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#### Abstract

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# A Method for Exploring Word-Colour Associations 

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#### Abstract

Strong associations exist between colours and concepts or words. Understanding these associations, sometimes referred to as colour emotions, is important for effective use of colour in art and design. Traditionally the relationships have been systematically explored in experiments where participants scale colours according to bi-polar adjectives such as warm-cool. In this paper a method for exploring the relationships between words and colours is suggested and is demonstrated. A psychophysical experiment is described where participants select colours based on words. The data are used to show that many similarities between the word-colour relationships for UK and Chinese participants although some interesting differences are also revealed. The method makes explicit the observation that there is not a one-to-one relationship between words and colours. The method could be used to explore word-colour relationship for specific words and participant groups or could be used to generate ground-truth data for testing methods for automatically generating the word-colour relationships.


## 1 Introduction

It is established that strong associations exist between colours and concepts ${ }^{1}$. Colour can also elicit an emotional response and the term colour emotion is often used to describe this ${ }^{2}$. Colours or colour combinations may evoke corresponding feelings such as excitement, energy and calmness ${ }^{2-5}$. It is likely that such colours emotions are triggered by learned associations ${ }^{6}$ or shaped by nature although the effect of light on the non-imaging-forming pathways of the human visual systems as a contributing cause cannot be ruled out ${ }^{7}$. It has been noted that people associate blue with calming, depressing, peaceful, quiet, serious, and nostalgic ${ }^{1}$; associate yellow with serene, happy and softly exciting ${ }^{8}$, or warm and sunny; associate green with envy, red with passion, black with death, yellow with cowardice, blue with loyalty ${ }^{9}$. Cherry has explored the link between colour and mood ${ }^{5}$. In terms of object colour at least, cool colours are linked with the moods as calm, serene and comfort whereas warm colours are relevant to stressful and exciting moods.

Colour is differentiated by hue, brightness and saturation ${ }^{10}$. In terms of hue, colours can be categorized into warm colours (e.g. orange and red) and cool colours (e.g. violet, blue). Brightness and saturation are also significant in colour perception however. Brightness plays an important role in determining lightness or darkness of colour; saturation suggests purity of colour. According to Hemphill, bright colours are linked with positive emotions like happiness, joy and hope ${ }^{11}$. Likewise, Elliot and Maier consider brighter colours as friendly, cultured, pleasant and beautiful ${ }^{12}$. Conversely, dark colours are associated with negative emotions such as boredom and sadness ${ }^{13}$. According to Elliot and Maier, colour may generate associations and responses, and they take meaning of colour as bipartite ${ }^{12}$.

In terms of colour preference there is a huge amount of previous literature. The long history of colour preference research has been described as being 'bewildering, confused and contradictory ${ }^{14-15}$. Most recent studies have tended to agree that, on average, people have a preference for cool shades such as blue and a dislike for warmer colours such as yellow and orange but there is still huge individual variation and some systematic differences between gender and culture ${ }^{15-16}$. For example, Choungourian studied people from four countries and identified variance in colour preferences representing individual variation ${ }^{17}$. It was found that the US consumers prefer red and blue, but least like bluegreen. However, respondents from Iran and Kuwait preferred blue and green. However,
whether there are effects of culture and gender on colour preference, and certainly exactly what those effects are, is still somewhat of an open question.

Although colour can elicit emotional responses, the sorts of associations that colours have are not limited to emotions. For example, the colour pink may be associated with femininity, but this does not necessarily mean that looking at pink (or wearing pink) makes one 'feel' more feminine (although it may); it may be that pink is simply associated (cognitively) with the concept of femininity. Such associations are important for the successful application of colour in product design, advertising and marketing ${ }^{18}$. In some case, "meaning" may represent a kind of mental stimulation ${ }^{19}$. Won defined colour meaning in her research thus ${ }^{20}$ : "Colour meaning is not about combinations that create pleasing responses (colour harmony), not about the processes with which people understand and react to colour (colour perception), and not about liking a particular colour among alternatives (colour preference). Instead, it is concerned with the meanings that are associated with certain colours." Grieve has argued that colours do not elicit a semiotic response per se; ${ }^{21}$ however, this can be interpreted as stating that without the cultural framework in which we live, red, for example, would have no specific meaning and would elicit no semiotic response. Nevertheless, we do live in a cultural framework and empirical studies show that colours do elicit response even when presented without any context ${ }^{22}$. These two views can be reconciled if we understand that colours can have a meaning even when presented without a context because of the associations that become formed when individuals previously encounter that colour in context. People associate colours with experience and memories; a colour may remind someone of a specific object, a certain substance, a person, a period time or a region and this has been referred to as colour association ${ }^{23}$.

