

软件质量分析期末考试

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1 第一题

1.1 相关知识点

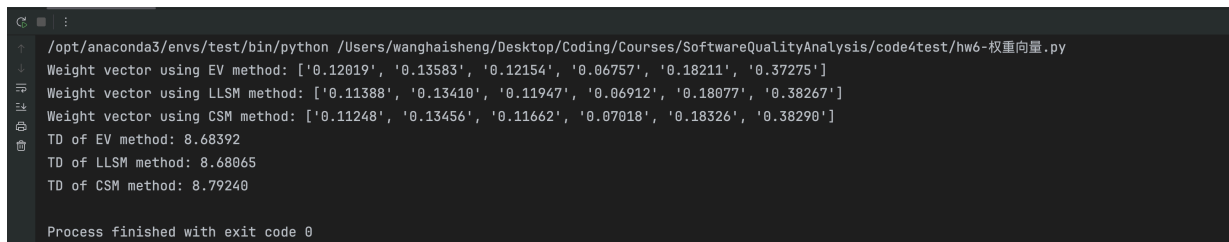
对于给定的正互反判断矩阵 A ，可以通过下面三种方法来估算权重向量：

1. 右特征向量法 (EigenVector Method, EV)
2. 对数最小二乘法 (Logarithmic Least Squares Method, LLSM)
3. 卡方最小二乘法 (Chi-Square Minimization Method, CSM)。

最终应当选择 TD 值最小的方法。

1.2 答案

1.2.1 代码运行结果



```
/opt/anaconda3/envs/test/bin/python /Users/wanghaisheng/Desktop/Coding/Courses/SoftwareQualityAnalysis/code4test/hw6-权重向量.py
Weight vector using EV method: ['0.12019', '0.13583', '0.12154', '0.06757', '0.18211', '0.37275']
Weight vector using LLSM method: ['0.11388', '0.13410', '0.11947', '0.06912', '0.18077', '0.38267']
Weight vector using CSM method: ['0.11248', '0.13456', '0.11662', '0.07018', '0.18326', '0.38290']
TD of EV method: 8.68392
TD of LLSM method: 8.68065
TD of CSM method: 8.79240
Process finished with exit code 0
```

图 1: 运行结果

1.2.2 题目答案

权重向量:

Weight vector using EV method: ['0.12019', '0.13583', '0.12154', '0.06757', '0.18211', '0.37275']

Weight vector using LLSM method: ['0.11388', '0.13410', '0.11947', '0.06912', '0.18077', '0.38267']

Weight vector using CSM method: ['0.11248', '0.13456', '0.11662', '0.07018', '0.18326', '0.38290']

“强度” TD 计算:

