



Education and Culture

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European Master In Computational Logic

Practical Project

**Implementation of An Atomized
Methodology for Evaluating Credibility
Values Concerning Similarity Relations
in Fuzzy Logic**

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Chapter 1

Practical Application

In [1] we introduce our own methodology for evaluating the similarity degree of fuzzy predicates. The method lacked a feature for realizing an error tolerance when working on knowledge bases with incomplete information. For that reason, three approaches are proposed in [1]. The most elegant of these approaches, namely the *Hybrid Approach* inherits the best features of the preceding ones, while avoiding the shortcomings via merging the two by attaining them some weights. Previously we displayed how these weights could be decided in an intuitionistic way, and hinted the possibility of an atomized procedure. In this chapter we show our proposal for atomizing the process with a real-world example.

1.1 The Problem

In the introduction chapter of [1], we mention about how today's search engines have limited querying capabilities. We give the example of the query *red car*, and say that an ideal frame work would return us the results of the query *orange car* in addition to the original one, with lower credibility values. In the light of this, as our real-world practical example, we inspect the similarity relations between the colors. We may observe colors as a domain of interest in many distinct fields, however one particular interesting example is the set of canonical colors that the web browsers use, namely the *X11 colors* [4]. The system makes use of the *RGB* framework, where colors are specified as triplets. Every color is depicted by three values which represent the *Red*, *Green* and *Blue* amount that the color consists of. The complete *X11* table is shown in the following page.

HTML name	Hex code R G B	Decimal code R G B	HTML name	Hex code R G B	Decimal code R G B	HTML name	Hex code R G B	Decimal code R G B
Red colors			Green colors			Brown colors		
IndianRed	CD 5C 5C	205 92 92	GreenYellow	AD FF 2F	173 255 47	Cornsilk	FF F8 DC	255 248 220
LightCoral	F0 80 80	240 128 128	Chartreuse	7F FF 00	127 255 0	BlanchedAlmond	FF EB CD	255 235 205
Salmon	FA 80 72	250 128 114	LawnGreen	7C FC 00	124 252 0	Bisque	FF E4 C4	255 228 196
DarkSalmon	E9 96 7A	233 150 122	Lime	00 FF 00	0 255 0	NavajoWhite	FF DE AD	255 222 173
LightSalmon	FF A0 7A	255 160 122	LimeGreen	32 CD 32	50 205 50	Wheat	F5 DE B3	245 222 179
Red	FF 00 00	255 0 0	PaleGreen	98 FB 98	152 251 152	BurlyWood	DE B8 87	222 184 135
Crimson	DC 14 3C	220 20 60	LightGreen	90 EE 90	144 238 144	Tan	D2 B4 8C	210 180 140
FireBrick	B2 22 22	178 34 34	MediumSpringGreen	00 FA 9A	0 250 154	RosyBrown	BC 8F 8F	188 143 143
DarkRed	8B 00 00	139 0 0	SpringGreen	00 FF 7F	0 255 127	SandyBrown	F4 A4 60	244 164 96
Pink colors			MediumSeaGreen	3C B3 71	60 179 113	Goldenrod	DA A5 20	218 165 32
Pink	FF C0 CB	255 192 203	SeaGreen	2E 8B 57	46 139 87	DarkGoldenrod	B8 86 0B	184 134 11
LightPink	FF B6 C1	255 182 193	ForestGreen	22 8B 22	34 139 34	Peru	CD 85 3F	205 133 63
HotPink	FF 69 B4	255 105 180	Green	00 80 00	0 128 0	Chocolate	D2 69 1E	210 105 30
DeepPink	FF 14 93	255 20 147	DarkGreen	00 64 00	0 100 0	SaddleBrown	8B 45 13	139 69 19
MediumVioletRed	C7 15 85	199 21 133	YellowGreen	9A CD 32	154 205 50	Sienna	A0 52 2D	160 82 45
PaleVioletRed	DB 70 93	219 112 147	OliveDrab	6B 8E 23	107 142 35	Brown	A5 2A 2A	165 42 42
Orange colors			Olive	80 80 00	128 128 0	Maroon	80 00 00	128 0 0
LightSalmon	FF A0 7A	255 160 122	DarkOliveGreen	55 6B 2F	85 107 47	White colors		
Coral	FF 7F 50	255 127 80	MediumAquamarine	66 CD AA	102 205 170	White	FF FF FF	255 255 255
Tomato	FF 63 47	255 99 71	DarkSeaGreen	8F BC 8F	143 188 143	Snow	FF FA FA	255 250 250
OrangeRed	FF 45 00	255 69 0	LightSeaGreen	20 B2 AA	32 178 170	Honeydew	F0 FF F0	240 255 240
DarkOrange	FF 8C 00	255 140 0	DarkCyan	00 8B 8B	0 139 139	MintCream	F5 FF FA	245 255 250
Orange	FF A5 00	255 165 0	Teal	00 80 80	0 128 128	Azure	F0 FF FF	240 255 255
Yellow colors			Blue/Cyan colors			AliceBlue	F0 F8 FF	240 248 255
Gold	FF D7 00	255 215 0	Aqua	00 FF FF	0 255 255	GhostWhite	F8 F8 FF	248 248 255
Yellow	FF FF 00	255 255 0	Cyan	00 FF FF	0 255 255	WhiteSmoke	F5 F5 F5	245 245 245
LightYellow	FF FF E0	255 255 224	LightCyan	E0 FF FF	224 255 255	Seashell	FF F5 EE	255 245 238
LemonChiffon	FF FA CD	255 250 205	PaleTurquoise	AF EE EE	175 238 238	Beige	F5 F5 DC	245 245 220
LightGoldenrodYellow	FA FA D2	250 250 210	Aquamarine	7F FF D4	127 255 212	OldLace	FD F5 E6	253 245 230
PapayaWhip	FF EF D5	255 239 213	Turquoise	40 E0 D0	64 224 208	FloralWhite	FF FA F0	255 250 240
Moccasin	FF E4 B5	255 228 181	MediumTurquoise	48 D1 CC	72 209 204	Ivory	FF FF F0	255 255 240
PeachPuff	FF DA B9	255 218 185	DarkTurquoise	00 CE D1	0 206 209	AntiqueWhite	FA EB D7	250 235 215
PaleGoldenrod	EE E8 AA	238 232 170	CadetBlue	5F 9E A0	95 158 160	Linen	FA F0 E6	250 240 230
Khaki	F0 E6 8C	240 230 140	SteelBlue	46 82 B4	70 130 180	LavenderBlush	FF F0 F5	255 240 245
DarkKhaki	BD B7 6B	189 183 107	LightSteelBlue	B0 C4 DE	176 196 222	MistyRose	FF E4 E1	255 228 225
Purple colors			PowderBlue	B0 E0 E6	176 224 230	Gray colors		
Lavender	E6 E6 FA	230 230 250	LightBlue	AD D8 E6	173 216 230	Gainsboro	DC DC DC	220 220 220
Thistle	D8 BF D8	216 191 216	SkyBlue	87 CE EB	135 206 235	LightGrey	D3 D3 D3	211 211 211
Plum	DD A0 DD	221 160 221	LightSkyBlue	87 CE FA	135 206 250	Silver	C0 C0 C0	192 192 192
Violet	EE 82 EE	238 130 238	DeepSkyBlue	00 BF FF	0 191 255	DarkGray	A9 A9 A9	169 169 169
Orchid	DA 70 D6	218 112 214	DodgerBlue	1E 90 FF	30 144 255	Gray	80 80 80	128 128 128
Fuchsia	FF 00 FF	255 0 255	CornflowerBlue	64 95 ED	100 149 237	DimGray	69 69 69	105 105 105
Magenta	FF 00 FF	255 0 255	RoyalBlue	41 69 E1	65 105 225	LightSlateGray	77 88 99	119 136 153
MediumOrchid	BA 55 D3	186 85 211	Blue	00 00 FF	0 0 255	SlateGray	70 80 90	112 128 144
MediumPurple	93 70 DB	147 112 219	MediumBlue	00 00 CD	0 0 205	DarkSlateGray	2F 4F 4F	47 79 79
BlueViolet	8A 2B E2	138 43 226	DarkBlue	00 00 8B	0 0 139	Black	00 00 00	0 0 0
DarkViolet	94 00 D3	148 0 211	Navy	00 00 80	0 0 128			
DarkOrchid	99 32 CC	153 50 204	MidnightBlue	19 19 70	25 25 112			
DarkMagenta	8B 00 8B	139 0 139						
Purple	80 00 80	128 0 128						
Indigo	4B 00 82	75 0 130						
DarkSlateBlue	48 3D 8B	72 61 139						
SlateBlue	6A 5A CD	106 90 205						
MediumSlateBlue	7B 68 EE	123 104 238						