Some studies have discussed colour meanings amongst different cultures and some crosscultural difference in colour meanings and associations have been identified. The colours themselves may differ because of dependence on lighting conditions, observation position and surrounding especially the adjacent colour and all of these factors can shape the ways of perceiving a specific colour ${ }^{19}$. Moreover, even when people are exposed to the same stimulus colour, the way they perceive that colour, and hence the meanings and emotions incurred, may differ as a result of variability in gender, age, educational background and culture, childhood association and others ${ }^{24}$. Nevertheless, broadly speaking, the empirical data suggest more similarities than differences in colour meanings between cultural
groups. Osgood and his colleagues carried out colour-meaning research across 20 countries ${ }^{19}$. They chose participants from high schools and asked them to rate seven (red, orange, yellow, green, blue, black and white) colours for each of 12 semantic differential items. The results, analysed using scaling, used "evaluation, potency and activity" as reference, were broadly similar for the 20 countries; blue was the most highly evaluated colour, green and white followed; black and red were the most potent colours; the most active colour was red, whereas grey and black were the most passive colours. In 2000, research indicated some differences and similarities in four cultures (Japan, China, South Korea, and United States $)^{25}$. The students from those four countries were asked to summarize the most closely colour associated with 13 words which are often used in describing objects from eight sample colours. As a result, red was associated with love and blue with high quality for all four cultures. Black was consistently associated with powerful and expensive. In contrast, purple associations showed a sharp contrast between three Asian countries (Japan, China and South Korea) and the United States. In the three Asian countries, purple was associated with expensive products while in the United States purple normally represents inexpensive. Thomas explored cross-cultural similarities and dissimilarities in colour preferences and colour meanings associations in eight cultures (Austria, Brazil, Canada, Colombia, Hong Kong, China, Taiwan and the United State) ${ }^{25}$. In his results, green, white and blue were consistently associated with "calming", "peaceful" and "gentle" in all eight countries. Green, white and blue were also associated with "pleasant" (Austria, Colombia, United States, and to a lesser extent China and Taiwan) and "beautiful" (Brazil, Hong Kong, China, and United States). However, those three colours also represent unique meanings separately in different countries while they share the meanings in the research. Thomas concluded that no universal meanings can be attributed to specific colours and proposed, instead, a "Spectrum of Colour Meaning".

Colour association may be a more accurate term than colour meaning since it implies that the relationship between colour and concept is the result of a learned association rather than being a property of the colours themselves. Colour association can be divided into concrete and abstract colour associations. Schloss and Palmer argued that colour can be associated with specific objects ${ }^{26}$. For instance, blue is associated with sky, red colour is associated with ripe berries and fruits, and brown is associated with faeces and rotten fruit. Moreover, the differences in colour associations between age, areas and environment have been discussed. In Goethe's research ${ }^{8}$, it is suggested that colour associations could be different because of some factors (for instance, women and children, or northern

Europeans as opposed to southern Europeans) and hue might not even be the most important element in colour associations ${ }^{23}$. Gage agrees that 'the same colour can be found to have quite antithetical connotations in different periods and cultures and even at the same time and place ${ }^{1}$. Tofle et al. also claimed emotional responses evoked by colour as an outcome of learned associations on basis of culture and individual-related features ${ }^{6}$.

In summary, despite Grieve's claim ${ }^{23}$ that colours elicit no semiotic response per se, empirical data supports the notion that colours are associated with ideas, concepts and emotions. Although there is some variation between individuals and between cultural and gender groups, there is a great deal that is agreed upon. These associations may result from human physiological responses or, more likely, from associations with key objects and experiences but nevertheless are relatively robust. Colours may, however, take on quite different meanings and associations when they occur in context (that is, a red dress may take on quite a different meaning to the colour red without any context). Note that this work is restricted to the associations of colour in an abstract sense rather than with any particular instantiation or context. By context we mean that the colours are not seen as belonging to an object but rather as seen as simple patches (albeit in the context of a background and other colour patches).

