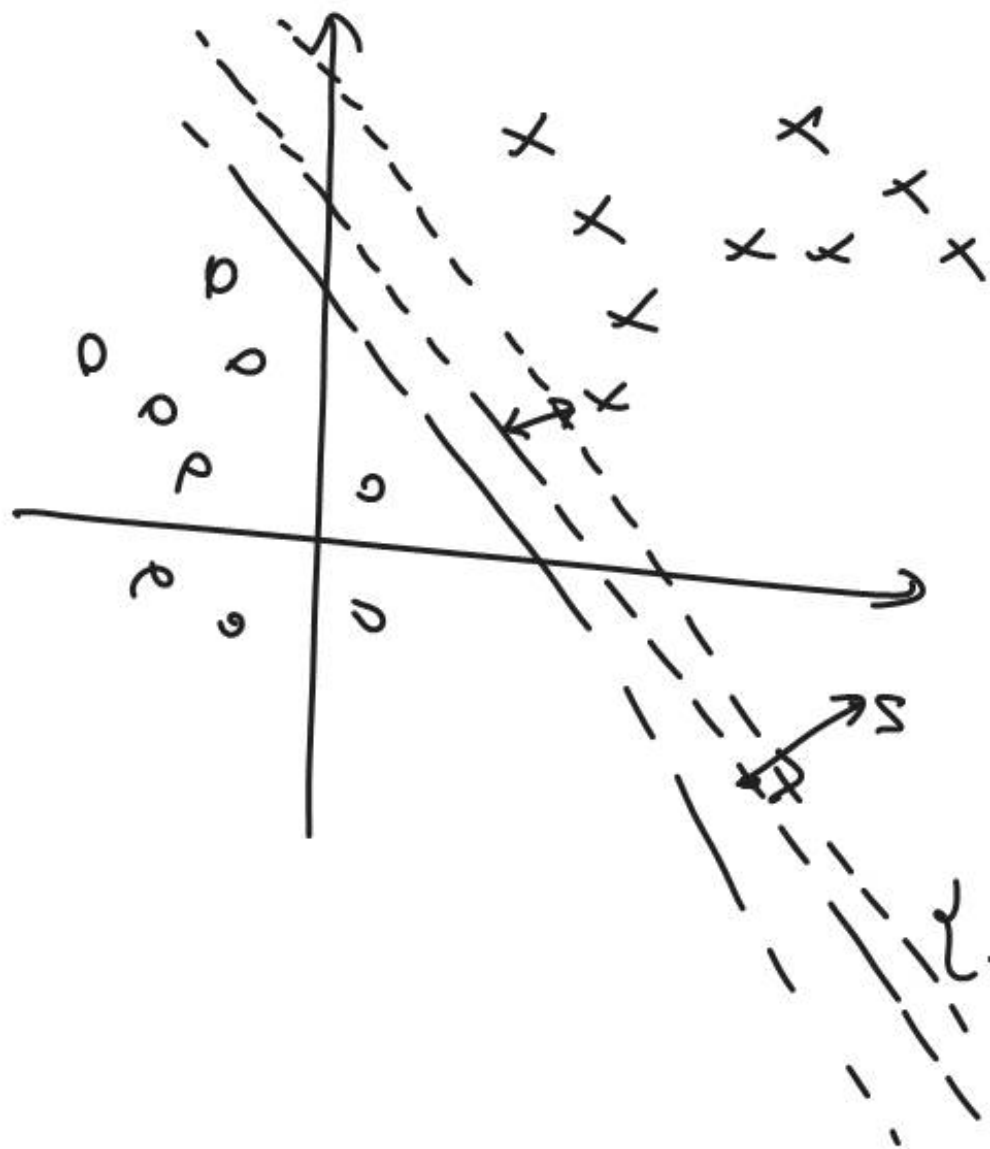




$$s^T x = r$$

$$\min. \|s\|^2$$

$$\text{s.t. } X^T s \geq (r+1) \mathbf{1}_K \leftarrow \mu \quad (\dots \leq 0)$$

$$Y^T s \leq (r-1) \mathbf{1}_L \leftarrow \nu \quad (\dots \leq 0)$$

$$1) \mathcal{L}(s, r; \mu, \nu) = \|s\|^2 + \mu^T ((r+1) \mathbf{1}_K - X^T s) + \nu^T (Y^T s - (r-1) \mathbf{1}_L)$$



