\tilde{c} - \bar{c}) = 0 \quad (1)$$

$$\mu \cdot [(\tilde{c} - \bar{c})^T \cdot \Sigma^{-1} \cdot (\tilde{c} - \bar{c}) - \Omega^2] = 0 \quad (2)$$

最终求得：

$$\tilde{c} = \bar{c} + \frac{\Omega}{\sqrt{y^T \Sigma y}} \cdot \Sigma \cdot y \quad (*)$$