The vast majority of experimental work that has been carried out in this field has used a scaling technique called semantic differential scaling ${ }^{19}$. In these studies the participants are presented with a colour and are asked to respond using a slider bar or by otherwise indicating the extent to which the colour is associated with two bi-polar adjectives; for example, male-female, soft-hard, old-new. Some computational models have been produced, fitted to the experimental data, that are able to predict the extent to which any given colour can elicit any of these responses ${ }^{22}$. However, in this work we suggest that there is merit in turning this problem around on its head. That is, rather than starting with a colour and asking to what extent it is, for example, modern, we start with an adjective, such as modern, and ask which colours are associated with this adjective. The reason for this is that in the design process the identification of intended feelings and/or ideas will normally come first, and the designer then seeks colours that can deliver the required message ${ }^{20}$. However, there are obviously a very large number of potential words that could be used as starting points for a colour palette; it is inconceivable that laboratorybased studies such as the one described in this paper could provide the required data for this (especially since this might be required in multiple languages). The purpose of this
study therefore is to provide some new psychophysical data on the relationship between word and colour. These data could be used as ground-truth data to validate automatic methods that may be developed to predict colours from words ${ }^{29,30}$. In one previously published study, for example, it was suggested that internet scraping could be used to derive the word-colour relationships automatically based on analyses of millions of images ${ }^{29}$. The second purpose of the work is to illustrate the method used in this study to obtain word-colour relationships from laboratory studies and to encourage other researchers to adopt these methods. Finally, we conduct a study with two distinct cultural groups. The aim of the work is not to undertake a cultural comparison (since that would require many more words and more participants than used in the current study). Nevertheless, the comparison of the results of this method used with two cultural groups will enhance the extent to which the method can be evaluated.

## 2 Experimental

A total of 30 participants, 15 from the UK ( 7 males and 8 females) and 15 from China ( 9 males and 6 females) all of whom had normal colour vision and were more than 18 years old, were recruited to take part in a psychophysical experiment. The number of males and females in the two groups were approximately equal.

There were two purposes to the experiment:
(i) to demonstrate a new method for collecting word-colour associations;
(ii) to produce psychophysical data that could be used to evaluate the performance of algorithms that automatically generate colour palettes.

In the experiment, 30 target English words were selected. Although to some extent the selection of the 30 words was arbitrary, the words chosen were amongst the most frequent words in the English language. A number of websites exist that provide these lists (for example, http://www.wordfrequency.info/.

However, words such as 'the' and ' $a$ ', although frequently occurring, would probably have very weak colour associations. Therefore, 30 words were selected (see Table 1) from a list of the 500 most-frequent words; in selecting the words some attention was paid to whether the words would likely have strong colour associations - in addition, some words were selected because they have been used extensively in other related studies (for example, the word 'active'). Chinese translations were made for each of the 30 words
（Table 1）．Since typeface may affect the mood or emotion created in the reader，Lanting Hei（兰亭黑）was chosen for the Chinese typeface，since it invokes quite similar responses to Calibri in English from different cultures ${ }^{27}$ ．

Table 1：The list of 30 words used in the experiment（note，that before this，each participant was also presented with the words，red，soft and twelve）．

| English Words | Chinese Words | English Words | Chinese Words |
| :---: | :---: | :---: | :---: |
| active | 积极的 | married | 婚姻的 |
| bad | 坏的 | medical | 医学的 |
| clean | 干净的 | modern | 现代的 |
| cold | 冷的 | natural | 自然的 |
| cultural | 文化的 | old | 年老的 |
| dangerous | 危险的 | poor | 贫穷的 |
| dead | 死亡的 | rewerful | 强大的 |
| female | 女性的 | rich | 宗教的 |
| fresh | 新鲜的 | safe | 富有的 |
| future | 未来的 | sweet | 安全的 |
| good | 好的 | traditional | 甜的 |
| healthy | unlucky | 传统的 |  |
| hot | 健康的 | urban | 不幸的 |
| lucky | 热的 | young | 城市的 |
| male | 幸运的 |  | 年轻的 |
|  | 男性的 |  |  |

