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(<http://cocl.us/pytorch> link top)



Differentiation in PyTorch

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In this lab, you will learn the basics of differentiation.

- [Derivatives](#)
- [Partial Derivatives](#)

Estimated Time Needed: **25 min**

Preparation

The following are the libraries we are going to use for this lab.

In [1]:

```
# These are the libraries will be useing for this lab.  
  
import torch  
import matplotlib.pyplot as plt
```

Derivatives

Let us create the tensor `x` and set the parameter `requires_grad` to true because you are going to take the derivative of the tensor.

In [2]:

```
# Create a tensor x

x = torch.tensor(2.0, requires_grad = True)
print("The tensor x: ", x)
```

The tensor x: tensor(2., requires_grad=True)

Then let us create a tensor according to the equation $y=x^2$.

In [3]:

```
# Create a tensor y according to y = x^2

y = x ** 2
print("The result of y = x^2: ", y)
```

The result of $y = x^2$: tensor(4., grad_fn=<PowBackward0>)

Then let us take the derivative with respect `x` at $x = 2$

In [4]:

```
# Take the derivative. Try to print out the derivative at the value x = 2

y.backward()
print("The dervative at x = 2: ", x.grad)
```

The dervative at $x = 2$: tensor(4.)

The preceding lines perform the following operation:

$$\frac{\mathrm{d}y(x)}{\mathrm{d}x}=2x$$

$$\frac{\mathrm{d}y(x=2)}{\mathrm{d}x}=2(2)=4$$

In []:

In [5]:

```
print('data:',x.data)
print('grad_fn:',x.grad_fn)
print('grad:',x.grad)
print("is_leaf:",x.is_leaf)
print("requires_grad:",x.requires_grad)
```

```
data: tensor(2.)
grad_fn: None
grad: tensor(4.)
is_leaf: True
requires_grad: True
```

In [6]:

```
print('data:',y.data)
print('grad_fn:',y.grad_fn)
print('grad:',y.grad)
print("is_leaf:",y.is_leaf)
print("requires_grad:",y.requires_grad)
```

```
data: tensor(4.)
grad_fn: <PowBackward0 object at 0x7f3bcd239cc0>
grad: None
is_leaf: False
requires_grad: True
```

Let us try to calculate the derivative for a more complicated function.

In [7]:

```
# Calculate the y = x^2 + 2x + 1, then find the derivative

x = torch.tensor(2.0, requires_grad = True)
y = x ** 2 + 2 * x + 1
print("The result of y = x^2 + 2x + 1: ", y)
y.backward()
print("The dervative at x = 2: ", x.grad)
```

```
The result of y = x^2 + 2x + 1:  tensor(9., grad_fn=<AddBackward0>)
The dervative at x = 2:  tensor(6.)
```

The function is in the following form: $y=x^2+2x+1$

The derivative is given by:

$$\frac{\mathrm{d}y(x)}{\mathrm{d}x}=2x+2$$

$$\frac{\mathrm{d}y(x=2)}{\mathrm{d}x}=2(2)+2=6$$

Practice

Determine the derivative of $y = 2x^3 + x$ at $x=1$

In [8]:

```
# Practice: Calculate the derivative of  $y = 2x^3 + x$  at  $x = 1$ 
x = torch.tensor(1.0, requires_grad = True)
y = 2* x ** 3 + x
print("y = ", y)
y.backward()
print("The derivative of y at x = 1: ", x.grad)
# Type your code here
```

```
y = tensor(3., grad_fn=<AddBackward0>)
The derivative of y at x = 1: tensor(7.)
```

Double-click **here** for the solution.

We can implement our own custom autograd Functions by subclassing `torch.autograd.Function` and implementing the forward and backward passes which operate on Tensors

In [9]:

```
class SQ(torch.autograd.Function):

    @staticmethod
    def forward(ctx,i):
        """
        In the forward pass we receive a Tensor containing the input and return
        a Tensor containing the output. ctx is a context object that can be used
        to stash information for backward computation. You can cache arbitrary
        objects for use in the backward pass using the ctx.save_for_backward metho
d.
        """
        result=i**2
        ctx.save_for_backward(i)
        return result

    @staticmethod
    def backward(ctx, grad_output):
        """
        In the backward pass we receive a Tensor containing the gradient of the los
s
        with respect to the output, and we need to compute the gradient of the loss
        with respect to the input.
        """
        i, = ctx.saved_tensors
        grad_output = 2*i
        return grad_output
```

We can apply it the function

In [10]:

```
x=torch.tensor(2.0,requires_grad=True )
sq=SQ.apply

y=sq(x)
y
print(y.grad_fn)
y.backward()
x.grad
```

<torch.autograd.function.SQBackward object at 0x7f3c5c3f2668>

Out[10]:

tensor(4.)

