# 千唐大学

## 数学建模校内竞赛论文



#### 论文题目:

组号:

成员:

选题:

姓名	学院	年级	专业	学号	联系电话	数学分析	高等代数	微积分	等 数	性代	率 统	实	数学模型	CET4	CET6
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数模校赛论文 摘 要

#### 摘 要

待补全

关键词:少样本分类;关系建模;对比学习;语义信息表示

**Key words:** 

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数模校赛论文

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## 表目录

数模校赛论文 1 绪论

#### 1 绪 论

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- 1.1 研究背景与意义
- 1.2 问题提出与研究内容

数模校赛论文 2 问题分析

## 2 问题分析

- 2.1 问题重述
- 2.2 理论基础

数模校赛论文 3 模型建立与求解

#### 3 模型建立与求解

3.1 问题 1: 颜色空间转换模型

3.2 问题 2: 四通道到五通道颜色转换模型

3.3 问题 3: LED 显示器颜色校正模型

#### 4 模型实现与结果分析

- 4.1 数据集介绍
- 4.2 实验设计
- 4.3 结果与分析

数模校赛论文 5 模型评价

## 5 模型评价

- 5.1 数据集介绍
- 5.2 实验设计
- 5.3 结果与分析

数模校赛论文 6 结论与展望

#### 6 结论与展望

- 6.1 主要结论
- 6.2 不足与改进方向

#### 参考文献

[1] Finn C, Abbeel P, Levine S. Model-agnostic meta-learning for fast adaptation of deep networks [C]//International Conference on Machine Learning. 2017: 1126-1135.