$$= \|s\|^2 - s^T (X\mu - Y\nu) + r(\mu^T \mathbf{1}_K - \nu^T \mathbf{1}_L)$$

$$\nabla(\cdot)(s^*) = 0$$

$$2s^* - (X\mu - Y\nu) = 0$$

$$2) \mathcal{L}(\mu, \nu) = \inf_{(s, r)} \mathcal{L}(s, r; \mu, \nu)$$

$$\rightarrow s^* = \frac{X\mu - Y\nu}{2}$$

$$\mathcal{L}(s, r; \mu, \nu) = \begin{cases} -\infty, & \text{if } \mu^T \mathbf{1}_K - \nu^T \mathbf{1}_L \neq 0 \\ -\frac{1}{4} \|X\mu - Y\nu\|^2, & \text{if } \mu^T \mathbf{1}_K = \nu^T \mathbf{1}_L \end{cases}$$

$$2) L(\mu, \nu) = \begin{cases} -\infty & ; \quad l_k^T \mu \neq l_L^T \nu \\ -\frac{\|x_\mu - y_\nu\|^2}{2} + l_k^T \mu + l_L^T \nu & ; \end{cases}$$

$$3) \max. \quad L(\mu, \nu)$$

s.t.

$$\mu \geq 0 \\ \nu \geq 0$$

$$= \max -\frac{\|x_\mu - y_\nu\|^2}{2} + l_k^T \mu + l_L^T \nu$$

s.t.

