

Homework #1

2017012251 윤영훈

Data Preparation					
AND		OR		XOR	
0,0	0	0,0	0	0,0	0
0,1	0	0,1	1	0,1	1
1,0	0	1,0	1	1,0	1
1,1	1	1,1	1	1,1	0

1. Boolean operators : AND

Logistic regression (AND)

```
In [1]: import random
from math import exp, log
import numpy as np
```

Data prepration

```
In [2]: X = np.array([(0,0),(1,0),(0,1),(1,1)])
Y = np.array([0,0,0,1])
```

◀ Data Set을 numpy array로 나타냄

Model

```
In [3]: class logistic_regression_model():
def __init__(self):
self.w = np.random.normal(size = 2)
self.b = np.random.normal(size = 1)

def sigmoid(self, z):
return 1/(1 + exp(-z))

def predict(self, x):
z = np.inner(self.w, x) + self.b
a = self.sigmoid(z)
return a
```

◀ self.w, self.b를 numpy array로 나타냄

◀ numpy에서 지원하는 inner 함수 사용

```
In [4]: model = logistic_regression_model()
```

Training

```
In [5]: def train(X, Y, model, lr = 0.1):
dw0 = 0.0
dw1 = 0.0
db = 0.0
m = len(X)
cost = 0.0
for x,y in zip(X,Y):
a = model.predict(x)
if y == 1:
cost -= log(a)
else:
cost -= log(1-a)

dw0 += np.multiply(a-y, x[0])
dw1 += np.multiply(a-y, x[1])
db += (a-y)

cost /= m
model.w[0] -= lr * dw0/m
model.w[1] -= lr * dw1/m
model.b -= lr * db/m

return cost
```

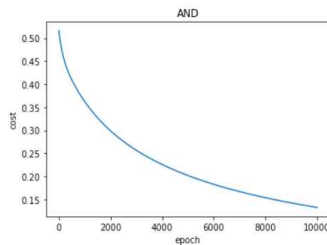
◀ numpy에서 지원하는 multiply 함수 사용

```
In [6]: for epoch in range(10000):
cost = train(X, Y, model, 0.1)
if epoch % 100 == 0:
print(epoch, cost)
```

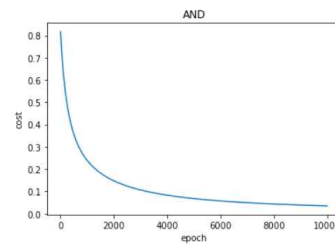
```
0 1.141848974190514
100 0.44571765404746155
200 0.3522160157117974
```

...

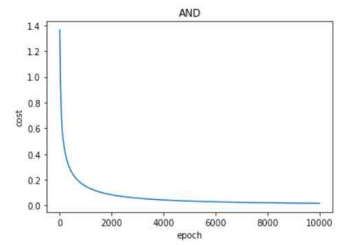
```
9700 0.017952426502209032
9800 0.017769159010130174
9900 0.017589554356108004
```



learning rate = 0.01



learning rate = 0.05



learning rate = 0.1

※ plot graph의 경우 epoch와 cost 값들을 각각 numpy array로 만들어 plot 함수를 통해 그래프화

▲ cost의 값이 점점 감소 → 학습이 정상적으로 진행중

Testing

```
In [7]: model.predict((0,0))
```

```
Out[7]: 1.2444400780361084e-05
```

```
In [8]: model.predict((0,1))
```

```
Out[8]: 0.020236703586892318
```

```
In [9]: model.predict((1,0))
```

```
Out[9]: 0.020236692418202265
```

```
In [10]: model.predict((1,1))
```

```
Out[10]: 0.9716563914463304
```

∴ (0,0), (0,1), (1,0)에서는 0과 가까운 값이, (1,1)에서는 1과 가까운 값이 출력되는 것을 볼 때 AND 연산자가 잘 학습되었음

2. Boolean operators : OR

Logistic regression (OR)

```
In [1]: import random
from math import exp, log
import numpy as np
```

Data preparation

```
In [2]: X = np.array([(0,0),(1,0),(0,1),(1,1)])
Y = np.array([0,1,1,1])
```

◀ Data Set을 numpy array로 나타냄

Model

```
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def sigmoid(self, z):
return 1/(1 + exp(-z))

def predict(self, x):
z = np.inner(self.w, x) + self.b
a = self.sigmoid(z)
return a
```

◀ self.w, self.b를 numpy array로 나타냄

◀ numpy에서 지원하는 inner 함수 사용

```
In [4]: model = logistic_regression_model()
```

Training

```
In [5]: def train(X, Y, model, lr = 0.1):
dw0 = 0.0
dw1 = 0.0
db = 0.0
m = len(X)
cost = 0.0
for x,y in zip(X,Y):
    a = model.predict(x)
    if y == 1:
        cost += log(a)
    else:
        cost += log(1-a)

    dw0 += np.multiply(a-y, x[0])
    dw1 += np.multiply(a-y, x[1])
    db += (a-y)

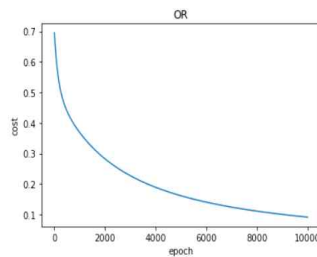
cost /= m
model.w[0] -= lr * dw0/m
model.w[1] -= lr * dw1/m
model.b -= lr * db/m

return cost
```

