AND

```
In [3]:
x = [[0,0],[0,1],[1,0],[1,1]]
y = [[0], [0], [0], [1]]
x_tensor = torch.tensor(x).float().to(device)
y_tensor = torch.tensor(y).float().to(device)
for epoch in range(5000):
    out = net(x tensor)
    loss = loss_func(out,y_tensor)
    optimizer.zero grad()
    loss.backward()
    optimizer.step()
    if epoch % 1000 ==0:
        print(f'迭代次数:{epoch}')
        print(f'误差:{loss}')
 迭代次数:0
 误差:0.3514098525047302
 迭代次数:1000
 误差:0.09353005886077881
 迭代次数:2000
 误差:0.05892698094248772
 迭代次数:3000
 误差:0.042214177548885345
 迭代次数:4000
 误差:0.032368630170822144
In [4]:
out = net(x_tensor).cuda()
print(f'out:{out.data}')
torch.save(net, './AND.pkl')
 out:tensor([[0.0115],
        [0.1728],
        [0.1729],
        [0.7903]], device='cuda:0')
```

```
x= [[0,0],[0,1],[1,0],[1,1]]
x_tensor = torch.tensor(x).float().to(device)
out_tensor = net(x_tensor).cuda()
for out in out_tensor:
    if out>0.5:
        print(1)
    else:
        print(0)
```

权重输出

```
for name, param in net.named_parameters():
    print(name)
    print(param.data)
    print("-----")

0.weight
    tensor([[2.8926, 2.8921]], device='cuda:0')
-------
0.bias
    tensor([-4.4578], device='cuda:0')
```

结论

权重如上一个单元格所示,为[2.8926, 2.8921],偏置为[-4.4578] 使用的激活函数为Sigmoid()

```
print(net)

Sequential(
   (0): Linear(in_features=2, out_features=1, bi as=True)
   (1): Sigmoid()
)
```

网络结构如上所示