Partial Derivatives

We can also calculate **Partial Derivatives**. Consider the function: $f(u,v)=vu+u^2$

Let us create `u` tensor, `v` tensor and `f` tensor

In [11]:

```
# Calculate  $f(u, v) = v * u + u^2$  at  $u = 1, v = 2$ 

u = torch.tensor(1.0, requires_grad=True)
v = torch.tensor(2.0, requires_grad=True)
f = u * v + u ** 2
print("The result of  $v * u + u^2$ : ", f)
```

The result of $v * u + u^2$: tensor(3., grad_fn=<AddBackward0>)

This is equivalent to the following:

$$f(u=1, v=2) = (2)(1) + 1^2 = 3$$

Now let us take the derivative with respect to `u` :

In [12]:

```
# Calculate the derivative with respect to u

f.backward()
print("The partial derivative with respect to u: ", u.grad)
```

The partial derivative with respect to `u`: tensor(4.)

the expression is given by:

$$\frac{\partial f(u, v)}{\partial u} = v + 2u$$

$$\frac{\partial f(u=1, v=2)}{\partial u} = 2 + 2(1) = 4$$

Now, take the derivative with respect to `v` :

In [13]:

```
# Calculate the derivative with respect to v

print("The partial derivative with respect to v: ", v.grad)
```

The partial derivative with respect to `v`: tensor(1.)

The equation is given by:

$$\frac{\partial f(u,v)}{\partial v} = u$$

$$\frac{\partial f(u=1,v=2)}{\partial v} = 1$$

Calculate the derivative with respect to a function with multiple values as follows. You use the sum trick to produce a scalar valued function and then take the gradient:

In [14]:

```
# Calculate the derivative with multiple values

x = torch.linspace(-10, 10, 10, requires_grad = True)
Y = x ** 2
y = torch.sum(x ** 2)
```

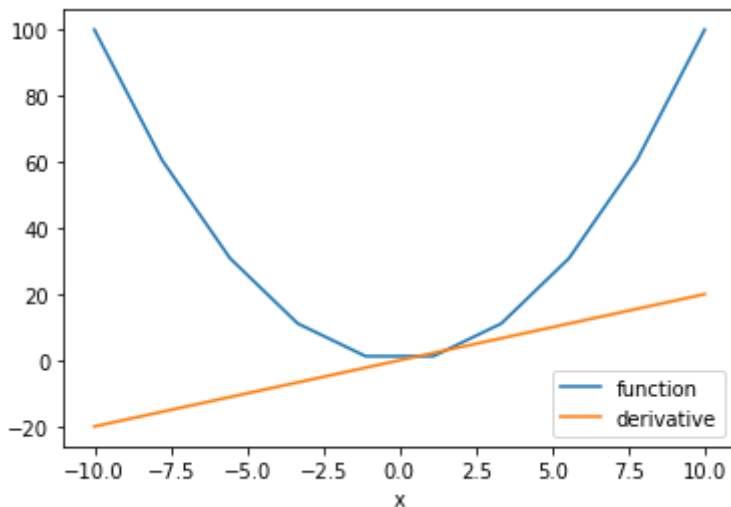
We can plot the function and its derivative

In [15]:

```
# Take the derivative with respect to multiple value. Plot out the function and its derivative

y.backward()

plt.plot(x.detach().numpy(), Y.detach().numpy(), label = 'function')
plt.plot(x.detach().numpy(), x.grad.detach().numpy(), label = 'derivative')
plt.xlabel('x')
plt.legend()
plt.show()
```



The orange line is the slope of the blue line at the intersection point, which is the derivative of the blue line.