1.1.1 Preliminaries

Before starting to proceed with the problem itself, we ought to mention about some problem specific concepts.

1.1.1.1 Real World Similarity

As stated in section 1.1 in the *RGB* scheme, all colors are represented by triplets of *Red*, *Green* and *Blue* values. With this in mind, when observing the similarity value between two colors, directly from the real-world data, we may consider them as if they were placed in *3D space*. Thus each of the *RGB* values acts as a coordinate for the corresponding color, and we may utilize a geometric distance formulation for this problem. In this scenario we adopt the *Euclidean distance*, which is indeed one of the common definitions in the field of *color science*. [3]

In Euclidean three-space, the distance between points (x_1, y_1, z_1) and (x_2, y_2, z_2) is

$$d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2 + (z_2 - z_1)^2}$$

Thus in our problem, the similarity distance between two colors C and S follows as:

$$similarity_distance = \sqrt{(C_{Red} - S_{Red})^2 + (C_{Green} - S_{Green})^2 + (C_{Blue} - S_{Blue})^2}$$

We then follow by normalizing this metric distance value with the following algorithm

$$degree_of_similarity = 1 - \frac{similarity_distance}{\sqrt{3 * 255^2}}$$

so that the resulting values fall between the *real number* interval of $[0, 1]$.

As one will notice the denominator of the function is the highest possible distance in the metric space. Real world correspondence of this scenario consists the color pair of *Black* and *White*. The algorithm would return 0 as a result for this case. Moreover as expected, the pairing of any color with itself would result the algorithm computing 1 as the similarity degree.

1.1.1.2 Evaluated Similarity

The gist of our methodology for evaluating the similarity degree between two concepts in [1] is conserved in the practical case with minor modifications.

The predicate trees are relatively simple in structural terms. All are depth one, have three children.

