

基于机器学习的冷轧板行功效系数 矩阵辨识实验

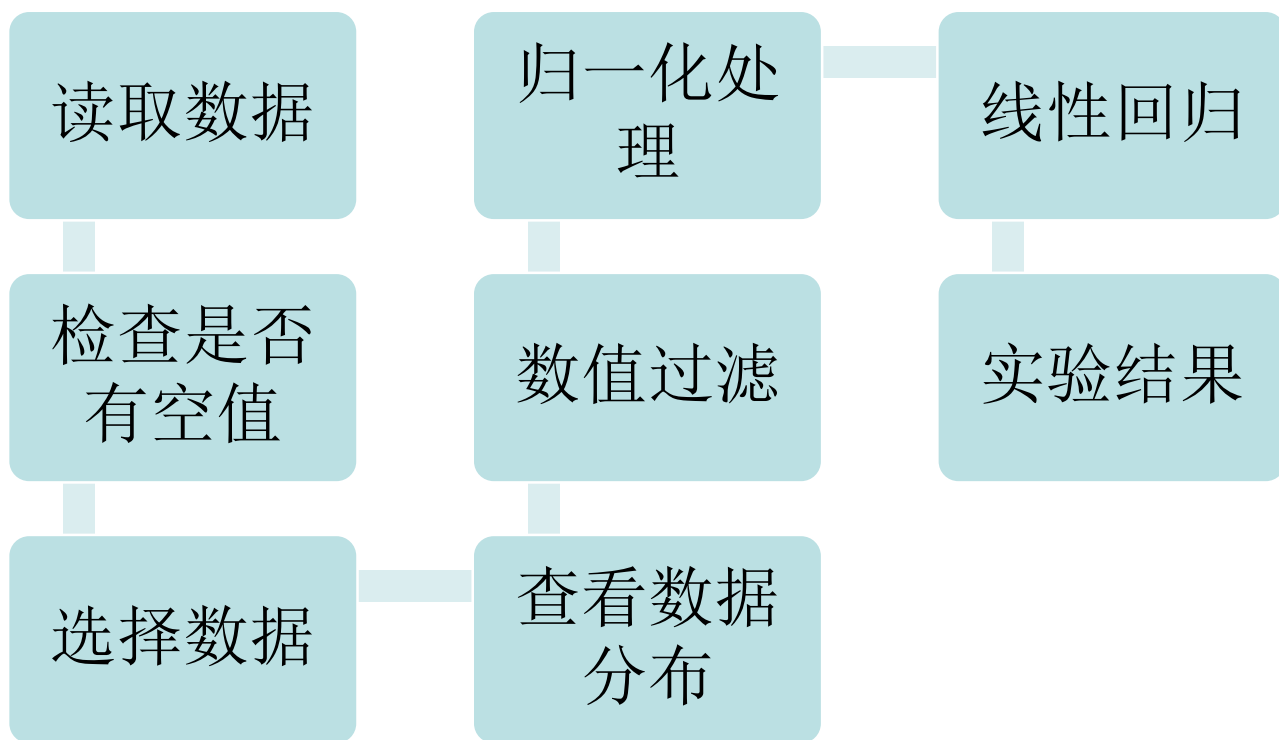
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2070712
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- 一、数据处理
- 二、机器学习方法
 - 1、最小二乘法
 - 2、脊回归
 - 3、神经网络
- 三、实验结果

一、数据处理

流程图



一、数据处理

读取数据，并检查数据是否有空，然后每行数据减去上一行，数据维度（7113， 70）

	V6	V7	V8	V9	V10	V11	...	V90	V91	V92	V93	V94	V95
1	-18.1	-2.0	-2.2	8.3	-0.4	-0.94	...	1.0090	1.2890	1.123	0.650	1.077	1.4575
2	-16.9	2.0	9.3	6.8	9.5	-0.38	...	0.6490	1.1940	0.849	0.550	0.837	1.5765
3	-2.5	3.0	2.9	-1.3	1.1	-0.41	...	1.1236	1.5866	1.129	1.270	1.914	3.5270
4	-5.5	5.0	3.4	-0.3	22.0	-0.42	...	-0.5625	-0.5359	0.299	1.589	2.167	5.1750
5	-10.3	3.0	3.7	1.0	13.2	-0.46	...	-0.9451	-0.6617	-0.112	0.126	1.507	3.6870