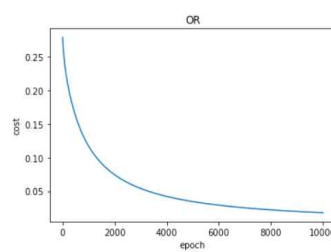
◀ numpy에서 지원하는 multiply 함수 사용

```
In [6]: for epoch in range(10000):
cost = train(X, Y, model, 0.1)
if epoch % 100 == 0:
    print(epoch, cost)
```

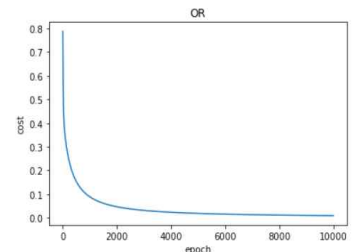
```
0 1.6654057288269368
100 0.3330699164779229
200 0.2541081392257821
...
9700 0.00955360761852166
9800 0.009454766373164133
9900 0.009357934477127045
```



learning rate = 0.01



learning rate = 0.05



learning rate = 0.1

※ plot graph의 경우 epoch와 cost 값들을 각각 numpy array로 만들어 plot 함수를 통해 그래프화

▲ cost의 값이 점점 감소 → 학습이 정상적으로 진행중

Testing

```
In [7]: model.predict((0,0))
```

```
Out[7]: 0.020442614722577
```

```
In [8]: model.predict((0,1))
```

```
Out[8]: 0.9918433567595195
```

```
In [9]: model.predict((1,0))
```

```
Out[9]: 0.9918273348811616
```

```
In [10]: model.predict((1,1))
```

```
Out[10]: 0.9999985858289085
```

∴ (0,0)에서는 0과 가까운 값이, (0,1), (1,0), (1,1)에서는 1과 가까운 값이 출력되는 것을 볼 때 OR 연산자가 잘 학습되었음

3. Boolean operators : XOR

Logistic regression (XOR)

```
In [1]: import random
        from math import exp, log
        import numpy as np
```

Data preparation

```
In [2]: X = np.array([(0,0),(1,0),(0,1),(1,1)])
        Y = np.array([0,1,1,0])
```

◀ Data Set을 numpy array로 나타냄

Model

```
In [3]: class logistic_regression_model():
        def __init__(self):
            self.w = np.random.normal(size = 2)
            self.b = np.random.normal(size = 1)

        def sigmoid(self, z):
            return 1/(1 + exp(-z))

        def predict(self, x):
            z = np.inner(self.w, x) + self.b
            a = self.sigmoid(z)
            return a
```

◀ self.w, self.b를 numpy array로 나타냄

◀ numpy에서 지원하는 inner 함수 사용

```
In [4]: model = logistic_regression_model()
```

Training

```
In [5]: def train(X, Y, model, lr = 0.1):
        dw0 = 0.0
        dw1 = 0.0
        db = 0.0
        m = len(X)
        cost = 0.0
        for x,y in zip(X,Y):
            a = model.predict(x)
            if y == 1:
                cost += log(a)
            else:
                cost += log(1-a)

            dw0 += np.multiply(a-y, x[0])
            dw1 += np.multiply(a-y, x[1])
            db += (a-y)

        cost /= m
        model.w[0] -= lr * dw0/m
        model.w[1] -= lr * dw1/m
        model.b -= lr * db/m

        return cost
```

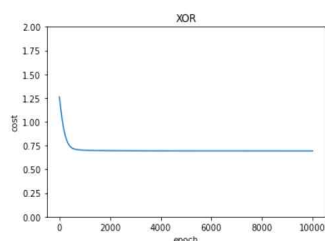
◀ numpy에서 지원하는 multiply 함수 사용

```
In [6]: for epoch in range(10000):
        cost = train(X, Y, model, 0.1)
        if epoch % 100 == 0:
            print(epoch, cost)
```

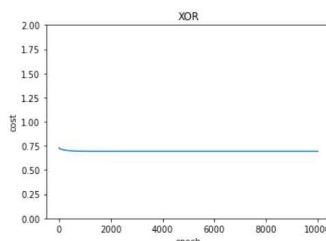
```
0 0.7053910348959286
100 0.6970513620804439
200 0.6945353790751042
```

...

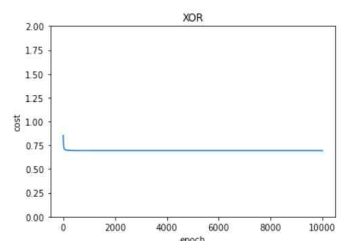
```
9700 0.6931471805599453
9800 0.6931471805599453
9900 0.6931471805599453
```



learning rate = 0.01



learning rate = 0.05



learning rate = 0.1

※ plot graph의 경우 epoch와 cost 값들을 각각 numpy array로 만들어 plot 함수를 통해 그래프화

▲ cost의 값이 감소하지 않음 → 학습이 정상적으로 진행되지 않음

Testing

```
In [7]: model.predict((0,0))
```

```
Out[7]: 0.5
```

```
In [8]: model.predict((0,1))
```

```
Out[8]: 0.5
```

```
In [9]: model.predict((1,0))
```

```
Out[9]: 0.5
```

```
In [10]: model.predict((1,1))
```

```
Out[10]: 0.5
```

∴ 이론상 (0,0), (1,1)에서는 0과 가까운 값이, (0,1), (1,0)에서는 1과 가까운 값이 출력되어야 하지만 그렇지 못함
→ XOR 연산자가 잘 학습되지 못함