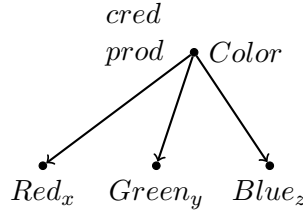


FIGURE 1.1: Example of a color predicate tree

As we know main focus of our algorithm is utilizing predefined similarity relations between the subconcepts. In this particular example, we assume the only predefined similarity relations are the ones between the same RGB main colors. So for every Red_α and Red_β a similarity relations is defined where α and β are two integer values from the interval $[0, 255]$.

The corresponding function is as follows:

$$similarity_single_color = 1 - \frac{|c_1 - c_2|}{255}$$

which informally equals to the normalization of the metric distance on a single coordinate of the colors, so that it falls into the interval of $[0, 1]$.

Lastly, the evaluation function follows the same procedure described in [1] .

1.1.1.3 Pseudo Credibility Values

The main focus of interest is finding the weight constants in the *Hybrid Approach(HA)* in an automatic sense. As one might observe, the problem consists the necessary information for calculating the values of VA and EA thus for fixing the weights, solely the credibility value should also be known. So for training weights, there is a need for some initial *pseudo credibility values*.

In the end we want the product of our similarity and credibility estimation, close to the real degree of similarity.

$$degree_of_similarity \times degree_of_credibility \cong real_similarity$$

Through the preceding sections, we know evaluating both of values concerning similarities. So by putting these values in the equation, we may compute the variable of interest.

1.1.1.4 Computing Weights

In [1], we introduce the formula for the *Hybrid Approach* as follows:

$$Credibility = (w \times [\mathbf{Vertex_Approach}]) + ((1 - w) \times [\mathbf{Edge_Approach}]) \quad (1.1)$$

where $w \in [0..1]$.

As mentioned earlier, when one has the value of credibility, getting the value of the weights from the formula is trivial.

One noteworthy concept here is the *Global Weight*. The focus of this methodology is training the weights from some pre-given info, then fixing the weights through some procedure so the credibility values for new fuzzy rules can be determined.

For maintaining the Multi-Adjoint properties as shown in [2], we adopt the minimum operator for evaluating the global weight. Informally, we train a single weight for every similarity pair in our training set, then set the minimum of those values as our global weight.

1.1.1.5 The Credibility

When the weights are computed, the credibility values are easily evaluated via the *Hybrid Approach* evaluation function. This consists the last step of the methodology, which could just be followed by checking both the credibility value and the similarity value together in comparison to the real degree of similarity.

Here the desired outcome is getting a conservative results, so to speak a value which does not exceed the original degree of similarity, but still finding one which is relatively close.

1.2 A Practical Example

Suppose we have the following six color pairings in our training set:

Salmon Coral
 Tomato Peru
 YellowGreen MediumAquamarine
 FineBrick LightSeaGreen
 Crimson RoyalBlue
 Snow MidnightBlue

LISTING 1.1: The training set

For the sake of an interesting display, let us select a pairing from the training set and one completely outside of the initial knowledge base, in order to construct our test set:

Tomato Peru
Pink Violet

LISTING 1.2: The training set

With these information in hand, we are ready for pursuing with the calculations.

1.2.1 Training Process for Salmon and Coral

In this section we show step by step the process of training a weight via utilizing the info of a color pairing from the training set. The task is simply following the explicit process described in section 1.1.1.

The first step is finding the real-world similarity proximity between the colors, namely *Salmon* and *Coral*. And for that, one needs to find out the metric distance of similarity of the colors.

We input the corresponding *RGB* values of the colors into the modified euclidian function, defined in section 1.1.1.1.

$$\begin{aligned}
 similarity_distance &= \sqrt{(C_{Red} - S_{Red})^2 + (C_{Green} - S_{Green})^2 + (C_{Blue} - S_{Blue})^2} \\
 \\
 similarity_distance &= \sqrt{(C_{Red} - S_{Red})^2 + (C_{Green} - S_{Green})^2 + (C_{Blue} - S_{Blue})^2} \\
 &= \sqrt{(C_{250} - S_{255})^2 + (C_{128} - S_{127})^2 + (C_{114} - S_{80})^2} \\
 &\cong \mathbf{34.38}
 \end{aligned}
 \tag{1.2}$$

Then use this metric distance value in the similarity evaluation function:

$$\begin{aligned}
degree_of_similarity &= 1 - \frac{similarity_distance}{\sqrt{3 * 255^2}} \\
&= 1 - \frac{34.48}{\sqrt{3 * 255^2}} \\
&\cong \mathbf{92.21}
\end{aligned} \tag{1.3}$$

Second step of the process is finding the degree of similarity between the colors via using our own evaluation algorithm. As we know first thing to do here is handling with the calculation of the similarity proximities of subconcepts. There will be three such relations that have to be taken into account. Let us compute one of them, namely the one concerning the *Red* value of the colors:

$$\begin{aligned}
similarity_single_color &= 1 - \frac{|c_1 - c_2|}{255} \\
&= 1 - \frac{|250 - 255|}{255} \\
&\cong \mathbf{0.98}
\end{aligned} \tag{1.4}$$