TD of EV method: 8.68392

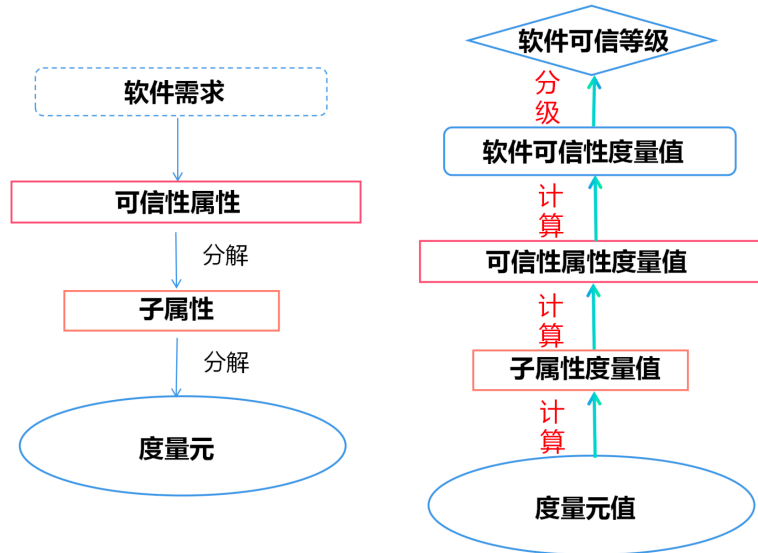
TD of LLSM method: 8.68065

TD of CSM method: 8.79240

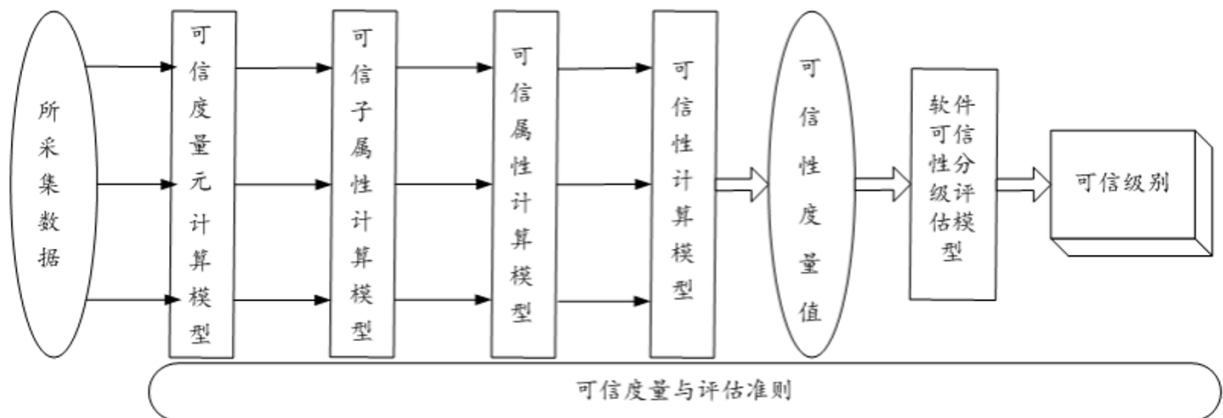
最终选择 TD 值最小的方法, 即 LLSM 对应的 8.68065。

2 第二题

2.1 相关知识点



由课堂内容可知，分析一个软件的可信等级，需要先自上而下地将软件需求解构为可信性属性、子属性、度量元；在计算时，要自下而上地计算每一个子类，然后得到最终的可信等级。更细节的流程图如下所示。



在最后一步，通过下述分级模型来判定软件可信等级。

软件可信度量值要求（黄金分割）	可信属性要求（多数原则）	可信等级
$9.5 \leq T$	1. 低于 9.5 分的关键属性个数不超过 $n - \lceil n \cdot \frac{2}{3} \rceil$ 个 2. 没有低于 8.5 分的可信属性	V
$8.5 \leq T < 9.5$ 或者 $T > 9.5$ 且不能评为 V 级别	1. 低于 8.5 分的关键属性个数不超过 $n - \lceil n \cdot \frac{2}{3} \rceil$ 个 2. 没有低于 7.0 分的可信属性	IV
$7.0 \leq T < 8.5$ 或者 $T > 8.5$ 且不能评为 IV 级别及以上者	1. 低于 7.0 分的关键属性个数不超过 $n - \lceil n \cdot \frac{2}{3} \rceil$ 个 2. 没有低于 4.5 分的可信属性	III
$4.5 \leq T < 7.0$ 或者 $T > 7.0$ 且不能评为 III 级别及以上者	1. 低于 4.5 分的关键属性个数不超过 $n - \lceil n \cdot \frac{2}{3} \rceil$ 个	II
$T < 4.5$ 或者 $T > 4.5$ 且不能评为 II 级别及以上者	1. 无要求	I

2.2 答案

2.2.1 代码运行结果

```

/opt/anaconda3/envs/test/bin/python /Users/wanghaisheng/Desktop/Coding/Courses/SoftwareQualityAnalysis/code4test/hw7-可信等级判定.py
ID1: 8.50652 8.27245 8.07854 8.49687 8.15320 8.94651 8.93288 9.24878 9.64811 软件信任值为 8.561 信任等级为 III
ID2: 7.83745 8.51402 8.38193 8.44011 8.42742 8.37854 8.55674 8.68562 8.90258 软件信任值为 8.468 信任等级为 III
ID3: 8.51307 8.99790 7.82639 8.39207 8.50961 8.52526 8.39207 7.64984 9.12939 软件信任值为 8.428 信任等级为 III
ID4: 8.87679 9.15556 8.80373 8.73551 8.92864 9.04323 8.99919 8.99500 9.32376 软件信任值为 8.966 信任等级为 IV

Process finished with exit code 0

```

图 2: 运行结果

2.2.2 题目答案

ID1: 8.50652 8.27245 8.07854 8.49687 8.15320 8.94651 8.93288 9.24878 9.64811 软件信任值为 8.561 信任等级为 III