$$\mu \geq 0 \\ \nu \geq 0 \\ l_k^T \mu = l_L^T \nu$$

$$l_k^T \mu = l_L^T \nu \quad \text{if} \quad l_k^T \mu = l_L^T \nu$$



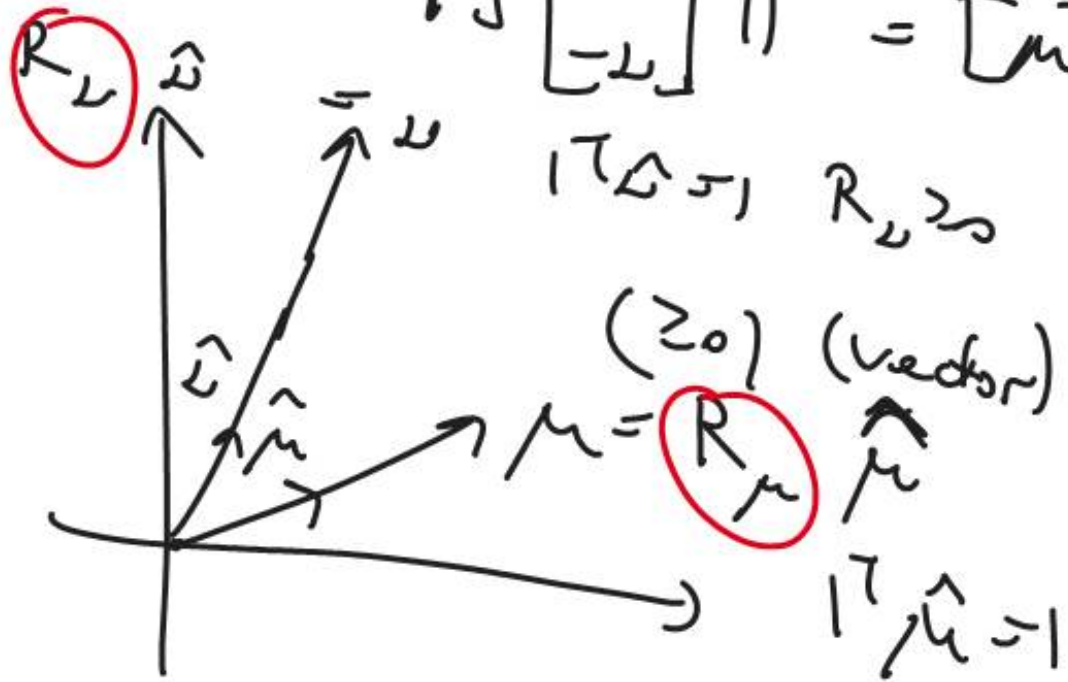
$$l_k^T \mu \neq l_L^T \nu$$

max. $-\frac{1}{4} (\|X\mu - Yv\|^2) + (\bar{k}^T \mu + \bar{l}^T v) R_v$

s.t. $\mu \geq 0, v \geq 0, \underline{l}^T \mu = \underline{l}^T v$

$R_\mu \hat{\mu}$
 $R_v \hat{v}$

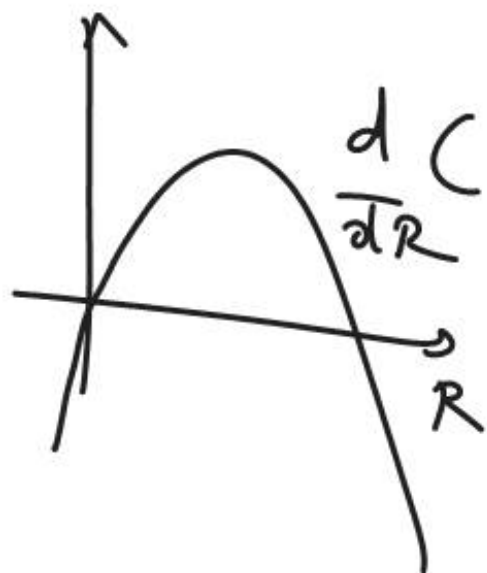
$\| [X \ Y] \begin{bmatrix} \mu \\ -v \end{bmatrix} \|^2 = \begin{bmatrix} \mu^T & -v^T \end{bmatrix} \begin{bmatrix} X^T X & X^T Y \\ Y^T X & Y^T Y \end{bmatrix} \begin{bmatrix} \mu \\ -v \end{bmatrix}$



"kernel trick"

$$\max. \quad -\frac{1}{4} R^2 \|x_{\hat{\mu}} - \gamma \hat{z}\|^2 + 2R$$

$$s.t. \quad \hat{\mu}, \hat{z} \geq 0 \quad \mathbf{1}^T \hat{\mu} = \mathbf{1}^T \hat{z} = 1 \quad R \geq 0$$



$$d(R) \big|_{R=R^*} = 0$$



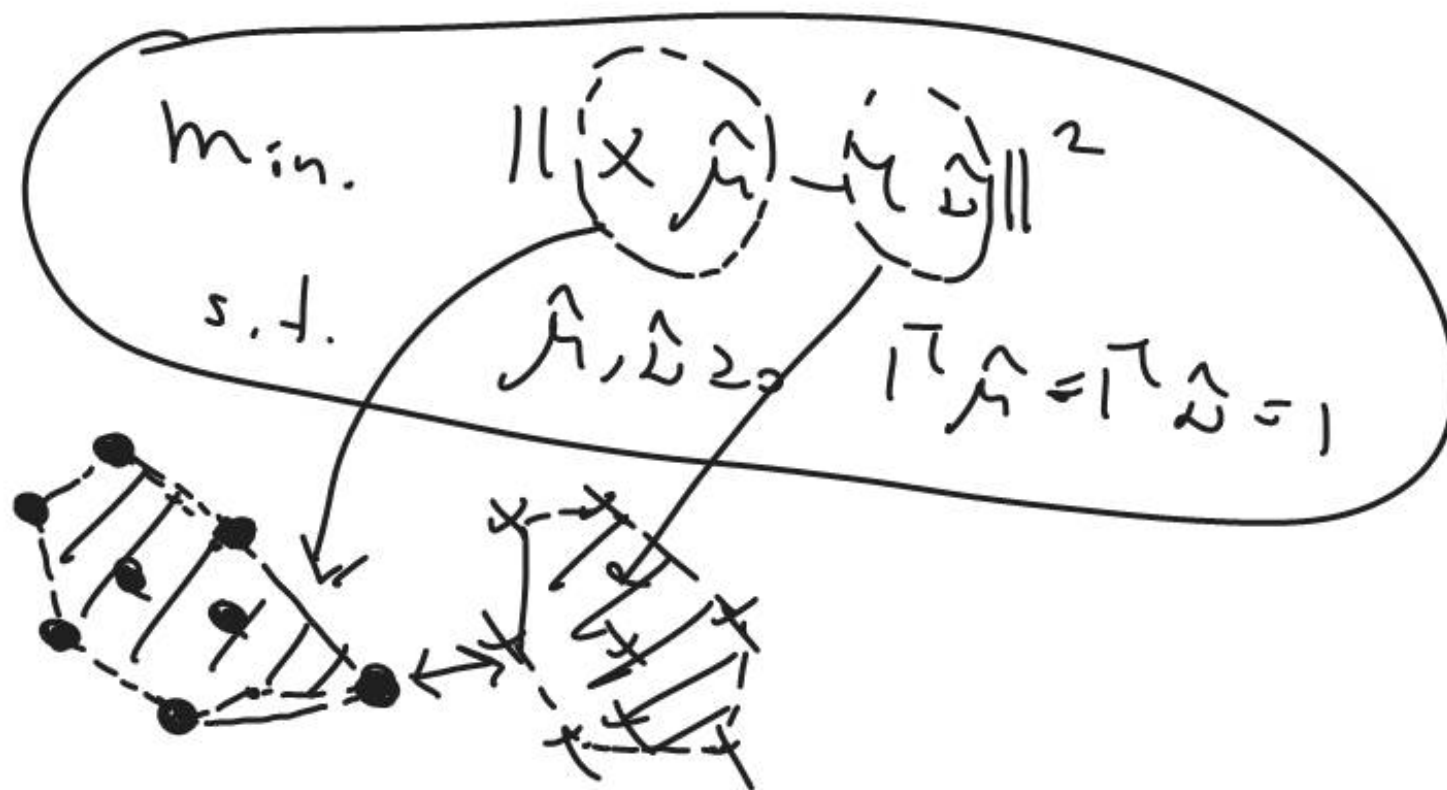
$$-\frac{1}{2} R^* \|x_{\hat{\mu}} - \gamma \hat{z}\|^2 + 2 = 0$$