After doing the same computation for the subconcept similarity relations regarding *Green* and *Blue*, and putting them in the evaluation function, we get:

$$degree_of_similarity = 0.947$$

In section 1.1.1.3 we have stated that one we have the both results of the similarity evaluations, we may conclude the credibility value which will be used for training the weights. The corresponding formula and the computation is as follows:

$$\begin{aligned}
degree_of_similarity \times degree_of_credibility &= real_similarity \\
degree_of_credibility &= \frac{real_similarity}{degree_of_similarity} \\
&= 1 - \frac{0.947}{0.98} \\
&\cong \mathbf{0.973}
\end{aligned} \tag{1.5}$$

The last computation enables us to continue with the last step of training, which is evaluating the weight. In addition to the credibility value, the *VA* and *EA* values should also be calculated. Since all of the subconcepts are included in some similarity relation,

the *VA* value is 1. Moreover as there are three predefined subconcept similarity relations out of nine possible relations(i.e. $3 \times 3 = 9$), the corresponding *EA* value is $1/3$.

With the help of all these information, the so-called *trained weight value* is:

$$\begin{aligned}
 \text{Credibility} &= (w \times [\text{Vertex_Approach}]) + ((1 - w) \times [\text{Edge_Approach}]) \\
 w &= \frac{\text{Credibility} - [\text{Edge_Approach}]}{[\text{Vertex_Approach}] - [\text{Edge_Approach}]} \\
 &= \frac{0.973 - 0.33}{1 - 0.33} \\
 &\cong \mathbf{0.959}
 \end{aligned} \tag{1.6}$$

And this concludes our calculation for the first color pair of the training set.

1.2.2 The Outcome

A complete implementation of the process in *C++* programming language is displayed at the appendix.

Regarding this example, the console output of the program is depicted in the following page.

As stated in section 1.1.1.4, in order to preserve *Multi-Adjoint* properties, we take the *minumum* operation as the merging tool of the weight values from the training set.

By multiplying the credibility and the similarity values evaluated from our algorithm, we compare the resulting number with the real similarity degree.

As can be seen in both examples, particularly both in the case of the example from the training set and also the newly introduced color pair, the results prove to be conservative. In other words they do not exceed the real values, which is a desired feature. In addition to that, we observe that the evaluated values are relatively close to the real-word correspondences.

As both of these features are satisfied by the result of our methodology where a relatively small knowledge base is utilized, we consider our method to be displaying extreme promise for real world applications.

```

Please select your choice of program: a
~~~~~
~~~~~TRAINING SET~~~~~
~~~~~
Color_pair[0], Salmon and Coral:
sim1:  0.947712      sim2:  0.922159
cred2: 0.973037      weight: 0.959555

Color_pair[1], Tomato and Peru:
sim1:  0.879739      sim2:  0.8615
cred2: 0.979268      weight: 0.968902

Color_pair[2], YellowGreen and MediumAquaMarine:
sim1:  0.775163      sim2:  0.703893
cred2: 0.908058      weight: 0.862087

Color_pair[3], FireBrick and LightSeaGreen:
sim1:  0.443137      sim2:  0.44288
cred2: 0.999418      weight: 0.999128

Color_pair[4], Crimson and RoyalBlue:
sim1:  0.470588      sim2:  0.4525
cred2: 0.961562      weight: 0.942343

Color_pair[5], Snow and MidnightBlue:
sim1:  0.224837      sim2:  0.207335
cred2: 0.92216       weight: 0.88324

*****
Global weight: 0.862087
*****

~~~~~
~~~~~TEST SET~~~~~
~~~~~
Color_pair[6], Tomato and Peru:
sim1:  0.879739      sim2:  0.8615
cred1: 0.908058
simFin: 0.798854

*****
0.798854    =<    0.8615
*****

Color_pair[7], Pink and Violet:
sim1:  0.85098       sim2:  0.83427
cred1: 0.908058
simFin: 0.77274

*****
0.77274    =<    0.83427
*****

```

Chapter 2

Implementation

Here we display the implementation of the methodology that is discussed in detail in chapter 1.

Firstly we should briefly mention about the flow of the program and the structure of the input files.

The main program contains of two subprograms. The first one of those takes two input files, namely *colorNames.txt* and *colors.txt*. The program takes a user-defined number of color pairs. The last two pairs are the test set, and the rest of the pairs consist the training set. One of the two pairings of the test set is expected to be from the training set, and the other one outside of the domain of the training set. So the program trains the weights via the color pairings from the training set and evaluate the credibility value of the one from the test set. For this regard, *colorNames.txt* store the information concerning the names of the colors. Every line displays one pair, and in each pair colors are separated by the *space* escape character. In a similar manner, in each line *colors.txt* saves the *RGB* values of the corresponding colors from the other input file. The six values, i.e. three for each of the colors, are separated by *space* characters and lines are split with an *enter key*.

An example case for the input files is displayed as follows:

```
250 128 114 255 127 80
255 99 71 210 105 30
154 205 50 102 205 170
178 34 34 32 178 170
220 20 60 65 105 225
255 250 250 25 25 112
255 99 71 210 105 30
255 192 203 238 130 238
```

LISTING 2.1: colors.txt

```

Salmon Coral
Tomato Peru
YellowGreen MediumAquamarine
FineBrick LightSeaGreen
Crimson RoyalBlue
Snow MidnightBlue
Tomato Peru
Pink Violet

```

LISTING 2.2: colorNames.txt

The second subprogram works with a single input file, namely *colorsAll.txt*. The input file contains all the information regarding the complete set of colors in *X11* scheme. At each line you have the name and the *RGB* values of a particular color.