ID2: 7.83745 8.51402 8.38193 8.44011 8.42742 8.37854 8.55674 8.68562 8.90258 软件信任值为 8.468 信任等级为 III

ID3: 8.51307 8.99790 7.82639 8.39207 8.50961 8.52526 8.39207 7.64984 9.12939 软件信任值为 8.428 信任等级为 III

ID4: 8.87679 9.15556 8.80373 8.73551 8.92864 9.04323 8.99919 8.99500 9.32376 软件信任值为 8.966 信任等级为 IV

2.2.3 可视化效果

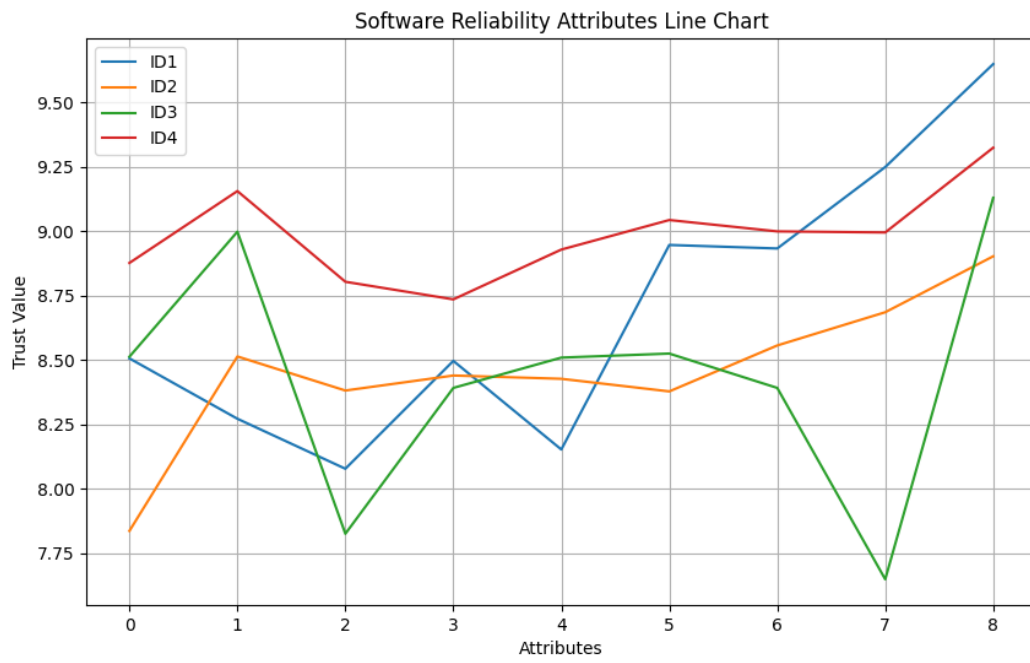


图 3: 软件可信属性折线图

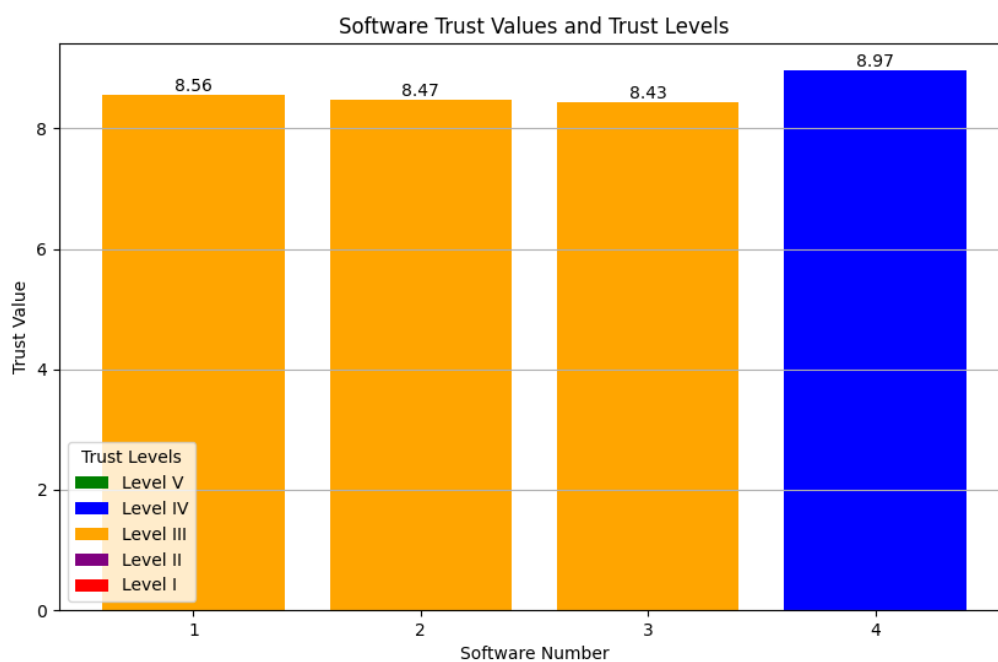


图 4: 软件可信等级柱状图

3 第三题

3.1 相关知识点

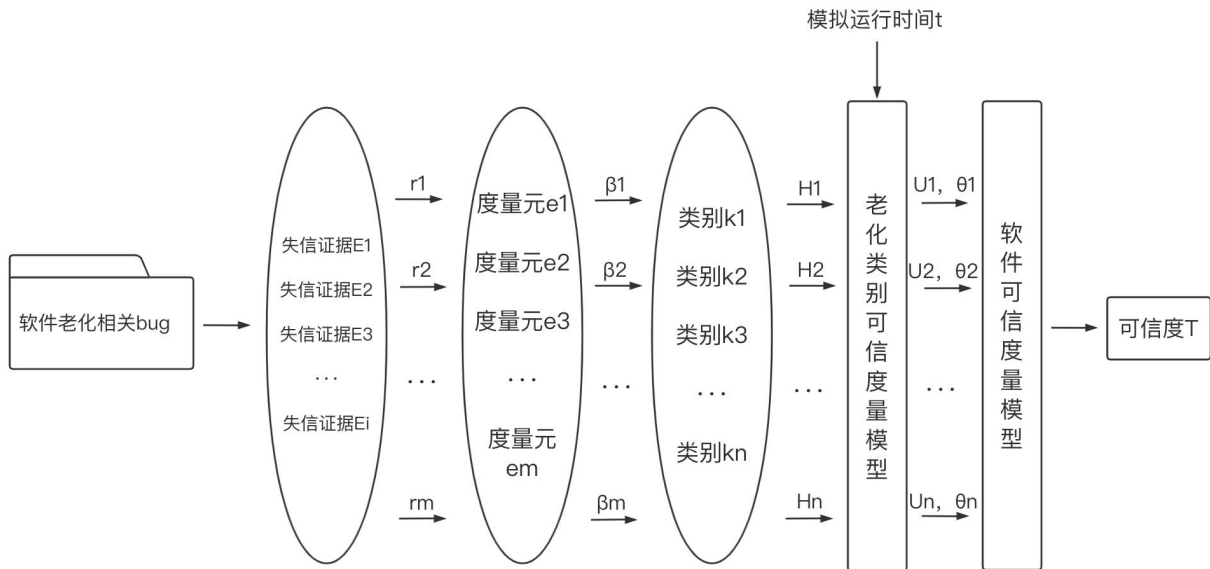


图 5: 软件老化可信计算框架

3.2 答案

3.2.1 代码运行结果

```

/opt/anaconda3/envs/test/bin/python /Users/wanghaisheng/Desktop/Coding/Courses/SoftwareQualityAnalysis/code4test/hw12-软件老化可信计算框架.py
类别数: 5
