

Strong Lagrange Duality (with equality constraints)

Let $f : R^n \rightarrow R$ and $g : R^n \rightarrow R^m$ are **convex** functions,

$h : R^n \rightarrow R^k$ is **affine**

X is a closed convex set over R^n

there exists $\hat{x} \in X : g(\hat{x}) < 0$ and $h(\hat{x}) = 0$.

Then:

if x^* **minimizes** $f(x)$ subject to $g(x) \leq 0, h(x) = 0, x \in X$

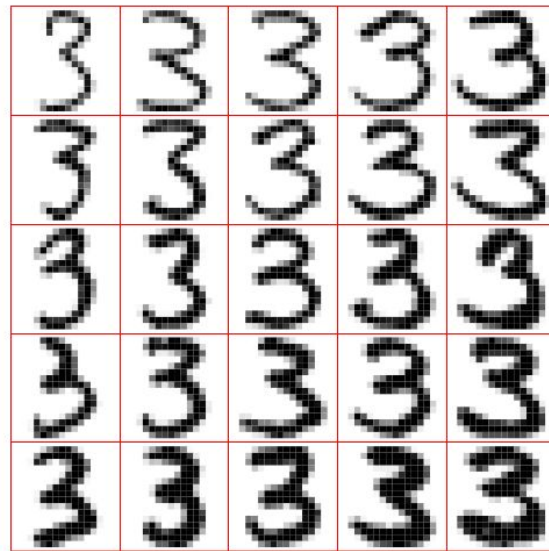
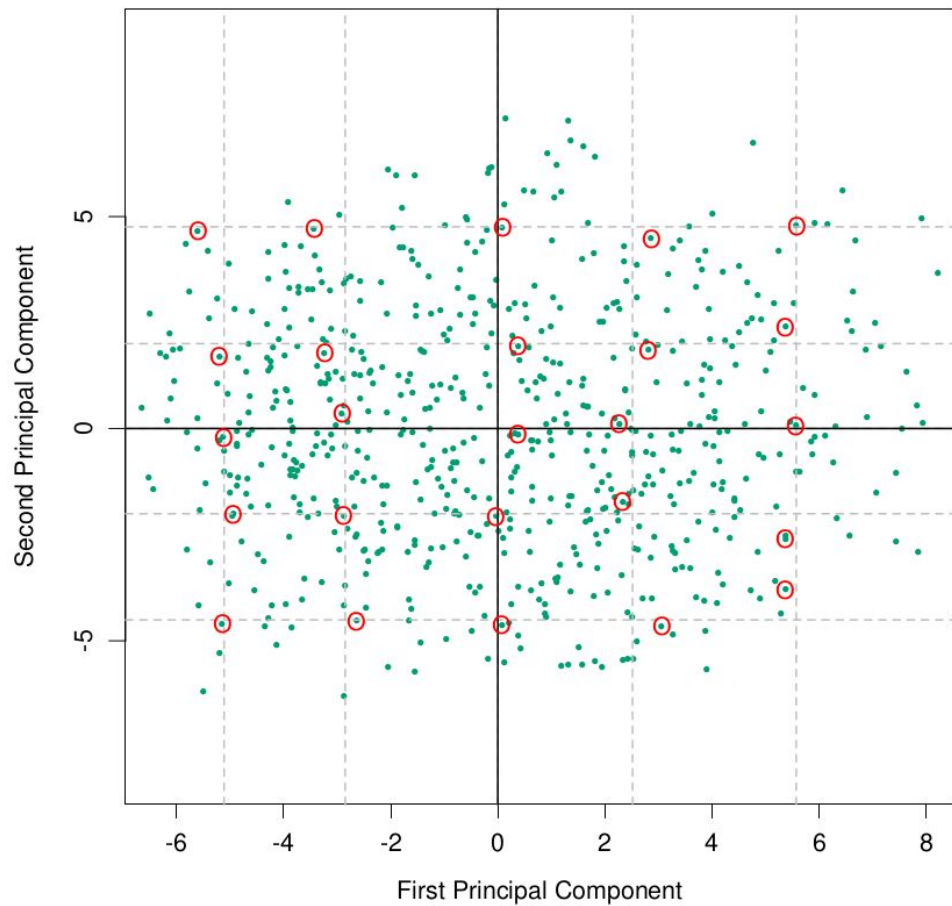
$(\lambda_g^*, \lambda_h^*)$ **maximizes** $L(\lambda_g, \lambda_h)$ subject to $\lambda_g \geq 0$ ← dual program

↖ primal program

then $f(x^*) = L(\lambda_g^*, \lambda_h^*)$

$$L(\lambda_g, \lambda_h) = \min_{x \in X} \underbrace{\{f(x) + \langle \lambda_g, g(x) \rangle + \langle \lambda_h, h(x) \rangle\}}_{\ell(\lambda_g, \lambda_h, x)}$$

vectors λ_g, λ_h : so called Lagrange multipliers



$$\begin{aligned}\hat{f}(\lambda) &= \bar{x} + \lambda_1 v_1 + \lambda_2 v_2 \\ &= \boxed{\text{3}} + \lambda_1 \cdot \boxed{\text{3}} + \lambda_2 \cdot \boxed{\text{3}}.\end{aligned}$$