The subprogram asks the user to name two colors from the domain. It then takes out those two values and their associated information from the knowledge base. All of the possible pairings between the rest of the colors are computed by the program, which exactly corresponds to 9591 pairings. With this huge amount of information, the weights are trained and later utilized for evaluating the credibility degree of the similarity value concerning the color pairing of interest.

You may observe an example instance of the *colorsAll.txt* input file. That is followed by two example outputs of the program for the same color pair, but with different methodologies of evaluating the global weight, namely *minimum rule* and *average rule*. And finally you may inspect the complete implementation of the methodology in *C++* programming language.

```

IndianRed 205 92 92
LightCoral 240 128 128
Salmon 250 128 114
DarkSalmon 233 150 122
LightSalmon 255 160 122
Red 255 0 0
Crimson 220 20 60
FireBrick 178 34 34
DarkRed 139 0 0
Pink 255 192 203
LightPink 255 182 193
HotPink 255 105 180
DeepPink 255 20 147
MediumVioletRed 199 21 133
PaleVioletRed 219 112 147
LightSalmon 255 160 122
Coral 255 127 80

```

Tomato	255	99	71
OrangeRed	255	69	0
DarkOrange	255	140	0
Orange	255	165	0
Gold	255	215	0
Yellow	255	255	0
LightYellow	255	255	224
LemonChiffon	255	250	205
LightGoldenrodYellow	250	250	210
PapayaWhip	255	239	213
Moccasin	255	228	181
PeachPuff	255	218	185
PaleGoldenrod	238	232	170
Khaki	240	230	140
DarkKhaki	189	183	107
Lavender	230	230	250
Thistle	216	191	216
Plum	221	160	221
Violet	238	130	238
Orchid	218	112	214
Fuchsia	255	0	255
Magenta	255	0	255
MediumOrchid	186	85	211
MediumPurple	147	112	219
BlueViolet	138	43	226
DarkViolet	148	0	211
DarkOrchid	153	50	204
DarkMagenta	139	0	139
Purple	128	0	128
Indigo	75	0	130
DarkSlateBlue	72	61	139
SlateBlue	106	90	205
MediumSlateBlue	123	104	238
GreenYellow	173	255	47
Chartreuse	127	255	0
LawnGreen	124	252	0
Lime	0	255	0
LimeGreen	50	205	50
PaleGreen	152	251	152
LightGreen	144	238	144
MediumSpringGreen	0	250	154
SpringGreen	0	255	127
MediumSeaGreen	60	179	113
SeaGreen	46	139	87
ForestGreen	34	139	34
Green	0	128	0
DarkGreen	0	100	0
YellowGreen	154	205	50
OliveDrab	107	142	35
Olive	128	128	0
DarkOliveGreen	85	107	47

MediumAquamarine 102 205 170
DarkSeaGreen 143 188 143
LightSeaGreen 32 178 170
DarkCyan 0 139 139
Teal 0 128 128
Aqua 0 255 255
Cyan 0 255 255
LightCyan 224 255 255
PaleTurquoise 175 238 238
Aquamarine 127 255 212
Turquoise 64 224 208
MediumTurquoise 72 209 204
DarkTurquoise 0 206 209
CadetBlue 95 158 160
SteelBlue 70 130 180
LightSteelBlue 176 196 222
PowderBlue 176 224 230
LightBlue 173 216 230
SkyBlue 135 206 235
LightSkyBlue 135 206 250
DeepSkyBlue 0 191 255
DodgerBlue 30 144 255
CornflowerBlue 100 149 237
RoyalBlue 65 105 225
Blue 0 0 255
MediumBlue 0 0 205
DarkBlue 0 0 139
Navy 0 0 128
MidnightBlue 25 25 112
Cornsilk 255 248 220
BlanchedAlmond 255 235 205
Bisque 255 228 196
NavajoWhite 255 222 173
Wheat 245 222 179
BurlyWood 222 184 135
Tan 210 180 140
RosyBrown 188 143 143
SandyBrown 244 164 96
Goldenrod 218 165 32
DarkGoldenrod 184 134 11
Peru 205 133 63
Chocolate 210 105 30
SaddleBrown 139 69 19
Sienna 160 82 45
Brown 165 42 42
Maroon 128 0 0
White 255 255 255
Snow 255 250 250
Honeydew 240 255 240
MintCream 245 255 250
Azure 240 255 255

```
AliceBlue 240 248 255
GhostWhite 248 248 255
WhiteSmoke 245 245 245
Seashell 255 245 238
Beige 245 245 220
OldLace 253 245 230
FloralWhite 255 250 240
Ivory 255 255 240
AntiqueWhite 250 235 215
Linen 250 240 230
LavenderBlush 255 240 245
MistyRose 255 228 225
Gainsboro 220 220 220
LightGrey 211 211 211
Silver 192 192 192
DarkGray 169 169 169
Gray 128 128 128
DimGray 105 105 105
LightSlateGray 119 136 153
SlateGray 112 128 144
DarkSlateGray 47 79 79
Black 0 0 0
```