The method `detach()` excludes further tracking of operations in the graph, and therefore the subgraph will not record operations. This allows us to then convert the tensor to a numpy array. To understand the sum operation [Click Here \(https://pytorch.org/tutorials/beginner/blitz/autograd_tutorial.html\)](https://pytorch.org/tutorials/beginner/blitz/autograd_tutorial.html).

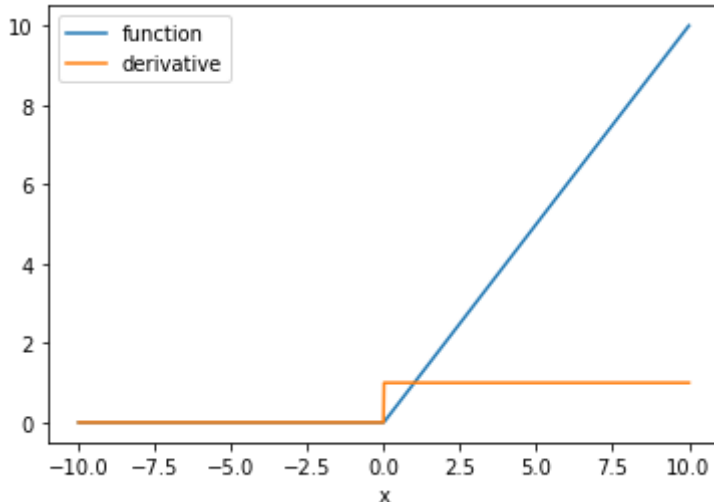
The **relu** activation function is an essential function in neural networks. We can take the derivative as follows:

In []:

In [16]:

```
# Take the derivative of Relu with respect to multiple value. Plot out the function and its derivative
```

```
x = torch.linspace(-10, 10, 1000, requires_grad = True)
Y = torch.relu(x)
y = Y.sum()
y.backward()
plt.plot(x.detach().numpy(), Y.detach().numpy(), label = 'function')
plt.plot(x.detach().numpy(), x.grad.detach().numpy(), label = 'derivative')
plt.xlabel('x')
plt.legend()
plt.show()
```



In [18]:

```
print(x)
print(Y)
print(y)
```

```
tensor([-10.0000, -9.9800, -9.9600, -9.9399, -9.9199, -9.8999, -
9.8799,
      -9.8599, -9.8398, -9.8198, -9.7998, -9.7798, -9.7598, -
9.7397,
      -9.7197, -9.6997, -9.6797, -9.6597, -9.6396, -9.6196, -
9.5996,
      -9.5796, -9.5596, -9.5395, -9.5195, -9.4995, -9.4795, -
9.4595,
      -9.4394, -9.4194, -9.3994, -9.3794, -9.3594, -9.3393, -
9.3193,
      -9.2993, -9.2793, -9.2593, -9.2392, -9.2192, -9.1992, -
9.1792,
      -9.1592, -9.1391, -9.1191, -9.0991, -9.0791, -9.0591, -
9.0390,
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6.0961,
      -6.0761, -6.0561, -6.0360, -6.0160, -5.9960, -5.9760, -
```

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	5.1351,	5.1552,	5.1752,	5.1952,	5.2152,	5.2352,
5.2553,						
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10.0000],
      grad_fn=<ReluBackward0>)
tensor(2502.5020, grad_fn=<SumBackward0>)

```

In [17]:

```
y.grad_fn
```

Out[17]:

```
<SumBackward0 at 0x7f3bcc97cc88>
```

Practice

Try to determine partial derivative $\frac{\partial f}{\partial u}$ of the following function where $u=2$ and $v=1$: $f=uv+(uv)^2$

In [19]:

```

# Practice: Calculate the derivative of f = u * v + (u * v) ** 2 at u = 2, v = 1
u = torch.tensor(2.0,requires_grad=True)
v = torch.tensor(1.0,requires_grad=True)
f = u * v + (u * v) ** 2
print("The result of v * u + u^2: ", f)
f.backward()
print("The partial derivative with respect to u: ", u.grad)
print("The partial derivative with respect to v: ", v.grad)

```

```


The result of v * u + u^2:  tensor(6., grad_fn=<AddBackward0>)
The partial derivative with respect to u:  tensor(5.)
The partial derivative with respect to v:  tensor(10.)

```

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About the Authors:

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