数据前5行

V6	0
V7	0
V8	0
V9	0
V10	0
...	..
V91	0
V92	0
V93	0
V94	0
V95	0

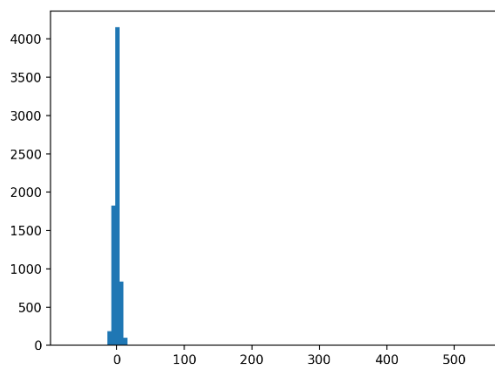
检查是否有空

一、数据处理

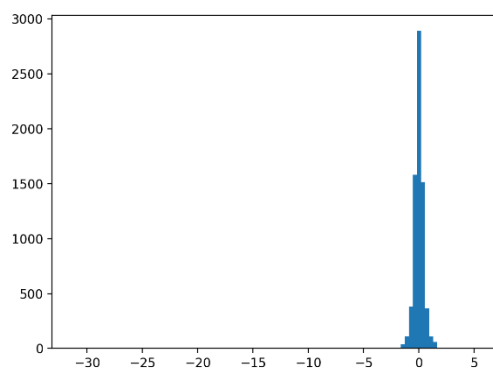
划分数据：选择五号机架的轧制力 V10、五号轧机的工作辊 V15、五号轧机的中间辊弯辊力 V16 作为自变量，V62-V95为因变量

X维度(7113, 3) Y维度(7113, 34)

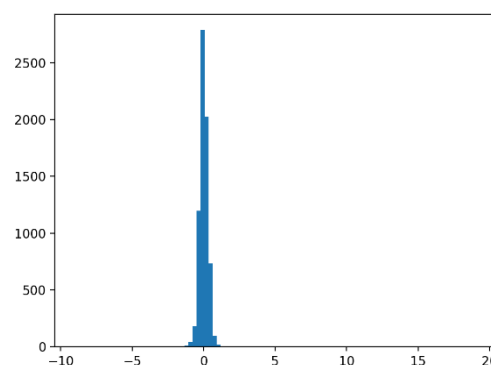
数据数值分布如下图（未处理）



V10 数据分布
最大值：533.7
最小值：-68.2



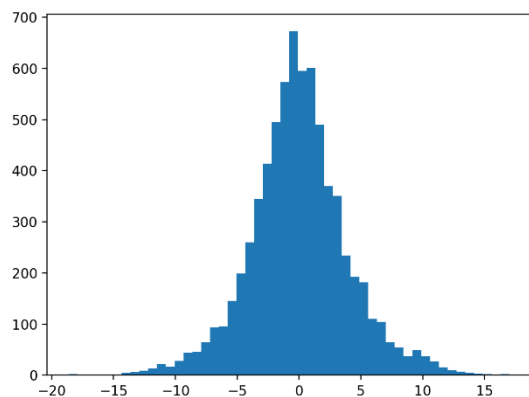
V15 数据分布
最大值：5.31
最小值：-31.33



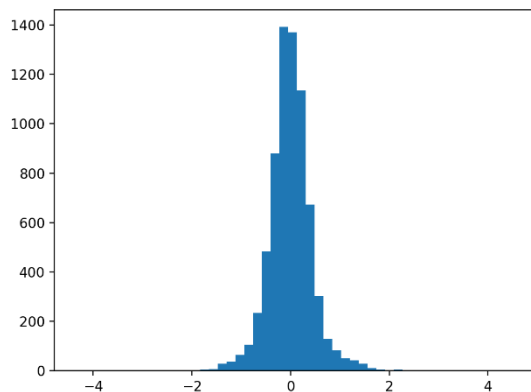
V16 数据分布
最大值：19.307
最小值：-9

一、数据处理

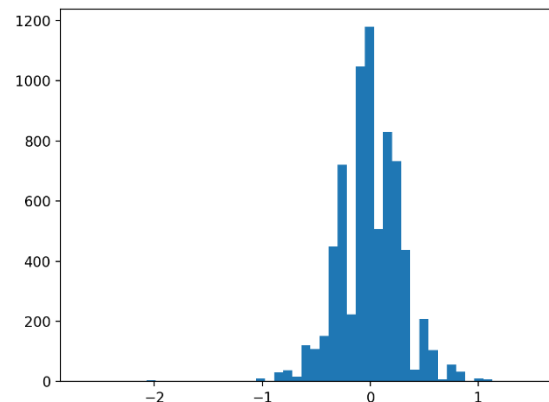
数据过滤：最大最小值限制后的数据，下方区间处理后的数据取值区间
X-(7098, 3) Y-(7098, 34)



