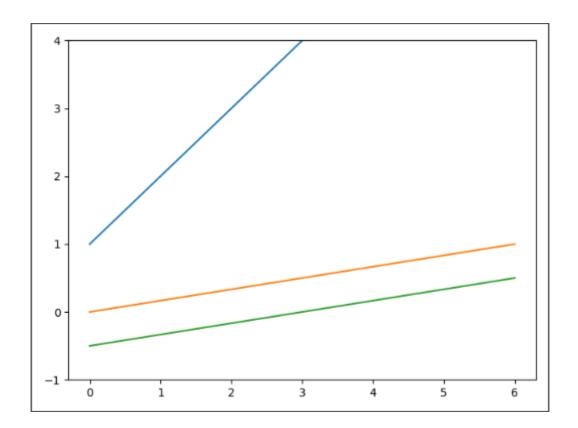
## 归一化

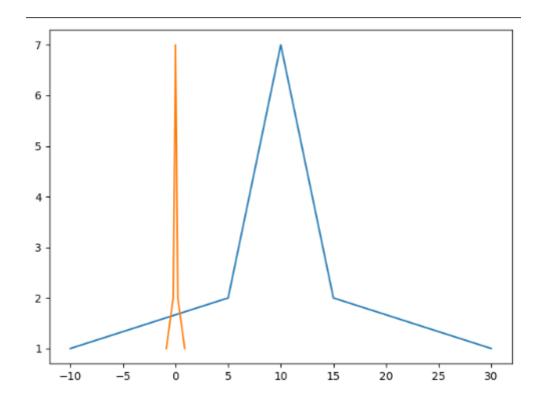
简而言之,归一化的目的就是使得预处理的数据被限定在一定的范围内(比如[0,1]或者[-1,1]),从而消除奇异样本数据导致的不良影响。[https://zhuanlan.zhihu.com/p/424518359]

## 对比效果

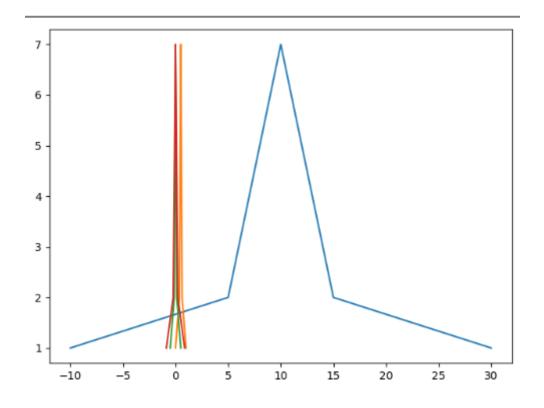
```
import numpy as np
import matplotlib.pyplot as plt
# 归一化的两种方式
def normalization1(x):
    '''归一化 (0~1) '''
    '''x = (x-x min)/(x max-x min)'''
    return [(float(i) - min(x)) / float(max(x) - min(x)) for i in x]
def normalization2(x):
    '''归一化 (-1~1) '''
    '''x_=(x-x_mean)/(x_max-x_min)'''
    return [(float(i) - np.mean(x)) / (max(x) - min(x)) for i in x]
test_list = [1,2,3,4,5,6,7]
normal_1 = normalization1(test_list)
normal_2 = normalization2(test_list)
plt.ylim(-1, 4)
plt.plot(test_list)
plt.plot(normal_1)
plt.plot(normal_2)
plt.show()
```



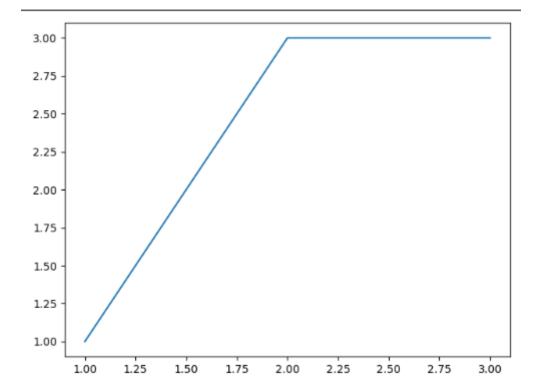
```
import numpy as np
 import matplotlib.pyplot as plt
#归一化的两种方式
 def Normalization1(x):
                   '''归一化 (0~1) '''
                   '''x = (x-x min)/(x max-x min)'''
                 return [(float(i)-min(x))/float(max(x)-min(x)) for i in x]
 def Normalization2(x):
                   '''归一化 (-1~1) '''
                   '''x = (x-x mean)/(x max-x min)'''
                 return [(float(i)-np.mean(x))/(max(x)-min(x)) for i in x]
#标准化
 def z score(x):
                  ""x*=(x-\mu)/\sigma""
                 x mean=np.mean(x)
                 s2=sum([(i-np.mean(x))*(i-np.mean(x)) for i in x])/len(x)
                 return [(i-x mean)/s2 for i in x]
cs=[]
for i in 1:
                 c=1.count(i)
                 cs.append(c)
print(cs)
n1 = Normalization1(1)
n2 = Normalization2(1)
z = z_score(1)
print('n1:\t', n1)
print('n2:\t', n2)
print('z:\t', z)
蓝线为原始数据,橙线为z
plt.plot(1,cs)
plt.plot(z,cs)
plt.show()
[1, 2, 2, 3, 3, 3, 4, 4, 4, 4, 5, 5, 5, 5, 5, 6, 6, 6, 6, 6, 6, 7, 7, 7, 7, 7, 7, 7, 7, 6, 6, 6, 6, 6, 6, 5, 5, 5, 5
                                 [0.0, 0.375, 0.375, 0.4, 0.4, 0.4, 0.425, 0.425, 0.425, 0.425, 0.45, 0.45, 0.45, 0.45, 0.45, 0.45, 0.475, 0.475
                                 [-0.5, -0.125, -0.125, -0.1, -0.1, -0.1, -0.075, -0.075, -0.075, -0.075, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.05, -0.0
n2:
                                  [-0.875, -0.21875, -0.21875, -0.175, -0.175, -0.175, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.0875, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0.13125, -0
z:
```



```
plt.plot(1, cs)
plt.plot(n1, cs)
plt.plot(n2, cs)
plt.plot(z, cs)
plt.show()
```



```
test_list = [1,2,3,4,5,6,7.1]
 print(type(test list))
 print(type(test_list[0]))
 test_list = [1,2,3,4,5,6,7.1]
 print(type(test_list))
 print(type(test_list[6]))
 <class 'list'>
 <class 'int'>
 <class 'list'>
 <class 'float'>
看看函数
 test_list = [1,2,3,4,5,6,7.1]
 print(type(max(test_list)))
 print(type(min(test_list)))
 <class 'float'>
 <class 'int'>
看看函数返回类型
 test_list = [1,2,3,4,5,6,7]
 print(np.mean(test_list)) # return numpy.float64
 print(type(np.mean(test_list)))
 4.0
 <class 'numpy.float64'>
画图:画图像描点一样,从前往后化,若x中有重复的那就画在同一y轴上。
 import matplotlib.pyplot as plt
 x = [1,2,2,2,3,3,3] # 画图等价[123]
 y = [x.count(i) for i in x] # 与x画图等价[133]
 # plt.plot(x) # 蓝
 plt.plot(x, y) # x[123] y[133]
 plt.show()
```



```
import matplotlib.pyplot as plt
x = [1,2,2,2,3,3,3]
y = [1,2,2,1,3,3,3]
plt.plot(x, y)
plt.show()
```