各个类别的权重: 0.539 0.125 0.238 0.049 0.049
各个类别的度量元数量: 15 4 7 2 2
第1个类别-各个度量元的权重: 0.0506 0.1845 0.0238 0.0238 0.0774 0.0774 0.0774 0.0238 0.0238 0.0506 0.0506 0.0238 0.0238 0.1042 0.1845
第1个类别-各个度量元的该时刻最大风险值: 4 4 4 1 1 1 1 1 1 4 4 4 4 4
第2个类别-各个度量元的权重: 0.25 0.25 0.25 0.25
第2个类别-各个度量元的该时刻最大风险值: 1 4 1 4
第3个类别-各个度量元的权重: 0.2340 0.1064 0.1064 0.2340 0.1064 0.1064 0.1064
第3个类别-各个度量元的该时刻最大风险值: 4 1 4 4 1 1 4
第4个类别-各个度量元的权重: 0.2 0.8
第4个类别-各个度量元的该时刻最大风险值: 1 1
第5个类别-各个度量元的权重: 0.10 0.90
第5个类别-各个度量元的该时刻最大风险值: 4 4
各类别的熵: ['0.38881', '0.30103', '0.40988', '0.00000', '0.60206']
各类别的可信值: ['6.77863', '7.40056', '6.63728', '10.00000', '5.47682']
可信值: 6.878

Process finished with exit code 0