V10
[-20, 20]



V15
[-5, 5]



V16
[-2, 2]

一、数据处理-归一化

由于不同的特征数据量级差异过大，需要对数据进行归一化处理，对所有特征进行 MinMax 缩放，使其数值在[-1, 1]范围内，计算公式如下：

$$x_{\text{std}} = \frac{x - \min(x)}{\max(x) - \min(x)}$$

$$x_{\text{scaled}} = x_{\text{std}} * (\text{MAX} - \text{MIN}) + \text{MIN}$$

其中 MAX 为 1，MIN 为-1，max()表示求特征的最大值，min()表示求特征的最小值。

注：将X按列进行MINMAX归一化，Y不做处理

二、机器学习方法-模型分析

板形调控功效系数是在一种调控手段的单位调节量作用下，冷轧机架的轧辊辊缝凸度沿带钢宽度方向上各处的变化量，可表示为：

$$\Delta f_j(x_i) = \Delta u_i \cdot eff_{i,j}^{(1)} + \Delta u_{i,j}^2 \cdot eff_{i,j}^{(2)} + \cdots + \Delta u_i^n \cdot eff_{i,j}^{(n)}$$

$$eff_{i,j} = \frac{\Delta f_j(x_i)}{\Delta u_j}$$

$$\begin{aligned} E = \frac{\Delta f(x)}{\Delta u} &= \begin{bmatrix} \Delta f(x_1) \\ \Delta f(x_2) \\ \vdots \\ \Delta f(x_{34}) \end{bmatrix} \begin{bmatrix} \frac{1}{\Delta u_1} & \frac{1}{\Delta u_2} & \cdots & \frac{1}{\Delta u_5} \end{bmatrix} \\ &= \begin{bmatrix} eff_{1,1} & eff_{1,2} & \cdots & eff_{1,5} \\ eff_{2,1} & eff_{2,2} & \cdots & eff_{2,5} \\ \vdots & \vdots & \ddots & \vdots \\ eff_{34,1} & eff_{34,2} & \cdots & eff_{34,5} \end{bmatrix} \end{aligned}$$

二、机器学习方法-最小二乘法

最小二乘法：

由于功效系数只考虑线性部分，故可用最小二乘法求解功效矩阵，最小二乘法的目标是最小化误差函数：

目标函数：
$$\min_W ||XW - Y||_2^2$$

正规方程解：
$$W = (X^T X)^{-1} X^T y$$

二、机器学习方法-脊回归

脊回归：

脊回归就是在线性回归的基础上，加上正则化

目标函数：
$$\min_W ||XW - Y||_2^2 + \alpha ||W||_2^2$$

正规方程解：
$$W = (X^T X + \alpha E)^{-1} X^T y$$

二、机器学习方法- 神经网络

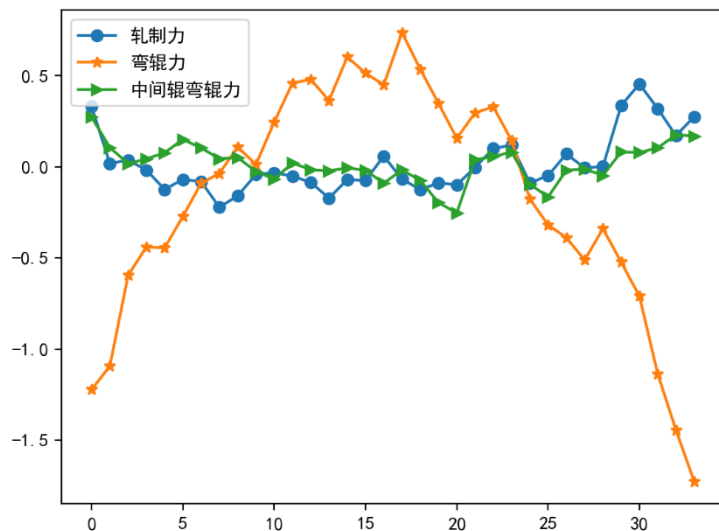
决定系数:

$$R^2 = 1 - \frac{\sum_{i=1}^N (y_i - \hat{y}_i)^2}{\sum_{i=1}^N (y_i - \bar{y})^2}$$