#### 附 录

#### A. 问题 1 使用代码

```
def greet(name):
    print(f"Hello, {name}!")

greet("ChatGPT")
```

#### B. 问题 2 使用代码

```
def greet(name):
    print(f"Hello, {name}!")

greet("ChatGPT")
```

#### C. 问题 3 使用代码

```
1
  import numpy as np
   import matplotlib.pyplot as plt
2
3
   import pandas as pd
   from scipy.optimize import minimize, differential_evolution
4
5
  # 设置中文字体
6
7
   plt.rcParams['font.sans-serif'] = ['SimHei', 'DejaVu Sans']
8
   plt.rcParams['axes.unicode_minus'] = False
9
10
   class LEDColorCorrection:
11
       基于三基色原理和CIE Lab色彩空间的颜色校正
12
       使用差分进化算法优化校正矩阵
13
14
15
       def ___init___(self):
16
           self.correction_matrix = None
17
18
           self.correction_bias = None
19
           self.gamma_correction = None
           self.measured\_data = None
20
           self.target\_data = None
2.1
22
23
       def load_excel_data(self, excel_path):
           """从Excel文件加载数据"""
24
25
           print(f"正在加载Excel文件: {excel_path}")
26
           sheets = ['R', 'G', 'B', 'target_R', 'target_G', 'target_B']
27
28
           data\_dict = \{\}
```

```
29
30
            for sheet_name in sheets:
                df = pd.read\_excel(excel\_path\;,\; sheet\_name=sheet\_name\;,\; header=None)
31
                    \hookrightarrow .iloc [0:64,0:64]
32
                data_dict[sheet_name] = df.values
                print(f"已加载工作表 '{sheet_name}': {df.shape}")
33
34
            #组织数据
35
36
            self.measured_data = np.stack([
37
                data_dict['R'],
38
                data_dict['G'],
                data_dict['B']
39
40
            ], axis=-1)
41
42
            self.target_data = np.stack([
43
                data_dict['target_R'],
44
                data_dict['target_G'],
45
                data_dict['target_B']
            ], axis=-1)
46
47
            print(f"测量数据形状: {self.measured_data.shape}")
48
            print(f"目标数据形状: {self.target_data.shape}")
49
50
51
        def rgb_to_xyz(self, rgb):
            """RGB转XYZ色彩空间"""
52
53
            rgb\_norm = rgb / 255.0
54
55
            # Gamma校正
56
            rgb\_linear = np.where(rgb\_norm \le 0.04045,
57
                                  rgb_norm / 12.92,
58
                                  np.power((rgb\_norm + 0.055) / 1.055, 2.4))
59
            # sRGB到XYZ的转换矩阵
60
61
            transform_matrix = np.array([
                [0.4124564, 0.3575761, 0.1804375],
62
                [0.2126729, 0.7151522, 0.0721750],
63
                [0.0193339\,,\ 0.1191920\,,\ 0.9503041]
64
65
            ])
66
            xyz = np.dot(rgb_linear, transform_matrix.T)
67
68
            return xyz
69
70
        def xyz_to_lab(self, xyz):
            """XYZ转CIE Lab色彩空间"""
71
72
            # D65 白点
            Xn, Yn, Zn = 0.95047, 1.00000, 1.08883
73
74
            x = xyz [..., 0] / Xn
75
```

```
y = xyz [\dots, 1] / Yn
76
77
              z = xyz[..., 2] / Zn
78
79
              # 立方根变换
              fx = np.where(x > 0.008856, np.power(x, 1/3), (7.787 * x + 16/116))
80
81
              fy = np.where(y > 0.008856, np.power(y, 1/3), (7.787 * y + 16/116))
              fz \, = \, np.\,where \, (\,z \, > \, 0.008856 \, , \, \, np.\,power \, (\,z \, , \, \, 1/3) \, , \, \, (7.787 \, \ ^* \, z \, + \, 16/116) \, )
82
83
84
              L = 116 * fy - 16
85
              a = 500 * (fx - fy)
86
              b = 200 * (fy - fz)
87
              return np. stack ([L, a, b], axis=-1)
88
89
90
         def calculate_color_difference(self, lab1, lab2):
              """计算CIE Delta E 2000色差"""
91
92
              L1, a1, b1 = lab1 [..., 0], lab1 [..., 1], lab1 [..., 2]
93
              L2, a2, b2 = lab2[..., 0], lab2[..., 1], lab2[..., 2]
94
              C1 = np. sqrt (a1**2 + b1**2)
95
              C2 = np. sqrt (a2**2 + b2**2)
96
              C_{bar} = 0.5 * (C1 + C2)
97
98
99
              G = 0.5 * (1 - np.sqrt(C_bar^{**7} / (C_bar^{**7} + 25^{**7})))
100
              a1p = (1 + G) * a1
              a2p = (1 + G) * a2
101
102
              C1p = np. sqrt (a1p**2 + b1**2)
103
104
              C2p = np. sqrt (a2p**2 + b2**2)
105
              h1p = np.degrees(np.arctan2(b1, a1p)) \% 360
106
