

**Problem 1:** Write down the gradient descent update rule for the objective function

$$J(\theta) = \theta^4 - 6\theta^3 + 11\theta^2 - 7\theta + 4$$

from class. Suppose the learning rate is  $\alpha$ .

The update rule is

$$\theta := \theta - \alpha J'(\theta) = \theta - \alpha(4\theta^3 - 18\theta^2 + 22\theta - 7).$$

The recurrence relation is

$$\theta_{t+1} = \theta_t - \alpha J'(\theta_t) = \theta_t - \alpha(4\theta_t^3 - 18\theta_t^2 + 22\theta_t - 7),$$

for  $t \geq 0$ .

**Problem 2:** Consider the affine objective function

$$J(\theta) = m\theta + b,$$

for some parameters  $m \neq 0$  and  $b \in \mathbb{R}$ .

(a) Write down the gradient descent update rule. Suppose the learning rate is  $\alpha$ .

The update rule is

$$\theta := \theta - \alpha m.$$

The recurrence relation is

$$\theta_{t+1} = \theta_t - \alpha m,$$

for  $t \geq 0$ .

(b) Find a closed form expression for  $\theta_t$ .

We have

$$\theta_t = \theta_0 - \alpha m t$$

for all  $t \geq 1$ .

(c) Using your answer to (b), discuss convergence of the gradient descent algorithm.

We have  $\theta_t \rightarrow -\infty$  as  $t \rightarrow \infty$ .

**Problem 3:** Consider the quadratic objective function

$$J(\theta) = \theta^2.$$

(a) Write down the gradient descent update rule. Suppose the learning rate is  $\alpha$ .

The update rule is

$$\theta := \theta - 2\alpha\theta.$$

The recurrence relation is

$$\theta_{t+1} = \theta_t - 2\alpha\theta_t,$$

for  $t \geq 0$ .

(b) Find a closed form expression for  $\theta_t$ .

We have

$$\theta_t = (1 - 2\alpha)^t \theta_0.$$

(c) Using your answer to (b), discuss convergence of the gradient descent algorithm.

We have  $\theta_t \rightarrow 0$  exponentially fast provided  $|1 - 2\alpha| < 1$ , which occurs if and only if  $\alpha < 1$ ; the value  $\theta_t$  orbits back and forth between  $-\theta_0$  and  $+\theta_0$  if  $\alpha = 1$ ; the algorithm diverges to  $\infty$  if  $\alpha > 1$ .

**Problem 4:** Consider again the affine objective function

$$J(\theta) = m\theta + b,$$

for some parameters  $m \neq 0$  and  $b \in \mathbb{R}$ .

(a) Write down the gradient descent update rule with learning rate  $\alpha$  and decay rate  $\beta$ .

The  $t$ -th update rule is

$$\theta := \theta - \alpha m(1 - \beta)^{t+1}.$$

The recurrence relation is

$$\theta_{t+1} = \theta_t - \alpha m(1 - \beta)^{t+1},$$

for  $t \geq 0$ .

(b) Find a closed form expression for  $\theta_t$ .

Setting  $\gamma = 1 - \beta$  for convenience, we have

$$\theta_t = \theta_0 - \alpha m \sum_{k=1}^t \gamma^k,$$

for  $t \geq 1$ . But

$$\sum_{k=1}^t \gamma^k = \frac{\gamma - \gamma^{t+1}}{1 - \gamma},$$

and so

$$\theta_t = \theta_0 - \alpha m \left( \frac{\gamma - \gamma^{t+1}}{1 - \gamma} \right),$$

for  $t \geq 1$ .

(c) Using your answer to (b), discuss convergence of the gradient descent algorithm.

As we saw in Problem 2, the algorithm diverges if  $\beta = 0$ . But if  $\beta > 0$ , then  $\gamma < 1$  and

$$\lim_{t \rightarrow \infty} \theta_t = \theta_0 - \alpha m \left( \frac{\gamma}{1 - \gamma} \right) = \theta_0 - \alpha m \left( \frac{1 - \beta}{\beta} \right).$$

Thus, the algorithm converges if the decay rate  $\beta$  is positive.