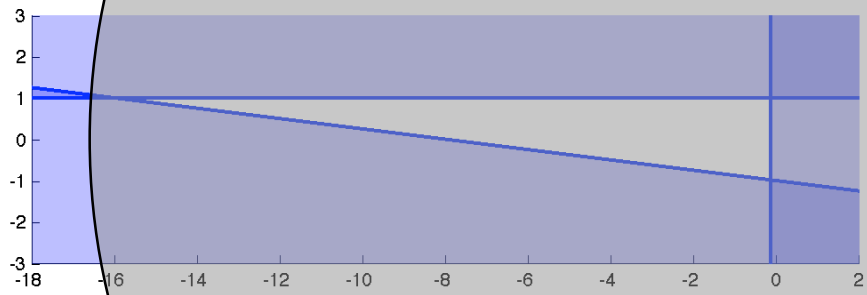
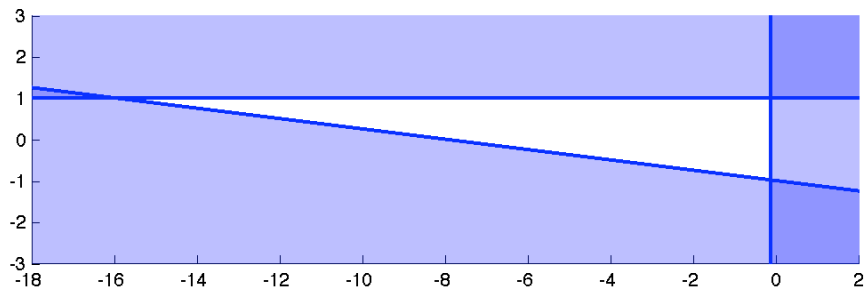