              h2p = np.degrees(np.arctan2(b2, a2p)) \% 360
107
108
109
              dLp = L2 - L1
              dCp = C2p - C1p
110
111
              dhp = h2p - h1p
112
113
              dhp = dhp - 360 * (dhp > 180) + 360 * (dhp < -180)
              dHp\,=\,2\ *\ np.\,sqrt\left(C1p\ *\ C2p\right)\ *\ np.\,sin\left(np.\,radians\left(dhp\ /\ 2\right)\right)
114
115
              L_bar = 0.5 * (L1 + L2)
116
117
              C_{bar_p} = 0.5 * (C1p + C2p)
118
              h_bar_p = (h1p + h2p + 360 * (np.abs(h1p - h2p) > 180)) / 2
119
120
              h_bar_p %= 360
121
122
              T = (1
123
                   - 0.17 * np.cos(np.radians(h_bar_p - 30))
```

```
124
                 + 0.24 * np.cos(np.radians(2 * h_bar_p))
125
                + 0.32 * np.cos(np.radians(3 * h_bar_p + 6))
                 - 0.20 * np.cos(np.radians(4 * h_bar_p - 63)))
126
127
128
             Sl = 1 + (0.015 * (L_bar - 50)**2) / np. sqrt(20 + (L_bar - 50)**2)
             Sc = 1 + 0.045 * C_bar_p
129
            Sh = 1 + 0.015 * C_bar_p * T
130
131
132
             delta\_theta = 30 * np.exp(-((h\_bar\_p - 275)/25)**2)
            Rc = 2 * np. sqrt(C_bar_p**7 / (C_bar_p**7 + 25**7))
133
134
            Rt = -np.sin(np.radians(2 * delta\_theta)) * Rc
135
136
            dE = np.sqrt
                 (dLp / S1)**2 +
137
138
                 (dCp / Sc)**2 +
139
                 (dHp / Sh)**2 +
140
                 Rt * (dCp / Sc) * (dHp / Sh)
141
             )
142
143
             return dE
144
145
        def estimate_gamma_parameters(self):
             """估计LED的Gamma参数 (保留线性比例偏移)"""
146
147
             print ("正在估计Gamma参数...")
148
            gamma\_params = \{\}
             for i, channel in enumerate (['R', 'G', 'B']):
149
150
                 meas = self.measured_data[..., i].flatten() / 255.0
151
                 targ = self.target_data[..., i].flatten() / 255.0
152
                 mask = (targ > 0.04) \& (targ < 0.96) \& (meas > 0)
153
                 m = meas[mask]
                 t = targ[mask]
154
                 if len(m) > 0:
155
                     # 拟合 \log(m) = \text{gamma} * \log(t) + \text{offset}
156
157
                     A = np.vstack([np.log(t + 1e-8), np.ones_like(t)]).T
158
                     gamma, offset = np.linalg.lstsq(A, np.log(m + 1e-8), rcond=
                         \hookrightarrow None) [0]
                     gamma = float(np.clip(gamma, 0.1, 3.0))
159
160
                     scale = float(np.exp(offset))
161
                 else:
                     gamma, scale = 1.0, 1.0
162
                 gamma_params[channel] = { 'gamma': gamma, 'scale': scale}
163
                 print(f"{channel}通道 Gamma: {gamma:.3f}, Scale: {scale:.3f}")
164
165
             self.gamma_correction = gamma_params
166
             return gamma_params
167
168
        def apply_gamma_correction(self, rgb_data, inverse=False):
             """应用Gamma校正: 在归一化 [0,1] 空间先应用线性比例, 再做幂运算"""
169
170
             if self.gamma_correction is None:
```

```
171
                 return rgb_data
172
             data = rgb_data.astype(np.float32) / 255.0
             out = np.zeros_like(data)
173
174
             for i, channel in enumerate(['R', 'G', 'B']):
                 gamma = self.gamma_correction[channel]['gamma']
175
176
                 scale = self.gamma_correction[channel]['scale']
177
                 ch = data[..., i]
178
                 if not inverse:
                     #前向: 先比例, 再幂
179
                     tmp = ch * scale
180
181
                     tmp = np.clip(tmp, 0.0, 1.0)
182
                     out_ch = np.power(tmp, gamma)
183
                 else:
184
                     # 反向: 开幂, 再去比例
185
                     tmp = np.power(ch, 1.0 / gamma)
186
                     out_ch = tmp / np.maximum(scale, 1e-8)
187
                 \operatorname{out}[\ldots, i] = \operatorname{np.clip}(\operatorname{out\_ch}, 0.0, 1.0)
188
            # 恢复到 [0,255]
             return (out * 255.0).astype(rgb_data.dtype)
189
190
191
        def correction_function(self, params, measured_lin, target_lin):
192
             优化函数:线性校正矩阵 M 和偏置 b, params 长度 12。
193
194