LISTING 2.3: colorsAll.txt

```
Color_pair[9584], DarkGray and LightSlateGray:
```

```
sim1: 0.870588      sim2: 0.859606
cred2: 0.987529      weight: 0.981293
```

```
Color_pair[9585], DarkGray and SlateGray:
```

```
sim1: 0.839216      sim2: 0.831251
cred2: 0.990621      weight: 0.985932
```

```
Color_pair[9586], Gray and DimGray:
```

```
sim1: 0.909804      sim2: 0.909804
cred2: 1            weight: 1
```

```
Color_pair[9587], Gray and LightSlateGray:
```

```
sim1: 0.945098      sim2: 0.937173
cred2: 0.991703      weight: 0.987554
```

```
Color_pair[9588], Gray and SlateGray:
```

```
sim1: 0.95817      sim2: 0.948769
cred2: 0.99029      weight: 0.985435
```

```
Color_pair[9589], DimGray and LightSlateGray:
```

```
sim1: 0.878431      sim2: 0.866801
cred2: 0.986909      weight: 0.980364
```

```
Color_pair[9590], DimGray and SlateGray:
```

```
sim1: 0.909804      sim2: 0.89627
cred2: 0.985286      weight: 0.977929
```

```
Color_pair[9591], LightSlateGray and SlateGray:
```

```
sim1: 0.968627      sim2: 0.968464
cred2: 0.999833      weight: 0.99975
```

```
*****
Global weight: 0.345403
*****
```

```
~~~~~
TEST SET ~~~~~
```

```
Color_pair[focus], Pink and Violet:
```

```
sim1: 0.85098      sim2: 0.83427
cred1: 0.563602
simFin: 0.479614
```

```
*****
0.479614 =< 0.83427
*****
```

```

sim1: 0.74902 sim2: 0.74902
cred2: 1 weight: 1

Color_pair[9584], DarkGray and LightSlateGray:
sim1: 0.870588 sim2: 0.859606
cred2: 0.987529 weight: 0.981293

Color_pair[9585], DarkGray and SlateGray:
sim1: 0.839216 sim2: 0.831251
cred2: 0.990621 weight: 0.985932

Color_pair[9586], Gray and DimGray:
sim1: 0.909804 sim2: 0.909804
cred2: 1 weight: 1

Color_pair[9587], Gray and LightSlateGray:
sim1: 0.945098 sim2: 0.937173
cred2: 0.991703 weight: 0.987554

Color_pair[9588], Gray and SlateGray:
sim1: 0.95817 sim2: 0.948769
cred2: 0.99029 weight: 0.985435

Color_pair[9589], DimGray and LightSlateGray:
sim1: 0.878431 sim2: 0.866801
cred2: 0.986909 weight: 0.980364

Color_pair[9590], DimGray and SlateGray:
sim1: 0.909804 sim2: 0.89627
cred2: 0.985286 weight: 0.977929

Color_pair[9591], LightSlateGray and SlateGray:
sim1: 0.968627 sim2: 0.968464
cred2: 0.999833 weight: 0.99975

*****
Global weight: 0.922583
*****

TEST SET
Color_pair[focus], Pink and Violet:
sim1: 0.85098 sim2: 0.83427
cred1: 0.948389
simFin: 0.80706

*****
0.80706 =< 0.83427
*****

```

```

#include<iostream>
using namespace std;
#include<string>
#include<math.h>
#include <vector>
#include <string>
#include <fstream>
#include <sstream>

struct color{
    int x;           //first coordinate
    int y;           //second coordinate
    int z;           //third coordinate
    string c;        //the color name
};

float euqDist(float , float , float , float , float , float);           //
    euclidian distance between two colors.
float simOneColor(float , float);
    //similarity between two instances of
    the same color.
float simByED(float);
    //similarity degree of two
    colors via euclidian distance.
float simByAlg(float , float , float , float , float , float);           //
    similarity degree of two colors via our algorithm.
float credBySim(float , float);
    //credibility value defined from proximity of
    similarity values.
float weightEval(float , float= 1, float = (1/3.0));
    //weight w, evaluated from cred function.
float credEval(float , float= 1, float = (1/3.0));
    //final credibility value, computed via global weight
void inputHandle(ifstream& f, vector<vector<float>> &c, string fileName);
void inputColor(ifstream& f, vector<string> &col, string fileName);
void inputComplete(ifstream& f, vector<color> &c, string fileName);

int main()
{
    string pc;
    begin:
    cout<<"Please select your choice of program: ";
    cin>>pc;

    if(pc == "1" || pc == "A" || pc == "a"){
        ifstream f1, f2;
        vector<vector<float>> colorPairs;
        inputHandle(f1, colorPairs, "colors.txt");
        vector<string> colorNames;
    }
}

```