```

图 6: $t = 0$ 运行结果

```

/opt/anaconda3/envs/test/bin/python /Users/wanghaisheng/Desktop/Coding/Courses/SoftwareQualityAnalysis/code4test/hw12-软件老化可信计算框架.py
类别数: 5
各个类别的权重: 0.539 0.125 0.238 0.049 0.049
各个类别的度量元数量: 15 4 7 2 2
第1个类别-各个度量元的权重: 0.0506 0.1845 0.0238 0.0238 0.0774 0.0774 0.0774 0.0238 0.0238 0.0506 0.0506 0.0238 0.0238 0.1042 0.1845
第1个类别-各个度量元的该时刻最大风险值: 7 4 7 7 9 7 4 7 7 7 7 7 7 7
第2个类别-各个度量元的权重: 0.25 0.25 0.25 0.25
第2个类别-各个度量元的该时刻最大风险值: 9 7 7 9
第3个类别-各个度量元的权重: 0.2340 0.1064 0.1064 0.2340 0.1064 0.1064 0.1064
第3个类别-各个度量元的该时刻最大风险值: 7 7 4 10 7 10 7
第4个类别-各个度量元的权重: 0.2 0.8
第4个类别-各个度量元的该时刻最大风险值: 4 7
第5个类别-各个度量元的权重: 0.10 0.90
第5个类别-各个度量元的该时刻最大风险值: 7 9
各类别的熵: ['0.80292', '0.89967', '0.87197', '0.79649', '0.94333']
各类别的可信值: ['4.48018', '4.06704', '4.18128', '4.50909', '3.89330']
可信值: 4.326

Process finished with exit code 0

```

图 7: $t = 10$ 运行结果

3.2.2 题目答案

$t = 0$ 时:

各类别的熵: ['0.38881', '0.30103', '0.40988', '0.00000', '0.60206'] 各类别的可信值: ['6.77863', '7.40056', '6.63728', '10.00000', '5.47682'] 可信值: 6.878

$t = 10$ 时:

各类别的熵: ['0.80292', '0.89967', '0.87197', '0.79649', '0.94333'] 各类别的可信值: ['4.48018', '4.06704', '4.18128', '4.50909', '3.89330'] 可信值: 4.326

4 附录：源代码

4.1 第一题

```

1      import numpy as np
2
3
4      # 定义正互反判断矩阵A
5      A = np.array([
6          [1, 1/2, 2, 2, 1/2, 1/4],
7          [2, 1, 1, 2, 1/2, 1/3],
8          [1/2, 1, 1, 2, 1, 1/3],
9          [1/2, 1/2, 1/2, 1, 1/2, 1/5],
10         [2, 2, 1, 2, 1, 1/2],
11         [4, 3, 3, 5, 2, 1]
12     ])
13

```



```

14     n = len(A)
15
16     # EV方法
17     def ev_method(A):
18         eigenvalues, eigenvectors = np.linalg.eig(A)
19         max_eigenvalue_index = np.argmax(eigenvalues.real)
20         principal_eigenvector = eigenvectors[:, max_eigenvalue_index].real
21         normalized_weights = principal_eigenvector / np.sum(principal_eigenvector)
22         return normalized_weights
23
24     # LLSM方法
25     def lls_method(A):
26         n = len(A)
27         product_list = []
28         for i in range(n):
29             product = 1
30             for j in range(n):
31                 if A[i][j] == 0:
32                     raise ValueError("Matrix element is zero.")
33                 product *= A[i][j]
34             product_list.append(product**(1/n))
35             sum_product = sum(product_list)
36             weights = [product / sum_product for product in product_list]
37             return list(map(float, weights))
38
39
40     def csm_method(A, epsilon=1e-10, max_iter=1000):
41         n = len(A)
42         w = np.ones(n) / n # Initial weight vector
43
44         for iteration in range(max_iter):
45             max_val, m = None, None
46
47             # Step 1: Find the index `m` with the largest discrepancy
48             for i in range(n):
49                 discrepancy = 0
50                 for j in range(n):
51                     discrepancy += ((1 + A[j, i] ** 2) * (w[i] / w[j]) - (1 + A[i, j] ** 2) * (w[j] / w[i]))
52                 discrepancy = abs(discrepancy)
53             if max_val is None or discrepancy > max_val:
54                 max_val = discrepancy
55                 m = i
56