模型预测越准确，上式中等式右侧趋于0

三、实验结果

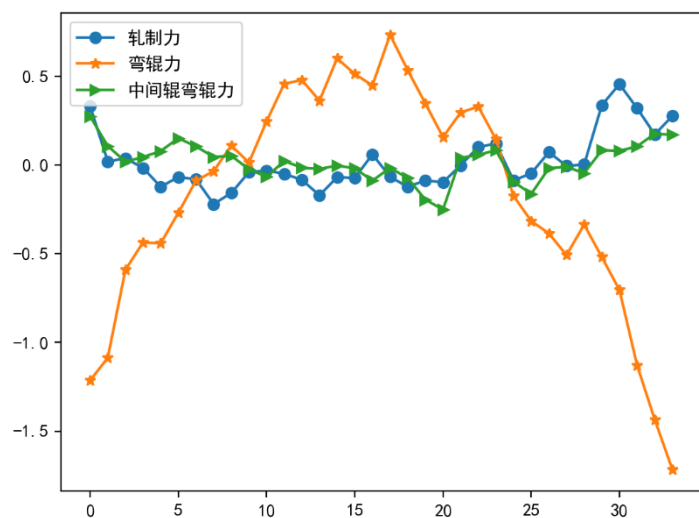
最小二乘法：



```
[[ 3.31865157e-01 -1.22178559e+00 2.73699753e-01]
 [ 1.86895045e-02 -1.09331430e+00 1.06137945e-01]
 [ 3.71776017e-02 -5.95523409e-01 1.82425399e-02]
 [-1.71287559e-02 -4.38868352e-01 4.38871121e-02]
 [-1.23651958e-01 -4.44668345e-01 7.78805550e-02]
 [-6.85024837e-02 -2.71277549e-01 1.50462358e-01]
 [-8.08998168e-02 -8.96930581e-02 1.04180497e-01]
 [-2.21194720e-01 -3.54954166e-02 4.49070455e-02]
 [-1.58712521e-01 1.09683207e-01 5.10757376e-02]
 [-4.15616219e-02 1.27686208e-02 -2.47734373e-02]
 [-3.16678605e-02 2.45850454e-01 -6.74549825e-02]
 [-5.08941901e-02 4.59187086e-01 2.18747500e-02]
 [-8.45792623e-02 4.80855766e-01 -1.60214917e-02]
 [-1.71740116e-01 3.63705456e-01 -2.23576362e-02]
 [-6.97224648e-02 6.03241333e-01 -4.65438152e-03]
 [-7.39983638e-02 5.14571020e-01 -1.94750804e-02]
 [ 5.90623213e-02 4.50135560e-01 -8.91606087e-02]
 [-6.74515780e-02 7.36872194e-01 -1.69190272e-02]
 [-1.24492169e-01 5.36092431e-01 -7.42458368e-02]
 [-8.76706162e-02 3.48079242e-01 -1.96505268e-01]
 [-9.85260245e-02 1.57075142e-01 -2.52418998e-01]
 [-2.45066875e-03 2.96590730e-01 3.91251403e-02]
 [ 1.00472998e-01 3.30110002e-01 5.68588077e-02]
 [ 1.19707561e-01 1.44805897e-01 8.17686804e-02]
 [-8.82634640e-02 -1.76281494e-01 -9.54750638e-02]
 [-4.76836767e-02 -3.18276088e-01 -1.64563210e-01]
 [ 7.13986544e-02 -3.88294734e-01 -1.86885572e-02]
 [-5.85933994e-03 -5.11429514e-01 -1.16786427e-02]
 [ 1.50578203e-03 -3.38836047e-01 -4.67033909e-02]
 [ 3.36265778e-01 -5.22396503e-01 8.23260968e-02]
 [ 4.55460201e-01 -7.08978666e-01 7.89164440e-02]
 [ 3.20023623e-01 -1.13488354e+00 1.04359300e-01]
 [ 1.71698687e-01 -1.44302375e+00 1.75580987e-01]
 [ 2.75609625e-01 -1.72596670e+00 1.69486419e-01]]
```