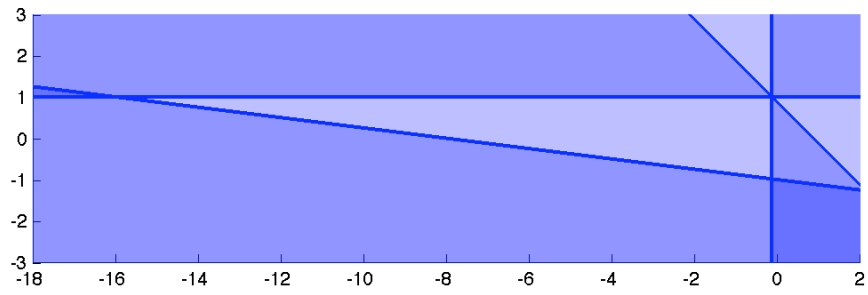
# Bit length example



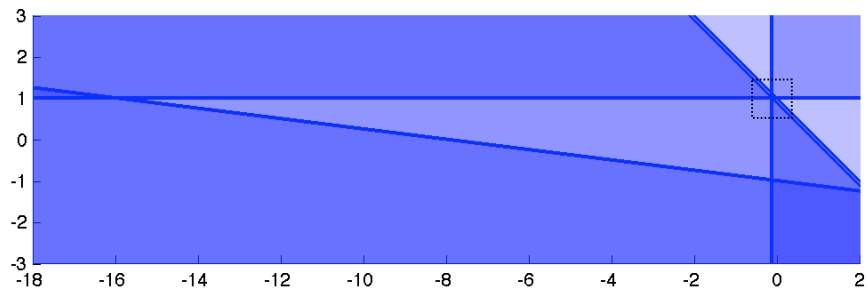
# Bit length example



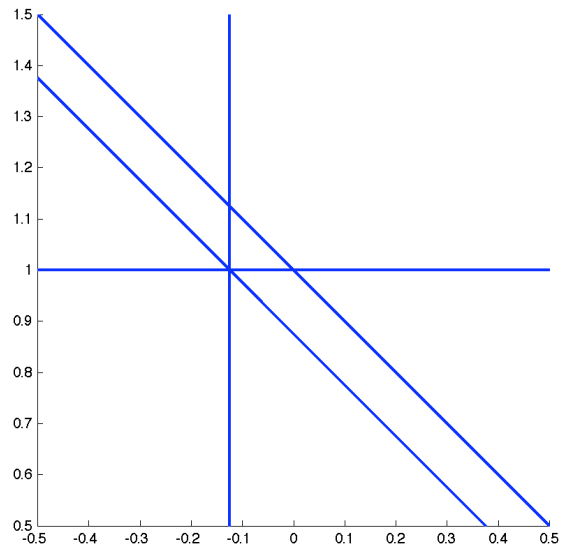
# Bit length example



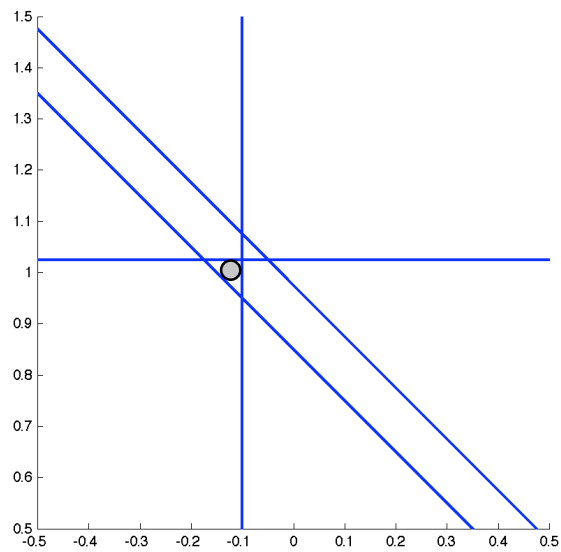
# Bit length example



# Bit length example




# Bit length example



# What's a subgradient?

## Subgradients for SVMs

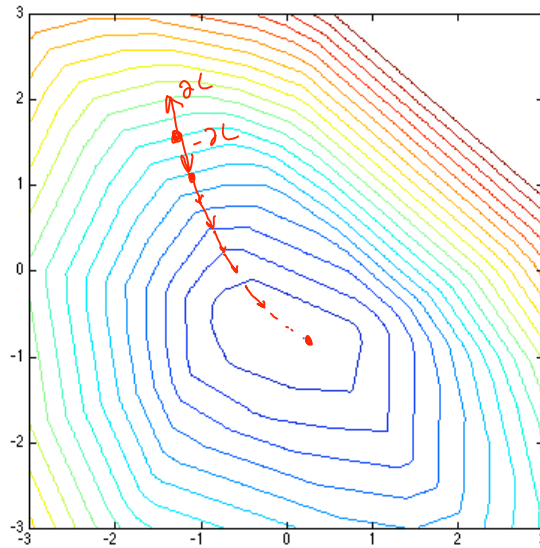
- $\min_w L(w) = \|w\|^2 + (C/m) \sum_i h(-y_i x_i^T w)$
- $h(z) = \max\{0, 1+z\}$
- Subgradient of  $h(z)$ :

$$\partial h(z) = \begin{cases} 0 & z < -1 \\ 1 & z > -1 \\ [0, 1] & z = -1 \end{cases}$$


- Subgradient of  $L(w)$  wrt  $w$ :

$$\partial L(w) = 2w + \frac{C}{m} \sum_i \partial h(-y_i x_i^T w) \cdot (-y_i x_i)$$

# Subgradient descent



# Subgradient descent

- Start w/  $x_0$
- While not tired:  $\eta_t = \text{learning rate}$   
 $g_t = (\text{estimate of}) \partial f(x_t)$   
 $x_{t+1} = x_t - \eta_t g_t$   
 $x_{t+1} := \Pi_F x_{t+1}$   
     $\uparrow$  projection onto feasible region  $F$

## Subgradient example

$$\begin{aligned} \min L(w) &= h(-z_1^T w) + h(-z_2^T w) + h(-z_3^T w) \\ \text{s.t. } \|w\|^2 &\leq 5 \end{aligned}$$

## Subgradient convergence

- Suppose  $\|\partial L(x)\|^2 \leq C$  for all  $x$  in  $F$
- Suppose  $L(x_t) \geq L(x^*) + \varepsilon$

## Setting step size

- If we knew  $\varepsilon$ , could set good step size  $\eta$
- But we don't!      So:

- Typical choices:

## Stochastic subgradient

- In SVM (and many other ML problems),  $L(w)$  contains big sum of simple terms

$$\min_w L(w) = \|w\|^2 + (C/m) \sum_i h(-y_i x_i^T w)$$

$$\partial L(w) =$$

- Approximate sum by sampling terms

$$\partial_i = \quad \quad \quad \partial_S =$$

$$E(-\partial_S^T(x-x^*)) =$$

$$S \text{ random, } |S| = k: \text{Var}(\eta \partial_S) \leq$$

## When do we stop?

- Feasible region diameter  $\|F\|$   
 $\geq f(x^*) \geq$
- Typical ML generalization bound:  
 $E(L(\text{new ex}, w)) \leq L(\text{train}, w) +$