```

```

57     # Step 2: Check for convergence
58     if max_val <= epsilon:
59         break
60
61     # Step 3: Update the weight w[m]
62     up, bottom = 0, 0
63     for j in range(n):
64         if j != m:
65             up += (1 + A[m, j] ** 2) * (w[j] / w[m])
66             bottom += (1 + A[j, m] ** 2) * (w[m] / w[j])
67
68     # Update weight vector
69     T = np.sqrt(up / bottom)
70     w[m] *= T
71     w /= np.sum(w) # Normalize weights
72
73     return w
74
75
76     # 计算每种方法的权重向量
77     W_EV = ev_method(A)
78     W_LLSM = lls_method(A)
79     W_CSM = csm_method(A)
80
81     # 打印权重向量
82     print("Weight_vector_using_EV_method:", [f"{w:.5f}" for w in W_EV])
83     print("Weight_vector_using_LLSM_method:", [f"{w:.5f}" for w in W_LLSM])
84     print("Weight_vector_using_CSM_method:", [f"{w:.5f}" for w in W_CSM])
85
86     # 计算TD (Total Deviation)
87     def total_deviation(W, A):
88         n = len(A)
89         TD = 0
90         for i in range(n):
91             for j in range(n):
92                 TD += abs(A[i][j] - W[i] / W[j])
93         return TD
94
95
96     TD_EV = total_deviation(W_EV, A)
97     TD_LLSM = total_deviation(W_LLSM, A)
98     TD_CSM = total_deviation(W_CSM, A)
99
100    # 打印总偏差 (TD)

```

```

101     print("TD_of_EV_method: {:.5f}".format(TD_EV))
102     print("TD_of_LLSM_method: {:.5f}".format(TD_LLSM))
103     print("TD_of_CSM_method: {:.5f}".format(TD_CSM))
104
105     # 最终应当选择TD值最小的方法

```

4.2 第二题

```

1     import math
2     import matplotlib.pyplot as plt
3
4
5     # 权重
6     weight = [0.05, 0.17, 0.20, 0.15, 0.09, 0.09, 0.11, 0.05, 0.09]
7
8     # 子权重
9     childWeight = [
10         0.31, 0.36, 0.33, # 指标1子权重
11         0.33, 0.33, 0.34, # 指标2子权重
12         0.16, 0.17, 0.17, 0.17, 0.17, 0.16, # 指标3子权重
13         0.33, 0.34, 0.33, # 指标4子权重
14         0.34, 0.33, 0.33, # 指标5子权重
15         0.5, 0.5, # 指标6子权重
16         0.33, 0.34, 0.33, # 指标7子权重
17         0.5, 0.5, # 指标8子权重
18         0.33, 0.33, 0.34 # 指标9子权重
19     ]
20
21
22
23     # 计算可信值
24     def calculate_trust(values, weights):
25         trust_value = 1.0
26         for i in range(len(values)):
27             trust_value *= math.pow(values[i], weights[i])
28         return trust_value
29
30     # 判断可信等级
31     def judge_trust_level_07(component_trust_values, key_component_count,
32                             overall_trust):
33         total_components = len(component_trust_values)
34         key_threshold = 3
35         low_key_count9_5 = 0