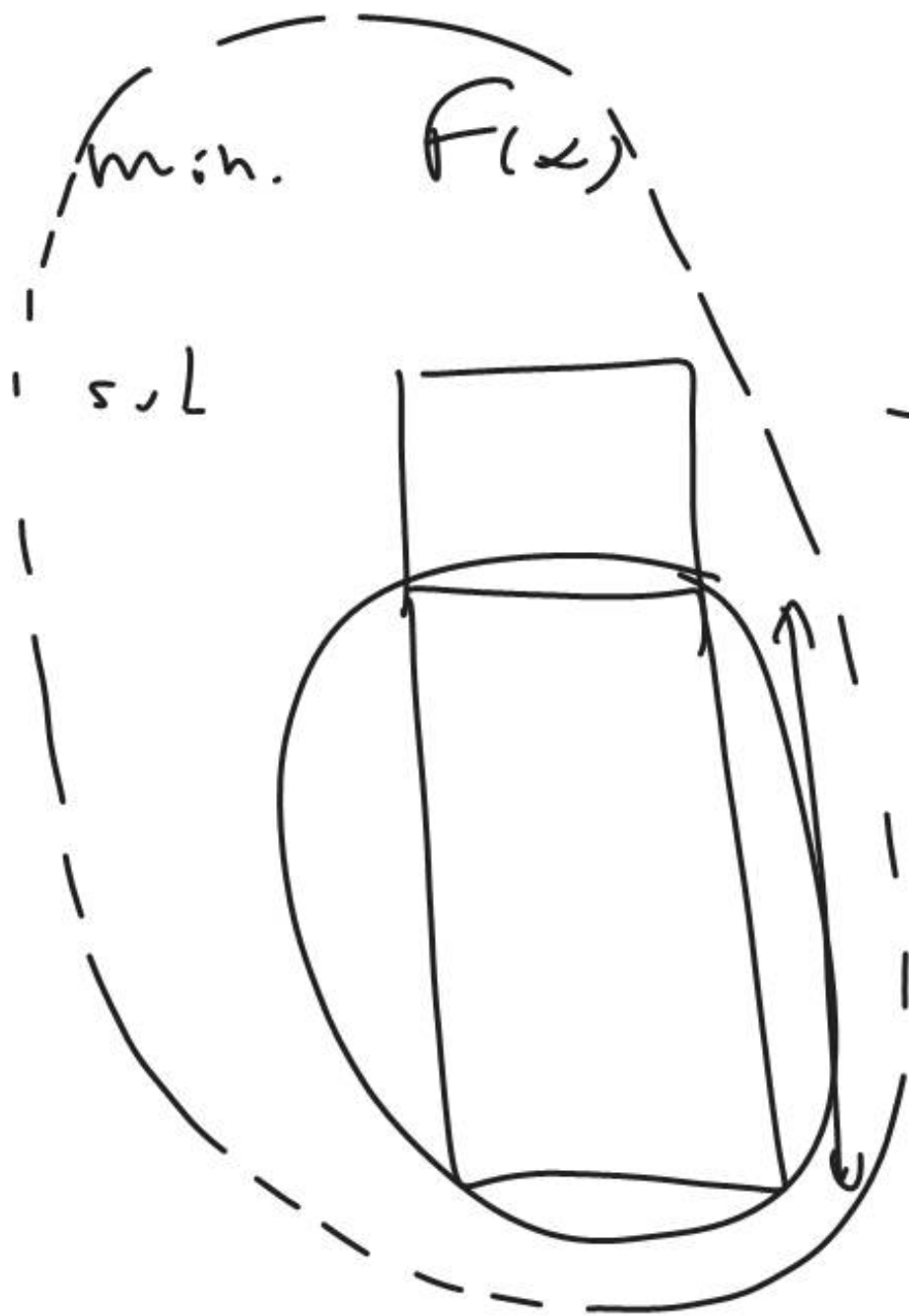
$$R^* = \frac{4}{\|x_{\hat{\mu}} - \gamma \hat{z}\|^2}$$

