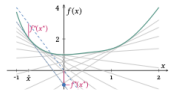
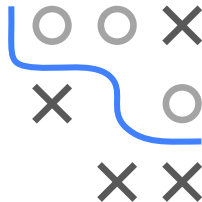


# Optimization in Machine Learning

## Other forms of duality



(a) The function  $f$  (solid) and the linear function of  $x^* = -4$  (dashed) and its shifted tangent (dotted).



(b) The Fenchel conjugate  $f^*$  of  $f$ .

### Learning goals

- Dual norms
- Conjugate functions
- Fenchel duality
- Examples in statistics

# CONSTRAINED MINIMIZATION AND DUAL NORMS

Consider the problem of norm minimization under linear constraints in its primal form:

$$\begin{array}{ll} \min_{\mathbf{x} \in \mathbb{R}^d} & \|\mathbf{x}\| \\ \text{s.t.} & \mathbf{G}\mathbf{x} = \mathbf{h}, \end{array}$$

where  $\|\cdot\|$  is some norm function. For instance, if the norm is the  $L_1$  norm, this problem is the famous [basis pursuit](#) problem.

**Question:** Is there a more straightforward way to solve constrained optimization problems involving norms?



# CONSTRAINED PROBLEMS AND CONJUGATE FUNCTIONS