In the experiment the participants viewed the words（presented one at a time in randomised order）on a PC display（HP DreamColor LP2480zx－a 24－inch LCD Backlit Monitor，max luminance $187 \mathrm{~cd} / \mathrm{m}^{2}$ ）from a distance of about 80 cm in a darkened room． The UK participants were presented with the English words and the Chinese participants were presented with the Chinese（Mandarin）words．A small reward was given to each participant at the end of the experiment to thank them for their participation．

The words were displayed in a black typeface on a uniform grey（CIELAB L＊＝50） background．For each word that was presented（in random order），each participant was requested to perform the following tasks：

1. To select the colour that most represents the word using a typical colour picker.
2. To indicate the strength of association between that colour and the word using a slider bar (in the range $0-100$ ).
3. To then select two more colours that also are associated with the word.
4. To indicate the strength of the association between the three selected colours and the word using a slider bar.
5. To enter text in a box to describe the reasons that they chose the colours that they did (note, however, that in this paper this text information has not been analysed).

No constraints on or recommendations to the participants were made in terms of which colours they selected. For example, particpants might choose three colours that were quite similar in response to a word prompt or they might choose three very different colours. Participants were not given any time limit to perform the task. However, each participant took about 45-60 minutes to complete the experiment. Before the experiment, participants were presented with three trial words in order to get used to the paradigm and also to provide some anchor points for the extremes of the slider bars (the use of such anchors has been shown to result in more consistent scaling between participants in such experiments). The three trial words were 'twelve', 'soft' and 'red' and these were chosen because they were thought to have little, moderate and substantial colour association respectively. Five Chinese and five UK participants were invited back (one or two days after the experiment) in order to repeat the experiment to provide information about intraobserver variability. The interface for displaying the words, for allowing the participants to select colours, and for automatically randomizing the words and collected the data was written using MATLAB.

The colours selected were automatically recorded as RGB values. After the experiment the colours selected ( 40 participants $\times 3$ colour selections $\times 33$ words $=3960$ colours) by each participant for each word were displayed on the computer screen and measured using a Minolta 2000 Spectroradiometer (the spectral data were subsequently converted to CIELAB values using the screen's white point, CIE $\mathrm{x}=0.3116, \mathrm{y}=0.3184$ ).

For each word (and for each nationality, UK and Chinese) this generated palettes containing 45 colours ( 3 selections $\times 15$ participants) and 15 colours ( 1 selection $\times 15$
participants) depending upon whether all three selections from each participant are included or whether only the first selected colour by each participant is used.

The colour measurements of each patch will be used to quantify colour differences within a palette or selection of colours and between palettes. For within a palette, each colour will be compared with each other and the CIELAB colour difference calculated; the mean of these colour differences will be used an indicator of similarity within the palette.

For comparing similarity between palettes, a method for quantifying the visual difference $\Delta \mathrm{E}_{\mathrm{P}}$ between two palettes has been developed by Pan and Westland ${ }^{28}$. The algorithm to calculate the $\Delta \mathrm{E}_{\mathrm{P}}$ is according to the following 5 steps given N colours in the palette:

1. For each colour in one palette, the CIELAB colour difference between this colour and each of the colours in the second palette are calculated. The minimum colour difference is recorded.
2. Step 1 is repeated for all the colours in the first palette, for each finding their closest corresponding colours in the second palette, resulting in N colour differences.
3. The N minimum color difference values are averaged and the mean value symbolized as $\mathrm{m}_{1}$.
4. Steps 1-3 are repeated, but this time for each of the colours in the second palette. In other words, for each of these colors the closest corresponding colour in the first palette is found. The mean value of these N color differences is symbolized as $\mathrm{m}_{2}$.
5. The values of $m_{1}$ and $m_{2}$ are averaged to obtain the visual colour difference $\Delta E_{p}$ between the two palettes

Smaller values of $\Delta \mathrm{E}_{\mathrm{P}}$ are associated with greater similarity between two palettes.