             corrected = clip (M @ measured + b, [0,1])
             计算 \Delta E + 正则化。
195
196
197
            M = params[:9]. reshape(3,3)
198
             b = params [9:]. reshape (1,3)
199
200
            # 应用矩阵和偏置
201
             corr = np.dot(measured_lin, M.T) + b
202
             corr = np. clip(corr, 0.0, 1.0)
203
204
            # 转到 XYZ → Lab
205
             transform = np.array([[0.4124564, 0.3575761, 0.1804375],
206
                                    [0.2126729, 0.7151522, 0.0721750],
207
                                    [0.0193339, 0.1191920, 0.9503041]])
208
             tgt_xyz = np.dot(target_lin, transform.T)
             corr_xyz = np.dot(corr, transform.T)
209
             tgt_lab = self.xyz_to_lab(tgt_xyz.reshape(-1,3)).reshape(corr.shape)
210
             corr\_lab = self.xyz\_to\_lab(corr\_xyz.reshape(-1,3)).reshape(corr.shape)
211
212
            #色差
213
214
             deltaE = self.calculate_color_difference(tgt_lab, corr_lab)
215
             loss = np.mean(deltaE)
216
217
            #矩阵正则+偏置正则
218
             loss += 0.001 * (np.sum((M - np.eye(3))**2) + np.sum(b**2))
```

```
219
             det = np. linalg. det (M)
220
             if \det \le 0 or abs(\det) < 0.1:
221
                 loss += 1000.0
222
             return loss
223
224
225
         def calibrate_correction_matrix(self):
             print("开始校正: 矩阵 + 偏置...")
226
227
             self.estimate_gamma_parameters()
228
             # 预处理: 线性化
229
             meas = self.apply_gamma_correction(self.measured_data.astype(np.
                 \hookrightarrow float 32), inverse=True)/255.0
             targ = self.apply_gamma_correction(self.target_data.astype(np.float32)
230
                \hookrightarrow , inverse=True)/255.0
231
             meas flat = meas.reshape(-1,3)
232
             targ_flat = targ.reshape(-1,3)
233
             #差分进化优化 12 参数
234
             bounds = [(-2,2)]*9 + [(-0.1,0.1)]*3
235
             res = differential_evolution(
236
                 self.correction_function, bounds,
237
                 {\tt args=(meas\_flat\,,\ targ\_flat\,)\,,\ maxiter=200,\ popsize=15,\ seed=42}
238
239
             x0 = res.x
240
             # 局部 L-BFGS-B
241
             local = minimize(
242
                 self.correction_function, x0, args=(meas_flat,targ_flat),
243
                 method='L-BFGS-B', options={'maxiter':500}
244
245
             M_{opt} = local.x[:9].reshape(3,3)
246
             b_{opt} = local.x[9:].reshape(3)
247
             self.correction_matrix = M_opt
             self.correction\_bias = b\_opt
248
             print("校正完成; 矩阵行列式: ", np.linalg.det(M_opt))
249
250
             print("偏置: ", b_opt)
251
             return M_opt, b_opt
252
253
254
         def apply_correction(self, input_rgb):
255
             """应用带偏置的线性校正"""
256
             lin = self.apply_gamma_correction(input_rgb.astype(np.float32),
                 \hookrightarrow inverse=True)/255.0
257
             flat = lin.reshape(-1,3)
258
             corr = np.dot(flat, self.correction_matrix.T) + self.correction_bias
259
             corr = np.clip(corr, 0.0, 1.0).reshape(input_rgb.shape)
             out = (corr * 255.0). astype(np.float32)
260
261
             final = self.apply_gamma_correction(out, inverse=False)
262
             return final.astype(np.uint8)
263
```

```
264
        def evaluate_correction(self):
            """评估校正效果"""
265
266
            corrected = self.apply_correction(self.measured_data.astype(np.float32
                \hookrightarrow ))
267
268
            measured_xyz = self.rgb_to_xyz(self.measured_data.astype(np.float32))
269
            corrected_xyz = self.rgb_to_xyz(corrected.astype(np.float32))
270
            target_xyz = self.rgb_to_xyz(self.target_data.astype(np.float32))
271
272
            measured_lab = self.xyz_to_lab(measured_xyz)
273
            corrected_lab = self.xyz_to_lab(corrected_xyz)
274
            target_lab = self.xyz_to_lab(target_xyz)
275
276
            diff_before = self.calculate_color_difference(measured_lab, target_lab
277
            diff_after = self.calculate_color_difference(corrected_lab, target_lab
                \hookrightarrow )