```

        inputColor(f2, colorNames, "colorNames.txt");

        cout<<"\n
        -----\n";
        cout<<"-----TRAINING SET
        -----\n";
        cout<<"
        -----\n";

        float weightMin = 1;
        for(int i = 0; i < colorPairs.size() - 2; i++){
            float sim1 = simByAlg(colorPairs[i][0],
colorPairs[i][1], colorPairs[i][2], colorPairs[i][3], colorPairs[i][4],
colorPairs[i][5]);
            float sim2 = simByED(euqDist(colorPairs[i
][0], colorPairs[i][1], colorPairs[i][2], colorPairs[i][3], colorPairs[i
][4], colorPairs[i][5]));
            float cred2 = credBySim(sim1, sim2);
            float weight = weightEval(cred2);
            if(weight < weightMin)
                weightMin = weight;

            cout<<"Color_pair["<<i<<"], "<<colorNames[i
*2]<<" and "<<colorNames[i*2+1]<<": "<<endl;
            cout<<"
            -----<<endl;
            cout<<"sim1:\t"<<sim1<<"\tsim2:\t"<<sim2<<"
\ncred2:\t"<<cred2<<" \tweight:\t"<<weight<<"\n\n\n";
        }

        cout<<"*****\n";
        cout<<"\tGlobal weight: "<<weightMin<<endl;
        cout<<"*****\n\n
\n";

        cout<<"
        -----\n";
        cout<<"-----TEST SET
        -----\n";
        cout<<"
        -----\n";

        int s = colorPairs.size();
        for(int t = s - 2; t < s; t++){
            float sim1 = simByAlg(colorPairs[t][0], colorPairs[t
][1], colorPairs[t][2], colorPairs[t][3], colorPairs[t][4], colorPairs
[t][5]);
            float sim2 = simByED(euqDist(colorPairs[t][0],
colorPairs[t][1], colorPairs[t][2], colorPairs[t][3], colorPairs[t][4],
colorPairs[t][5]));

```

```

        cout<<" Color_pair ["<<t<<"], "<<colorNames[t*2]<<"
and "<<colorNames[t*2+1]<<": "<<endl;
        cout<<" -----"<<
endl;

        cout<<"sim1:\t"<<sim1<<" \tsim2:\t"<<sim2<<"\n\n";

        float cred1 = credEval(weightMin);           //global
weight is used for finding the final credibility value
        cout<<"cred1:\t"<<cred1<<endl;
        float simFin = cred1 * sim1;
        cout<<"simFin:\t"<<simFin;

        cout<<"\n\n
*****\n";
        cout<<"\t"<<simFin<<"      =<      "<<sim2<<endl;
        cout<<"
*****\n\n\n";
    }
}

else if(pc == "2" || pc == "B" || pc == "b"){
    ifstream f3;
    vector<color> col;
    inputComplete(f3, col, "colorsAll.txt");
    string userColor1, userColor2;

    cout<<"\nPlease choose two colors: ";
    cin>>userColor1;
    cin>>userColor2;

    color temp1, temp2;
    for(int i = 0; i < col.size(); i++){
        if(col[i].c == userColor1){
            temp1.c = col[i].c;
//store the color at a temp variable
            temp1.x = col[i].x;
            temp1.y = col[i].y;
            temp1.z = col[i].z;

            col[i].c = col[col.size()-1].c; //switch it
with the last element of
            col[i].x = col[col.size()-1].x; //the
vector in order to pop it out
            col[i].y = col[col.size()-1].y; //of the
list
            col[i].z = col[col.size()-1].z;
            col.pop_back();
        }
    }

    for(int i = 0; i < col.size(); i++){

```

```

        if (col[i].c == userColor2){
            temp2.c = col[i].c;
            //store the color at a temp variable
            temp2.x = col[i].x;
            temp2.y = col[i].y;
            temp2.z = col[i].z;

            col[i].c = col[col.size()-1].c; //switch it
            //with the last element of
            col[i].x = col[col.size()-1].x; //the
            //vector in order to pop it out
            col[i].y = col[col.size()-1].y; //of the
            //list
            col[i].z = col[col.size()-1].z;
            col.pop_back();
        }
    }

    cout<<"\n Weights evaluated via Min or Average rule: ";
    cin>>pc;

    float weightGlo = 1.0; //global weight
    if (pc == "1" || pc == "m" || pc == "M"){
        color t1, t2;
        int pn = 1; //
        //pair counter

        for (int i = 0; i < col.size(); i++){
            for (int j = i+1; j < col.size(); j++){
                float sim1 = simByAlg(col[i].x, col[i].y, col[i].z, col[j].x, col[j].y, col[j].z);
                float sim2 = simByED(euqDist(col[i].x, col[i].y, col[i].z, col[j].x, col[j].y, col[j].z));
                float cred2 = credBySim(sim1, sim2);

                float weight = weightEval(cred2);
                if (weight < weightGlo)
                    weightGlo = weight;

                cout<<" Color-pair ["<<pn<<"], "<<col[i].c<<" and "<<col[j].c<<": "<<endl;
                cout<<"
                -----"<<endl;
                cout<<" sim1:\t"<<sim1<<"\tsim2:\t"
                <<sim2<<"\ncred2:\t"<<cred2<<" \tweight:\t"<<weight<<"\n\n";
                pn++;
            }
        }

        else if (pc == "a" || pc == "A" || pc == "2"){

```



```

        color t1, t2;
        vector<float> weights;
        int pn =1; //
pair counter
        for(int i = 0; i < col.size(); i++){
            for(int j = i+1; j < col.size(); j++){
                float sim1 = simByAlg(col[i].x, col
[i].y, col[i].z, col[j].x, col[j].y, col[j].z);
                float sim2 = simByED(euqDist(col[i
].x, col[i].y, col[i].z, col[j].x, col[j].y, col[j].z));
                float cred2 = credBySim(sim1, sim2)
;

                float weight = weightEval(cred2);
                weights.push_back(weight);