## 3 Results

The first colour selected by the UK and Chinese participants for the three trial words are illustrated in Figure 1. The mean strength of association for these words were 98.3 (red), 75.5 (soft) and 33.2 (twelve) for the 15 UK participants and 99.2 (red), 82.0 (soft) and 59.6 (twelve) for the 15 Chinese participants. This confirms the assumption made at the beginning that the words red, soft and twelve would have strong, moderate and weak
colour associations respectively. The mean colour difference within each palette was 13.6 (red), 44.4 (soft) and 99.6 (twelve) for the UK participants and 18.0 (red), 46.7 (soft) and 96.6 (twelve) for the Chinese participants. Based on these three words alone it seems that palettes that are strongly associated with the words contain more self-similar colours than palettes that are weakly associated with the words. This will be tested later for all 30 of the test words.


Figure 1: The trial results for the UK (left) and Chinese (right) participants. Each row represents the first colours chosen by one of the participants for the words red, soft and twelve (shown from left to right).

Figures 2 and 3 respectively show the first colours selected by the UK and Chinese participants for the 30 test words. For example, for the word 'hot' the vast majority of participants selected a red colour whereas for the word 'natural' the vast majority of participants selected a green colour. In some cases the colours selected are intuitively what one would expect (that is, for example, there is a large amount of published studies that would equate blue with cold, for example). Note the predominance of pink for female, and the predominance of blue for cold (and compare these to the reds that are generally associated with hot). Visually there are many similarities between the colours chosen by the UK participants and those chosen by the Chinese participants. Even when the colours that were collected were less obviously intuitive (such as the reds and blacks that are associated with bad and the colours associated with active) there is evidence of some common relationship between word and colour. However, some interesting differences can be observed between the UK and Chinese colours. For example, red is used much more by Chinese participants in association with traditional and with married. Note also the dominance of red for the word lucky in the Chinese data whereas there is a dominance of green for the same word in the UK data.

However, the similarity between palettes can be quantified using the palette comparison metric $\Delta \mathrm{E}_{\mathrm{P}}$. Table 2 lists the $\Delta \mathrm{E}_{\mathrm{P}}$ values for each of the 30 words where the UK palette is compared with the Chinese palette. From Table 2 it can be observed that the three words that produced the most similar palettes between the UK and Chinese participants were dead $\left(\Delta \mathrm{E}_{\mathrm{P}}=5.9\right)$, clean $\left(\Delta \mathrm{E}_{\mathrm{P}}=8.9\right)$ and hot $\left(\Delta \mathrm{E}_{\mathrm{P}}=9.8\right)$. The three words that produced the least similar palettes between the UK and Chinese participants were religious ( $\Delta \mathrm{E}_{\mathrm{P}}=$ 39.6), bad $\left(\Delta \mathrm{E}_{\mathrm{P}}=36.1\right)$ and rich $\left(\Delta \mathrm{E}_{P}=31.9\right)$. More work may be needed with the $\Delta \mathrm{E}_{P}$ metric to ascertain a threshold value above which palettes may be considered to be visually dissimilar. Words that produce similar colour palettes for UK and Chinese participants tend to have strong associations. However, words that produce less similar palettes may have weak associations for one or both of the participant groups (e.g. the strength of association for the word safe was only 55.5 for UK participants but is 82.8 for Chinese participants). However, the colour associations for the word bad are quite different for the two groups but both groups report strong associations (76.3 and 82.1 for UK and Chinese participants respectively) for the colours that they select. The strength of the associations are detailed in Table 3. Note that in nearly every case the strength of the association of three selected colours is a little less than for the first selected colour.


Figure 2: The experimental results for the UK participants the first colour selected. The words are, from left to right in the top row (active, bad, clean, cold, cultural, dangerous, dead, female, fresh, future, good, healthy, hot, lucky, male) and in the second row (married, medical, modern, natural, old, poor, powerful, religious, rich, safe, sweet, traditional, unlucky, urban, young).

Table 2: Visual colour differences DEP between the UK and Chinese colour palettes (smaller values of DEP indicate more similarity between the two palettes). Based on the first colour selected by each participant for each word.

| word | difference | word | difference | word | difference |
| :--- | ---: | :--- | ---: | :--- | ---: |
| dead | 5.88 | old | 16.73 | male | 23.54 |
| clean | 8.89 | dangerous | 17.25 | unlucky | 23.67 |
| hot | 9.80 | good | 17.57 | sweet | 25.28 |
| fresh | 12.95 | urban | 19.02 | young | 26.13 |
| medical | 13.41 | cold | 19.02 | future | 26.32 |
| poor | 14.38 | female | 19.39 | married | 27.06 |
| traditional | 14.72 | lucky | 19.43 | active | 27.64 |
| natural | 15.34 | powerful | 20.70 | rich | 31.93 |
| modern | 16.10 | cultural | 22.35 | bad | 36.10 |
| healthy | 16.13 | safe | 23.03 | religious | 39.64 |