$$\inf_{x,y} f(x,y) = \inf_x \left[\inf_y f(x,y) \right] = \inf_x \underbrace{\left[\inf_y f(x,y) \right]}_{\phi(x)}$$

$$\max. \quad +4 \frac{1}{\|x\hat{\mu} - \tau\hat{\Delta}\|^2}$$

$$\text{s.t.} \quad \tau\hat{\mu} = \tau\hat{\Delta} = 1 \quad \hat{\mu}, \hat{\Delta} \geq 0$$





min. $g(x, y)$

s.l.



max. $r^T x$

s.t. $x \geq 0, \quad 1^T x = T$

~~$x_{E1} + x_{E2} \leq 0.8 \cdot T$~~

$x_1 + x_2 \leq 0.8T$

$x_1 + x_3 \leq 0.8T$

$x_1 + x_4 \leq 0.8T$

$x_{h-1} + x_h \leq 0.8T$

min. $c^T x$
s.t. $Ax \leq b$

$\binom{n}{k} \sim n^k$

Bender's decomposition
Dantzig-Wolfe

$$\lambda_1(x) + \dots + \lambda_k(x) \leq \alpha$$

$$\lambda(x) = \begin{bmatrix} \lambda_1(x) \\ \vdots \\ \lambda_k(x) \end{bmatrix}$$