三、实验结果

脊回归：



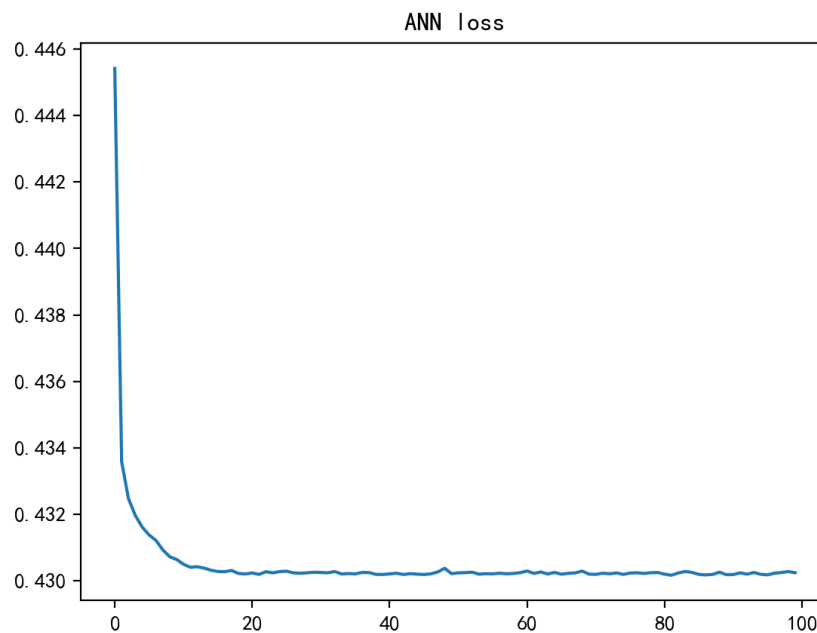
```
[[ 3.31187481e-01 -1.21426488e+00 2.73013842e-01]
 [ 1.84570457e-02 -1.08661882e+00 1.05953867e-01]
 [ 3.70059872e-02 -5.91864513e-01 1.82822705e-02]
 [-1.71867384e-02 -4.36185950e-01 4.38047811e-02]
 [-1.23553660e-01 -4.41977591e-01 7.76651999e-02]
 [-6.84354829e-02 -2.69651209e-01 1.49979429e-01]
 [-8.07856664e-02 -8.91760180e-02 1.03824777e-01]
 [-2.20879214e-01 -3.53300253e-02 4.47225819e-02]
 [-1.58455441e-01 1.08969517e-01 5.08569777e-02]
 [-4.15038279e-02 1.26857816e-02 -2.46974591e-02]
 [-3.15835918e-02 2.44346288e-01 -6.72687788e-02]
 [-5.07242659e-02 4.56355500e-01 2.17173357e-02]
 [-8.43631136e-02 4.77890455e-01 -1.60581137e-02]
 [-1.71424659e-01 3.61442227e-01 -2.23679674e-02]
 [-6.95006396e-02 5.99526239e-01 -4.74725638e-03]
 [-7.37909840e-02 5.11401543e-01 -1.95035739e-02]
 [ 5.90553766e-02 4.47399890e-01 -8.89181070e-02]
 [-6.72077187e-02 7.32339558e-01 -1.69910804e-02]
 [-1.24217402e-01 5.32789424e-01 -7.40984676e-02]
 [-8.75068322e-02 3.45957173e-01 -1.95902666e-01]
 [-9.83949543e-02 1.56131919e-01 -2.51595802e-01]
 [-2.38018156e-03 2.94763975e-01 3.89428332e-02]
 [ 1.00406498e-01 3.28095721e-01 5.66272523e-02]
 [ 1.19579891e-01 1.43928586e-01 8.14848573e-02]
 [-8.81893060e-02 -1.75202410e-01 -9.51340378e-02]
 [-4.77077251e-02 -3.16306260e-01 -1.63955916e-01]
 [ 7.12144952e-02 -3.85894484e-01 -1.85501305e-02]
 [-5.95733837e-03 -5.08290593e-01 -1.15570530e-02]
 [ 1.42677024e-03 -3.36749036e-01 -4.64884592e-02]
 [ 3.35693194e-01 -5.19135597e-01 8.21836371e-02]
 [ 4.54679011e-01 -7.04547965e-01 7.88353668e-02]
 [ 3.19352748e-01 -1.12787116e+00 1.04237544e-01]
 [ 1.71188073e-01 -1.43416234e+00 1.75240598e-01]
 [ 2.74892079e-01 -1.71534691e+00 1.69229651e-01]]
```