Figure 3: The experimental results for the Chinese participants the first colour selected. The words are, from left to right in the top row (active, bad, clean, cold, cultural, dangerous, dead, female, fresh, future, good, healthy, hot, lucky, male) and in the second row (married, medical, modern, natural, old, poor, powerful, religious, rich, safe, sweet, traditional, unlucky, urban, young).

The analysis so far has considered the first colour chosen by each participant for each word. Figures 4 and 5 show the colour palettes ( 3 colours $\times 15$ observers) that result by considering all three colours chosen by each participant. In this figures, each row shows the colours selected by a participant with the left-most colour in each case being the first colour that was selected.

Figure 6 shows the relationship between the self-similarity of the colours in a palette and the strength of the association between word and colour. The left-hand figure shows the average $\Delta \mathrm{E}$ between the 15 -colour palettes that result from considering the first colour selected only and the average strength of the association between the word and colour $\left(\mathrm{r}^{2}\right.$ $=0.20$ ). The right-hand figure shows the average $\Delta \mathrm{E}$ between the 45 -colour palettes that result from considering all colours selected and the average strength of the association between the word and colour ( $\mathrm{r}^{2}=0.29$ ). Correlations are relatively weak but indicate some evidence that palettes that are strongly associated with the words contain more self-
similar colours than palettes that are weakly associated with the words. Other metrics that quantify self-similarity of the palettes may show stronger correlations with the visual strength data. For example, the $\Delta \mathrm{E}$ could be calculated using other colour difference equations such as CIEDE2000.

Table 3: Mean strength of association reported by participants. Data are shown for UK and Chinese participants and based on the first selected colour and on all three colour selections.

|  | UK (1) | UK (3) | China (1) | China (3) |
| ---: | ---: | ---: | ---: | ---: |
| red | 98.3 | 95.2 | 99.2 | 91.4 |
| twelve | 33.2 | 23.9 | 59.6 | 44.8 |
| soft | 75.5 | 64.3 | 82.0 | 67.2 |
| active | 62.1 | 58.3 | 84.9 | 81.7 |
| bad | 76.3 | 68.2 | 82.1 | 77.4 |
| clean | 85.8 | 73.9 | 97.3 | 88.8 |
| cold | 87.3 | 83.8 | 88.1 | 83.7 |
| cultural | 34.9 | 35.3 | 75.8 | 71.2 |
| dangerous | 87.6 | 77.9 | 93.0 | 86.4 |
| dead | 77.4 | 71.2 | 91.5 | 82.2 |
| female | 75.8 | 70.6 | 86.0 | 80.0 |
| fresh | 79.0 | 72.9 | 89.8 | 86.4 |
| future | 48.4 | 40.9 | 83.5 | 74.0 |
| good | 78.5 | 63.6 | 79.7 | 72.6 |
| healthy | 69.8 | 65.8 | 87.2 | 80.3 |
| hot | 90.7 | 84.6 | 89.8 | 85.1 |
| lucky | 79.0 | 67.2 | 90.1 | 81.2 |
| male | 78.4 | 63.9 | 78.7 | 74.6 |
| married | 68.8 | 57.5 | 90.1 | 79.1 |
| medical | 75.2 | 72.3 | 91.6 | 87.1 |
| modern | 52.6 | 45.5 | 79.9 | 70.7 |
| natural | 82.7 | 81.8 | 91.1 | 83.1 |
| old | 59.2 | 46.3 | 86.2 | 79.8 |
| poor | 58.3 | 52.2 | 81.6 | 71.8 |
| powerful | 70.7 | 63.8 | 84.9 | 73.4 |
| religious | 47.3 | 45.5 | 81.4 | 77.3 |
| rich | 74.6 | 62.8 | 89.8 | 80.3 |
| safe | 55.5 | 50.9 | 82.8 | 73.6 |
| sweet | 63.8 | 55.1 | 86.2 | 80.8 |
| traditional | 49.1 | 55.7 | 86.7 | 79.2 |
| unlucky | 54.4 | 48.5 | 81.2 | 73.6 |
| urban | 68.7 | 61.3 | 79.1 | 75.4 |
| young | 65.0 | 60.4 | 82.0 | 76.5 |
|  |  |  |  |  |