$$K_{\lambda} F_{\lambda}$$

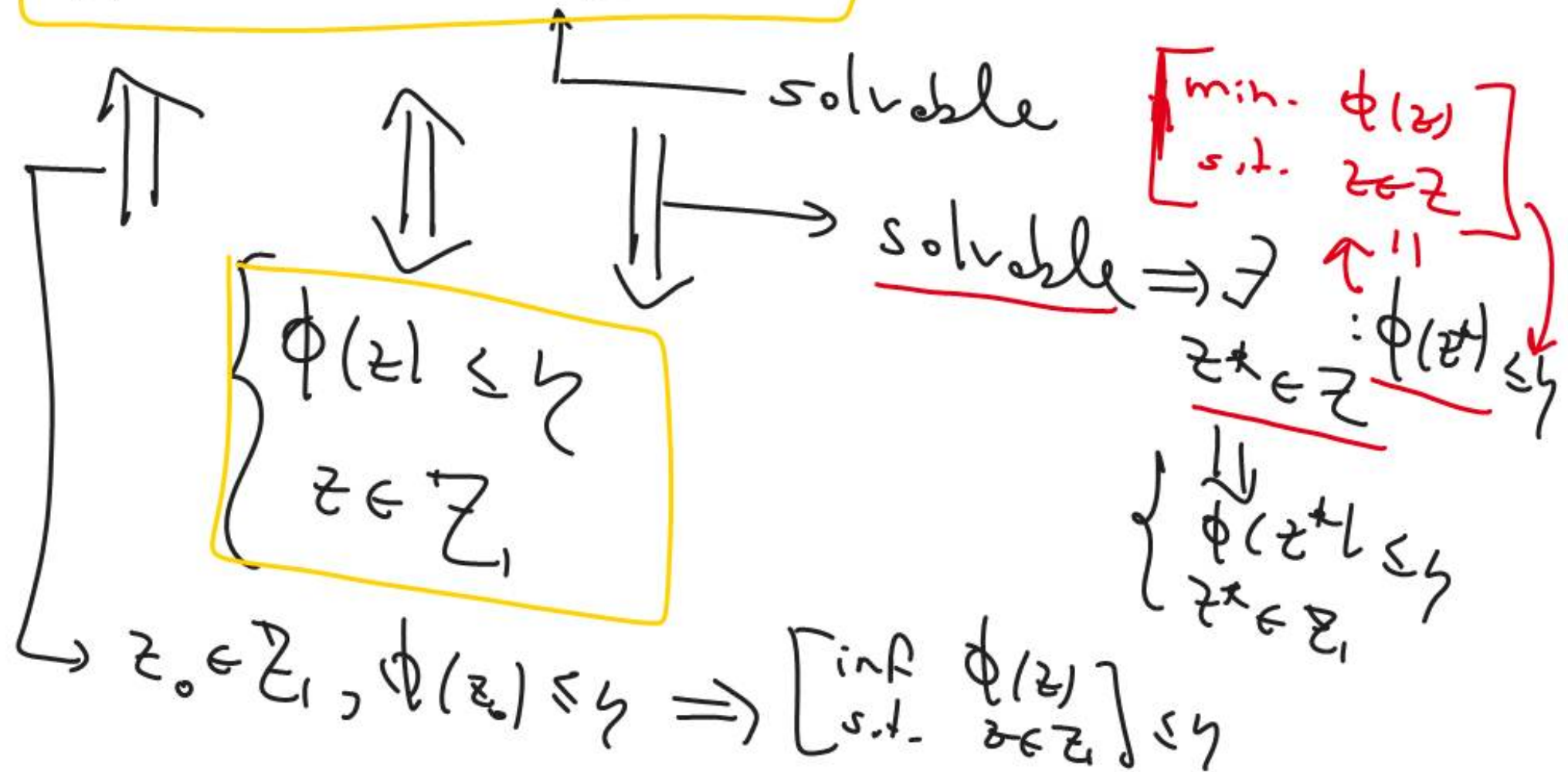
$$(x_{E_1} + \dots + x_{E_k}) \leq 0.3T$$

$$x_{E_1} + \dots + x_{E_k} = \begin{cases} \max. & F(x, y) \\ \text{s.t.} & y \in Y \end{cases}$$

↓ Duality (solvable)

$$\begin{aligned} & \begin{cases} f(x, z) \leq 0.3T \\ h(x, z) \leq 0 \end{cases} \\ & \begin{cases} \min. & g(x, z) \\ \text{s.t.} & h(x, z) \leq 0 \end{cases} \leq 0.8T \end{aligned}$$

$$\left[\begin{array}{l} \text{min. } \phi(z) \\ \text{s.t. } z \in Z \end{array} \right] \leq \gamma$$



$$\min. \quad \phi(z)$$

$$\text{s.t.} \quad z \in \mathcal{Z}$$

solvable



$$\exists z^* \in \mathcal{Z} : \phi(z^*) = \inf \{ \phi(z) : z \in \mathcal{Z} \}$$

$$= \left[\begin{array}{l} \min. \quad \phi(z) \\ \text{s.t.} \quad z \in \mathcal{Z} \end{array} \right]$$

$$\begin{bmatrix} \text{min.} & e^{-x} \\ \text{s.t.} & x \geq 0 \end{bmatrix} \leq 0$$



~~$$\begin{bmatrix} e^{-x} \\ x \geq 0 \end{bmatrix} \leq 0$$~~