三、实验结果

神经网络训练集损失和测试集决定系数：

测试集决定系数

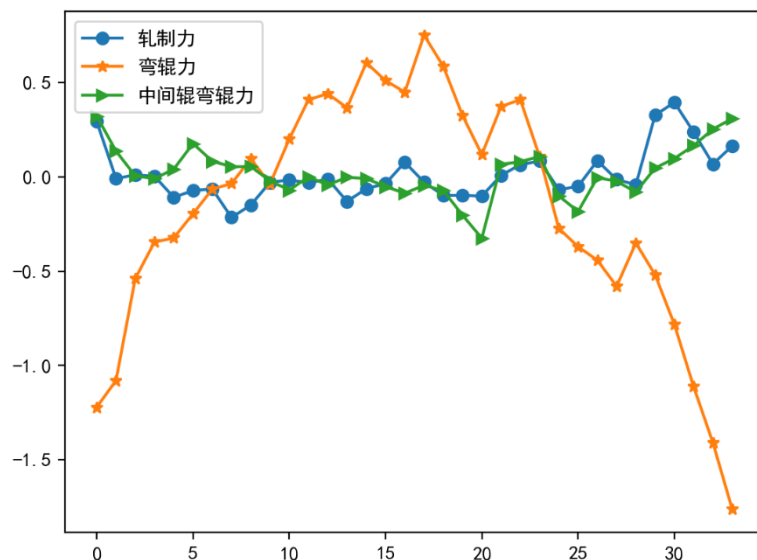
$R=0.005442743562043028$



训练过程中的损失

三、实验结果

神经网络：



```
[[ 2.93931636e-01 -1.22083872e+00 3.22382888e-01]
[-8.55330950e-03 -1.07983801e+00 1.37252545e-01]
[ 1.06539997e-02 -5.39463739e-01 5.43286514e-03]
[ 3.47359542e-03 -3.45256505e-01 -8.86834759e-03]
[-1.08286534e-01 -3.23696438e-01 4.12171520e-02]
[-7.25298446e-02 -1.95583805e-01 1.76887035e-01]
[-6.37489001e-02 -6.51925731e-02 8.34591262e-02]
[-2.15144393e-01 -3.51971256e-02 5.43029358e-02]
[-1.49386403e-01 9.63664284e-02 5.45737344e-02]
[-2.85801121e-02 -3.70751174e-02 -2.07788685e-02]
[-1.57791725e-02 2.01004874e-01 -6.97690388e-02]
[-3.13596424e-02 4.10313633e-01 2.89333761e-04]
[-1.15100248e-02 4.40836276e-01 -4.23377676e-02]
[-1.29908500e-01 3.66949980e-01 -1.49733993e-03]
[-6.25297632e-02 6.05756234e-01 -8.76269506e-03]
[-3.56811957e-02 5.10376605e-01 -5.13248816e-02]
[ 7.76406751e-02 4.49854964e-01 -8.66728303e-02]
[-2.78209625e-02 7.51669267e-01 -4.42000331e-02]
[-9.80728265e-02 5.86292954e-01 -7.23323584e-02]
[-9.72725301e-02 3.25283303e-01 -2.03129076e-01]
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[ 8.63090298e-03 3.73438483e-01 6.66607872e-02]
[ 6.48135062e-02 4.09818265e-01 8.09112679e-02]
[ 8.44138755e-02 1.07716335e-01 1.08280774e-01]
[-6.73255049e-02 -2.73175571e-01 -1.00943826e-01]
[-5.01277893e-02 -3.70116949e-01 -1.83330897e-01]
[ 8.59649863e-02 -4.41547866e-01 -3.83544762e-03]
[-1.33482931e-02 -5.80071224e-01 -2.32477974e-02]
[-4.19154827e-02 -3.50594959e-01 -8.01649043e-02]
[ 3.27394268e-01 -5.18863316e-01 4.71151284e-02]
[ 3.95044785e-01 -7.81748856e-01 9.77949351e-02]
[ 2.38031276e-01 -1.11226113e+00 1.65951084e-01]
[ 6.82939068e-02 -1.40879973e+00 2.53019085e-01]
[ 1.63939302e-01 -1.76020618e+00 3.09424470e-01]]
```


Github:

https://github.com/langdu/AI_course/tree/main/function_matrix_latest

Thank You