```

```
35     low_key_count8_5 = 0
36     low_key_count7_0 = 0
37     low_key_count4_5 = 0
38     has_low85 = False
39     has_low70 = False
40     has_low45 = False
41
42     # 统计关键组件低于对应阈值的数量
43     for i in range(key_component_count):
44         if component_trust_values[i] < 9.5:
45             low_key_count9_5 += 1
46         if component_trust_values[i] < 8.5:
47             low_key_count8_5 += 1
48         if component_trust_values[i] < 7.0:
49             low_key_count7_0 += 1
50         if component_trust_values[i] < 4.5:
51             low_key_count4_5 += 1
52
53     # 统计所有组件低于对应阈值的情况
54     for i in range(total_components):
55         if component_trust_values[i] < 8.5:
56             has_low85 = True
57         if component_trust_values[i] < 7.0:
58             has_low70 = True
59         if component_trust_values[i] < 4.5:
60             has_low45 = True
61
62     # 判断信任等级
63     if overall_trust >= 9.5 and low_key_count9_5 <= key_threshold and not
        has_low85:
64         return "V"
65     elif overall_trust >= 8.5 and not has_low70 and low_key_count8_5 <=
        key_threshold:
66         return "IV"
67     elif overall_trust >= 7.0 and not has_low45 and low_key_count7_0 <=
        key_threshold:
68         return "III"
69     elif overall_trust >= 4.5 and low_key_count4_5 <= key_threshold:
70         return "II"
71     else:
72         return "I"
73
74     # 绘制折线图
75     def plot_line_chart(title, x_label, y_label, data, labels):
```

```
76 plt.figure(figsize=(10, 6))
77 for i, series in enumerate(data):
78     plt.plot(series, label=labels[i])
79     plt.title(title)
80     plt.xlabel(x_label)
81     plt.ylabel(y_label)
82     plt.legend()
83     plt.grid(True)
84     plt.show()
85
86 # 绘制条形图
87 def plot_bar_chart(software_numbers, trust_values, trust_levels):
88     # 定义等级对应的颜色
89     level_colors = {
90         "V": "green",
91         "IV": "blue",
92         "III": "orange",
93         "II": "purple",
94         "I": "red"
95     }
96
97     # 获取每个软件对应的颜色
98     colors = [level_colors[level] for level in trust_levels]
99
100     plt.figure(figsize=(10, 6))
101     bars = plt.bar(software_numbers, trust_values, color=colors)
102
103     # 添加数值标签
104     for bar in bars:
105         height = bar.get_height()
106         plt.text(bar.get_x() + bar.get_width() / 2, height, f"{height:.2f}", ha='
            center', va='bottom')
107
108     plt.title("Software_Trust_Values_and_Trust_Levels")
109     plt.xlabel("Software_Number")
110     plt.ylabel("Trust_Value")
111     plt.xticks(software_numbers)
112     plt.grid(True, axis='y')
113
114     # 添加图例
115     from matplotlib.patches import Patch
116     legend_elements = [
117         Patch(facecolor='green', label='Level_V'),
118         Patch(facecolor='blue', label='Level_IV'),
```

```

119     Patch(facecolor='orange', label='Level_III'),
120     Patch(facecolor='purple', label='Level_II'),
121     Patch(facecolor='red', label='Level_I')
122 ]
123 plt.legend(handles=legend_elements, title="Trust Levels")
124
125 plt.show()
126
127 # 主函数
128 def main():
129     # 4组28个子属性平均值
130     child_trust_values_list = [
131         [9.1, 8.9, 7.6, 9.2, 7.8, 7.9, 8.9, 7.8, 7.9, 7.6, 7.5, 9.0, 9.1, 7.6, 8.9,
132          9.0, 7.9, 7.6, 9.2, 8.7, 8.9, 8.9, 9.0, 9.1, 9.4, 10, 10, 9.0],
133         [7.7, 7.9, 7.9, 9.0, 8.7, 7.9, 8.7, 8.2, 8.7, 8.2, 7.7, 8.9, 9.0, 7.7, 8.7,
134          8.7, 7.9, 8.7, 9.0, 7.8, 7.9, 8.9, 8.9, 9.2, 8.2, 8.9, 7.9, 10],
135         [7.9, 8.9, 8.7, 9.2, 8.9, 8.9, 7.9, 7.9, 8.7, 6.2, 7.9, 8.7, 8.7, 8.7, 7.8,
136          7.8, 8.9, 8.9, 9.2, 7.9, 8.7, 8.7, 7.8, 7.7, 7.6, 9.3, 9.2, 8.9],
137         [8.7, 9.2, 8.7, 9.3, 9.5, 8.7, 8.7, 9.3, 8.7, 7.7, 8.9, 9.7, 9.7, 8.7, 7.9,
138          8.7, 8.9, 9.2, 9.4, 8.7, 8.9, 9.4, 8.7, 8.7, 9.3, 9.5, 9.6, 8.9]
139     ]
140
141     # 处理每一组数据
142     attribute_trust_values_list = []
143     software_trust_values = []
144     trust_levels = []
145
146     for i in range(len(child_trust_values_list)):
147         child_trust_values = child_trust_values_list[i]
148
149         # 计算各属性信任值
150         attribute_trust_values = []
151         index = 0
152         for attr in range(len(weight)):
153             sub_trusts = []
154             sub_weights = []
155             if attr == 0:
156                 sub_trusts = [child_trust_values[index], child_trust_values[index+1],
157                             child_trust_values[index+2]]
158             index += 3
159             elif attr == 1:
160                 sub_trusts = [child_trust_values[index], child_trust_values[index+1],
161                             child_trust_values[index+2]]
162             index += 3