278
279
            print("="*50)
            print ("校正效果评估报告")
280
            print("="*50)
281
            print(f"校正前平均色差: {np.mean(diff_before):.3f}")
282
            print(f"校正后平均色差: {np.mean(diff_after):.3f}")
283
284
            print(f"色差改善: {np.mean(diff_before) - np.mean(diff_after):.3f}")
285
            print(f"改善百分比: {((np.mean(diff_before) - np.mean(diff_after)) /
                \hookrightarrow np.mean(diff_before) * 100):.1 f\\%")
            print(f"校正前最大色差: {np.max(diff_before):.3f}")
286
            print(f"校正后最大色差: {np.max(diff_after):.3f}")
287
288
            print(f"色差<1.0的像素比例:校正前{np.mean(diff_before < 1.0)*100:.1f
                → }%, 校正后{np.mean(diff_after < 1.0)*100:.1f}%")
289
            print("="*50)
290
291
            return corrected, diff_before, diff_after
292
293
        def visualize_results(self):
            """可视化校正结果"""
294
295
            corrected_data = self.apply_correction(self.measured_data.astype(np.
                \hookrightarrow float 32))
296
297
            fig, axes = plt.subplots(3, 4, figsize=(20, 15))
298
            # 第一行: 测量数据
299
            for i, (channel, color) in enumerate(zip(['R', 'G', 'B'], ['Reds', '
300

    Greens', 'Blues'])):
                im = axes[0, i].imshow(self.measured_data[:, :, i], cmap=color,
301
                    \rightarrow vmin=0, vmax=255)
                 axes[0, i].set_title(f'测量值 - {channel} 通道')
302
303
                axes[0, i].axis('off')
```

```
304
                  plt.colorbar(im, ax=axes[0, i], fraction=0.046, pad=0.04)
305
              measured_rgb = np.clip(self.measured_data / 255.0, 0, 1)
306
307
              axes[0, 3].imshow(measured_rgb)
              axes[0, 3].set_title('测量值 - RGB合成')
308
              axes[0, 3].axis('off')
309
310
             # 第二行: 目标数据
311
312
              for i, (channel, color) in enumerate(zip(['R', 'G', 'B'], ['Reds', '

    Greens', 'Blues'])):
313
                  im = axes[1, i].imshow(self.target_data[:, :, i], cmap=color, vmin
                      \rightarrow =0, vmax=255)
                  axes[1, i].set_title(f'目标值 - {channel} 通道')
314
315
                  axes[1, i].axis('off')
316
                  plt.colorbar(im, ax=axes[1, i], fraction=0.046, pad=0.04)
317
318
              target_rgb = np.clip(self.target_data / 255.0, 0, 1)
319
              axes[1, 3].imshow(target_rgb)
              axes[1, 3].set_title('目标值 - RGB合成')
320
321
              axes[1, 3].axis('off')
322
             # 第三行: 校正后数据
323
324
              for i, (channel, color) in enumerate(zip(['R', 'G', 'B'], ['Reds', '

    Greens', 'Blues'])):
325
                  im = axes[2, i].imshow(corrected_data[:, :, i], cmap=color, vmin
                      \hookrightarrow =0, vmax=255)
                  axes[2, i].set_title(f'校正后 - {channel} 通道')
326
                  axes[2, i].axis('off')
327
328
                  plt.colorbar(im, ax=axes[2, i], fraction=0.046, pad=0.04)
329
330
              corrected_rgb = np.clip(corrected_data / 255.0, 0, 1)
331
              axes [2, 3]. imshow(corrected_rgb)
              axes[2, 3].set_title('校正后 - RGB合成')
332
333
              axes[2, 3].axis('off')
334
335
              plt.tight_layout()
336
              plt.show()
337
338
    # 主函数
339
     i \ f \ \underline{\quad} name\underline{\quad} = \ "\underline{\quad} main\underline{\quad} ":
340
         files = ["MathModel\_Code \setminus data \setminus preprocess \setminus p3 \setminus RedPicture.xlsx", "
341
             \hookrightarrow MathModel_Code\\data\\preprocess\\p3\\GreenPicture.xlsx", "
             \hookrightarrow MathModel_Code\\data\\preprocess\\p3\\BluePicture.xlsx"]
342
343
         corrector = LEDColorCorrection()
344
345
         for filepath in files:
```

```
corrector.load_excel_data(filepath)
346
347
             correction_matrix = corrector.calibrate_correction_matrix()
348
349
             print ("\n评估校正效果:")
350
             corrected_display , diff_before , diff_after = corrector.
                \hookrightarrow evaluate_correction()
351
352
             corrector.visualize_results()
353
354
             print("\n校正完成!")
```

#### D. 像素数据集