

# 1 Automatic Differentiation

## 1.1 Multiple choice (5 pts)

Which of the following statements is true? In all these questions, you can assume the functions are differentiable.

True   False

- Reverse-mode automatic differentiation computes the derivative of the input of a function with regards to its output.
- To differentiate a function that has one input and 10 outputs, we should use reverse-mode automatic differentiation.
- The Jacobian of a function is a symmetric matrix.
- Reverse-mode differentiation is generally more expensive to use than forward-mode automatic differentiation because it requires recording all the operations.
- Central finite differences always gives the same result as forward-mode automatic differentiation.

**Solution:** True: D, False: A, B, C, E

## 1.2 Jacobian (5 pts)

Consider a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ . Find a way to use automatic differentiation in order to compute the Jacobian matrix  $\mathbf{J}_f(\mathbf{x})$  of that function in the least number of differentiation passes possible. Describe the steps you would take and which AD mode you would use.

**Solution:** If  $n < m$ , we should use forward-mode AD. We can compute  $n$  passes of forward-mode AD to compute the products  $\mathbf{J}_f(\mathbf{x})\mathbf{e}_i$ , where  $\mathbf{e}_i$  is the  $i$ -th canonical basis vector of  $\mathbb{R}^n$ . This will give us the rows of the Jacobian. Conversely, if  $n > m$ , we should use reverse-mode AD. We can compute  $m$  passes of reverse-mode AD to compute the products  $\mathbf{d}_i^T \mathbf{J}_f(\mathbf{x})$ , where  $\mathbf{d}_i$  is the  $i$ -th canonical basis vector of  $\mathbb{R}^m$ . This will give us the columns of the Jacobian.

### 1.3 Differentiation of a function (5 pts)

Consider the following computation:

```
a = exp(-2*x)
b = 3 * a + 1
y = sqrt(b)
```

We now want to apply automatic-differentiation to this computation. Write the sequence of steps that forward-mode automatic differentiation would take to compute the derivative of  $y$  with regards to  $x$ . Do the same for reverse-mode automatic differentiation.

**Solution:** Forward-mode AD:

```
dx = 1
da = -2*dx*exp(-2*x) = -2 * exp(-2*x)
db = 3 * da = -6 * exp(-2*x)
dy = - db / (2*sqrt(b)) = -3 * exp(-2*x) / sqrt(b)
```

Reverse-mode AD:

```
dy = 1
db = - dy / (2*sqrt(b)) = -1 / (2*sqrt(b))
da = 3 * db = -3 / (2 * sqrt(b))
dx = -2*da*exp(-2*x) = -3 * exp(-2*x) / sqrt(b)
```

### 1.4 Solving a linear system (5 pts)

We provide you with a black-box function  $linear\_solve(b)$ , that takes as input a vector  $b$  and returns the solution  $x$  of the linear system  $Ax = b$  for a constant symmetric matrix  $A$ .

(1) Is solving a linear system a differentiable operation? Why? (2 pts)

**Solution:** Solving a (square) linear system is the same operation as computing a matrix-vector product:  $x = A^{-1}b$ , which is differentiable.

(2) similar to Homework 4, write the backward pass of the function: (3pts)

**Solution:**

```
@mytorch_bwd  
def solve_system(b, out):  
    return linear_solve(A, out.grad),
```