```

```

157     elif attr == 2:
158         sub_trusts = [child_trust_values[index], child_trust_values[index+1],
159                     child_trust_values[index+2],
160                     child_trust_values[index+3], child_trust_values[index+4], child_trust_values
161                     [index+5]]
162         index += 6
163     elif attr == 3:
164         sub_trusts = [child_trust_values[index], child_trust_values[index+1],
165                     child_trust_values[index+2]]
166         index += 3
167     elif attr == 4:
168         sub_trusts = [child_trust_values[index], child_trust_values[index+1],
169                     child_trust_values[index+2]]
170         index += 3
171     elif attr == 5:
172         sub_trusts = [child_trust_values[index], child_trust_values[index+1]]
173         index += 2
174     elif attr == 6:
175         sub_trusts = [child_trust_values[index], child_trust_values[index+1],
176                     child_trust_values[index+2]]
177         index += 3
178     elif attr == 7:
179         sub_trusts = [child_trust_values[index], child_trust_values[index+1]]
180         index += 2
181     elif attr == 8:
182         sub_trusts = [child_trust_values[index], child_trust_values[index+1],
183                     child_trust_values[index+2]]
184         index += 3
185
186     sub_weights = childWeight[index-len(sub_trusts):index]
187     attribute_trust_values.append(calculate_trust(sub_trusts, sub_weights))
188
189     attribute_trust_values_list.append(attribute_trust_values)
190
191     # 计算软件信任值
192     software_trust_value = calculate_trust(attribute_trust_values, weight)
193     software_trust_values.append(software_trust_value)
194
195     # 判断信任等级
196     trust_level = judge_trust_level_07(attribute_trust_values, len(
197         attribute_trust_values), software_trust_value)
198     trust_levels.append(trust_level)
199
200     # 输出每组的各属性信任值

```

```

194     print(f"ID{i+1}: ", end=" ")
195     for attr_trust in attribute_trust_values:
196         print(f"{attr_trust:.5f}", end=" ")
197
198     # 输出软件信任值及信任等级
199     print(f"软件信任值为{software_trust_value:.3f}信任等级为{trust_level}")
200
201     # 可视化属性信任值
202     labels = [f"ID{i+1}" for i in range(len(child_trust_values_list))]
203     plot_line_chart("SoftwareReliabilityAttributesLineChart", "Attributes",
204                     "TrustValue", attribute_trust_values_list, labels)
205
206     # 可视化软件信任值（条形图）
207     software_numbers = range(1, 1 + len(child_trust_values_list))
208     plot_bar_chart(software_numbers, software_trust_values, trust_levels)
209
210     if __name__ == "__main__":
211         main()

```

4.3 第三题

```

1     n = int(input("类别数: "))
2     theta = list(map(float, input("各个类别的权重: ").split()))
3     m = list(map(int, input("各个类别的度量元数量: ").split()))
4     R = []
5     BETA = []
6     for i in range(n):
7         beta = list(map(float, input("第{0}个类别-各个度量元的权重: ".format(i + 1))
8                                     .split()))
9         r = list(map(float, input("第{0}个类别-各个度量元的该时刻最大风险值: ".
10                                format(i + 1)).split()))
11        BETA.append(beta)
12        R.append(r)
13
14    import math
15    Hs = []
16    Us = []
17    for i in range(n):
18        H = 0
19        for j in range(m[i]):
20            H += BETA[i][j] * math.log10(R[i][j])
21        U = max(10 * math.exp(-H), 1)
22        Hs.append(H)

```



```
21     Us.append(U)
22
23     formatted_Hs = [f"{x:.5f}" for x in Hs]
24     print("各类别的熵: ", formatted_Hs)
25
26     formatted_Us = [f"{x:.5f}" for x in Us]
27     print("各类别的可信值: ", formatted_Us)
28
29     T = 1
30     for i in range(n):
31         T *= math.pow(Us[i], theta[i])
32     print("可信值: {0:.3f}".format(T))
```