                cout<<" Color_pair["<<pn<<"], "<<col
[i].c<<" and "<<col[j].c<<": "<<endl;
                cout<<"
-----"<<endl;
                cout<<" sim1:\t"<<sim1<<"\tsim2:\t"
<<sim2<<"\ncred2:\t"<<cred2<<" \tweight:\t"<<weight<<"\n\n";
                pn++;
            }
        }
        for(int i = 0; i < col.size(); i++)
            weightGlo += weights[i];
        weightGlo /= col.size();

        cout<<"*****\n";
        cout<<"\tGlobal weight: "<<weightGlo<<endl;
        cout<<"*****\n\n";

        cout<<"
-----\n";
        cout<<"-----TEST SET
-----\n";
        cout<<"
-----\n";

        float sim1 = simByAlg(temp1.x, temp1.y, temp1.z, temp2.x,
temp2.y, temp2.z);
        float sim2 = simByED(euqDist(temp1.x, temp1.y, temp1.z,
temp2.x, temp2.y, temp2.z));
        cout<<" Color_pair[ focus ], "<<temp1.c<<" and "<<temp2.c<<" :
"<<endl;
        cout<<"-----"<<endl;
        cout<<" sim1:\t"<<sim1<<" \tsim2:\t"<<sim2<<"\n\n";

```

```

        float cred1 = credEval(weightGlo);           //global weight is
        used for finding the final credibility value
        cout<<"cred1:\t"<<cred1<<endl;
        float simFin = cred1 * sim1;
        cout<<"simFin:\t"<<simFin;

        cout<<"\n\n
        *****\n";
        cout<<"\t"<<simFin<<"      =<      "<<sim2<<endl;
        cout<<"*****\n\n
\n";
    }

    else{
        cout<<"Wrong command!\n";
        goto begin;
    }

    getchar();
    getchar();
    return 0;
}

//euclidian distance between two colors
float euqDist(float a, float b, float c, float x, float y, float z)
{
    return sqrt(pow(fabs(a-x),2)+pow(fabs(b-y),2)+pow(fabs(c-z),2));
}

//similarity between two instances of the same color. (ex: red_90 and
red_187)
float simOneColor(float c1, float c2)
{
    return 1-((fabs(c1-c2))/255);
}

//similarity degree of two colors via euqlidian distance
float simByED(float dist)
{
    return 1-(dist/(255*sqrt(3.0)));
    /*
    float res = 1-(dist/(255*sqrt(3.0)));
    if(res < 0.1)
        return 0;
    else
        return res;*/
}

//similarity degree of two colors via our algorithm
float simByAlg(float a, float b, float c, float x, float y, float z)
{

```

```

        return (((simOneColor(a, x))+(simOneColor(b, y))+(simOneColor(c, z)
    ))/3);
}

//credibility value defined from proximity of similarity values
float credBySim(float sim1, float sim2)
{
    //return 1-(fabs(sim1 - sim2));
    float sim = (sim2+0.01) / (sim1+0.01); //in order to prevent
    division by zero
    if(sim > 1)
        return 1;
    else
        return sim;
}

//weight w, evaluated from cred function
float weightEval(float cred, float va, float ea)
    //default values for va and ea are 1 and 1/3 respectively
{
    //since
    those are the most common cases, but could be overwritten
    return ((cred-ea)/(va-ea));
}

float credEval(float weight, float va, float ea)
    //again default values for va and ea are 1 and 1/3 respectively
{
    return ((weight * va) + ((1-weight) * ea));
}

void inputHandle(istream& f, vector<vector<float>> &c, string fileName)
{
    string temp;
    f.open(fileName);

    if (!f) {
        cout << "Unable to open file";
        cin>>temp;
        exit(1); // terminate with error
    }

    int i = 0;
    char line[1024];

    while(f.getline(line, 1024)){
        istringstream ss(line, istringstream::in);
        float rate;
        vector<float> temp;

```

```

        c.push_back(temp);

        for(int j = 0; j < 6; j++){
            ss >> rate;
            c[i].push_back(rate);
        }
        i++;
    }

    f.close();
}

void inputColor(ifstream& f, vector<string> &c, string fileName)
{
    string temp;
    f.open(fileName);

    if (!f) {
        cout << "Unable to open file";
        cin>>temp;
        exit(1); // terminate with error
    }

    char line[1024];
    string tempColor;

    while(f.getline(line, 1024)){
        istringstream ss(line, istringstream::in);
        ss >> tempColor; //first
        color of the pairing
        c.push_back(tempColor);
        ss >> tempColor; //second
        color of the pairing
        c.push_back(tempColor);
    }

    f.close();
}

void inputComplete(ifstream& f, vector<color> &c, string fileName)
{
    string temp;
    f.open(fileName);

    if (!f) {
        cout << "Unable to open file";
        cin>>temp;
        exit(1); // terminate with error
    }

```

```
char line[1024];

while(f.getline(line, 1024)){
    istringstream ss(line, istringstream::in);
    color temp;

    ss >> temp.c;
    ss >> temp.x;
    ss >> temp.y;
    ss >> temp.z;

    c.push_back(temp);

    cout<<temp.x<<" \t"<<temp.y<<" \t"<<temp.z<<" \t"<<temp.c<<
endl;
}

f.close();
}
```

LISTING 2.4: The C++ implementation of the application concerning colors

Bibliography

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