Recall that 5 UK and 5 Chinese participants undertook the experiment twice within a few days in order to investigate intra-observer variability. Intra-observer variability varied greatly between words as would expected if some words elicit strong associations and other words elicit much weaker associations (or no associations at all). The words dead for example elicited strong associations from UK participants (see Figure 2 - where the colours for dead are displayed in the upper row and the $7^{\text {th }}$ column for a visual representation and note that in Table 3 the mean strength association was 77.4).


Figure 4: The experimental results for the UK participants for all three selected colours. The words are, from left to right in the top row (active, bad, clean, cold, cultural, dangerous, dead, female, fresh, future, good, healthy, hot, lucky, male) and in the second row (married, medical, modern, natural, old, poor, powerful, religious, rich, safe, sweet, traditional, unlucky, urban, young).


Figure 5: The experimental results for the Chinese participants for all three selected colours. The words are, from left to right in the top row (active, bad, clean, cold, cultural, dangerous, dead, female, fresh, future, good, healthy, hot, lucky, male) and in the second row (married, medical, modern, natural, old, poor, powerful, religious, rich, safe, sweet, traditional, unlucky, urban, young).

For this word, the mean CIELAB colour difference between the first colour selected by the UK participants during their $1^{\text {st }}$ and $2^{\text {nd }}$ visit was 2.0 . By contrast, the word future has much weaker associations for UK participants (see Figure 2 - where the colours for future are displayed in the upper row and the $10^{\text {th }}$ column for a visual representation and note that in Table 3 the mean strength association was 48.4). For this word, the mean CIELAB colour difference between the first colour selected by the UK participants during their $1^{\text {st }}$
and $2^{\text {nd }}$ visit was 34.6 . Participants were more consistent in their choices when the associations between the words and the colours were strongest.


Figure 6: Correlation between within-palette $\square$ Es and mean strength of word-colour association for all 33 words used in the study. Note that in general the stronger the association of word and colour, the greater the similarity of the colours within a palette. The left figure shows the correlation for when only the first colour selected is considered ( 15 colours per word); the right figure shows the correlation for when all three selected colours are considered ( 45 colours per word).

## Discussion

In this paper a new method for exploring the relationships between words and colours has been suggested. The data derived have been used to show, for example, that there are a great many similarities between the word-colour relationships for UK and Chinese participants although some interesting differences were also revealed.

The majority of the published literature explore word-colour relationships by starting from a colour and finding the strength of the relationship between that colour and various terms, often expressed as bi-polar adjectives such as weak-strong or warm-cool. It is suggested that there is merit in reversing the experimental paradigm and in this study participants were asked to identify colours that relate to various adjectives. One of the advantages of this method is that it makes explicit the observation that there is not a one-to-one relationship between words and colours. Numerous infographics can be found on the internet that grossly simplify the relationship between colour and meaning, suggesting
for example that red means this or green means that. For example, in our data the colours red and blue are found in many of the relationships (see Figures 2-5). The method described in this study could be used by those who wish to explore specific relationships. Nevertheless, for practical use the method is somewhat time consuming and more efficient methods to obtain the relationships should be sought. In one previous study, it was suggested that internet scraping could be used to derive the word-colour relationships automatically based on analyses of millions of images ${ }^{29}$. Further work may be required to ascertain whether the colour palettes derived using automatic methods are consistent with those derived from psychophysical experiments. In this regard the data generated in this paper can be regarded as test data that could be used to evaluate the performance of internet-based automatic word-colour extraction (the RGB values obtained are available on request from the authors to any researchers who wish to use them). The possibility of automatic extraction of word-colour relationships offers the potential for design tools that allows users to input a word and to be presented with a colour palette - a range of colours that represent that word from which the user can select. Such a system would present obvious and commercially valuable opportunities in design, marketing and branding. It would also allow a large-scale investigation of differences in word-colour relationships that may exist between cultures, sub-cultures, or even over time in a way that small-scale laboratory-based studies will always struggle to deliver.

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