MultiPic: A standardized set of 750 drawings

with norms for six European languages

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Abstract

Numerous studies in psychology, cognitive neuroscience and psycholinguistics have used pictures of objects as stimulus materials. Currently, authors engaged in cross-linguistic work or wishing to run parallel studies at multiple sites where different languages are spoken must rely on rather small sets of black-and-white or colored line drawings. These sets are increasingly experienced as being too limited. Therefore, we constructed a new set of 750 colored pictures of concrete concepts. This set, MultiPic, constitutes a new valuable tool for cognitive scientists investigating language, visual perception, memory and/or attention in monolingual or multilingual populations. Importantly, the MultiPic databank has been normed in six different European languages (British English, Spanish, French, Dutch, Italian and German). All stimuli and norms are freely available at http://www.bcbl.eu/databases/multipic

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When Joan G. Snodgrass and Mary Vanderwart published their standardized set of 260 pictures in 1980, they probably did not anticipate the massive use scientists and practitioners would make of the set: 35 years later their paper has been cited over 4,750 times (Google Scholar), and their materials have been widely used in studies investigating object recognition, word production, aphasia, and bilingualism, among others. In addition, the materials have been normed for several other languages, making them the primary item set for cross-linguistic research involving picture naming (e.g., Alario & Ferrand, 1999; Barry, Morrison, & Ellis, 1997; Martein, 1995; Nishimoto et al., 2005; Rogic et al., 2013; Sanfeliú & Fernández, 1996; Wang, Chen, & Zhu, 2014).

Despite its broad use in the field, Snodgrass and Vanderwart’s dataset of line drawings is not without limitations. First, as demonstrated by Rossion and Pourtois (2004), the original black-and-white version of the line drawings yields impoverished recognition as compared to colorized and texturized versions of the same items. The latter help participants to retrieve the names of objects faster and more reliably. For this reason, in recent years researchers have progressively moved to the colored and copyright-free version designed by Rossion and Purtois (e.g., Bonin, Guillemard-Tsarparina, & Méot, 2013; Dimitropoulou, Duñabeitia, Blitsas, & Carreiras, 2009; Raman, Raman, & Mertan, 2014).

Second, and contrary to common belief, the original items from the Snodgrass and Vanderwart set were not freely available. Researchers had to apply for copyright clearance to the authors and at some point had to pay for using the dataset. This was a particular problem for authors wanting to develop normative data in other languages, as they could not make the stimuli available together with the norms.
A third limitation was that Snodgrass and Vanderwart’s picture databank was limited to 260 drawings, which rapidly became a serious constraint in the design of experiments. As a result, new drawings were added and normed in various languages (see the International Picture-Naming Project, IPNP, by Bates et al., 2003, and Szekely et al., 2004, for 520 black-and-white drawings of common objects and 275 black-and-white drawings of actions from multiple sources; see also Bonin, Peereman, Malardier, Méot, & Chalard, 2003; Cycowicz, Friedman, Rothstein, & Snodgrass, 1997; Moreno-Martínez & Montoro, 2012).

As usual, the situation is better for English than for other languages, in particular since the publication of the BOSS dataset, which includes photos of 1,468 pictures with various norms (Brodeur, Dionne-Dostie, Montreuil, & Lepage, 2010; Brodeur, Guérard, & Bouras, 2014). Unfortunately, these data are limited to American English and, for a subset of the stimuli, to Canadian French (Brodeur et al., 2012).

Given the above limitations, researchers who want to use normed drawings and accumulate data across different European languages have limited choice: They can either choose the Rossion and Pourtois (2004) stimulus set of 260 colored pictures, or they can use the 520 black-and-white drawings from IPNP. The IPNP dataset comes from ten different sources which, despite being described as “comparable in style” (Bates et al., 2003, p. 350), include items with different pictorial styles and graphic properties related to the artists' individual styles.

The present paper describes the MultiPic databank, which is a new database of standardized pictures that brings together the strengths of existing datasets while avoiding some of their limitations. With 750 pictures, MultiPic is a large dataset of colored line drawings coming from the same source and normed for multiple
languages. It is the result of a collaborative European project intended to provide the scientific community with a set of 750 publicly available color drawings representing common concrete concepts created by the same artist, standardized for name agreement and visual complexity in several European languages. The languages tested so far are British English, German, Italian, Spanish, French, and Dutch (separately normed for the Netherlands and Belgium). In the following sections we describe how the materials were generated and normed in order to facilitate replication and extension to other languages in future research.

**Method**

**Participants.** 620 undergraduate and graduate students (443 females) with a mean age of 22.03 years (SD=4.26) were tested across sites (see Table 1 for the specific characteristics of the sample tested in each language). Only participants who were native speakers of the target language were selected. Ethical approval for conducting the study had been obtained from ethics boards of the participating institutions as required.

**Table 1.** Characteristics of the samples tested in each language, and means and standard deviations (in parentheses) of the different indices obtained in the norming study in each language.
**Materials.** The materials were commissioned by the first author and drawn by a local artist. An initial set of nearly 600 Spanish words with highly distinctive pictorial characteristics were selected from the Spanish lexeme database ESPAL on the basis of their imageability and concreteness values (ranging between 4 and 7 on the 1-to-7 Likert scales; see Duchon et al., 2013). These words were selected to cover a wide range of frequencies and semantic fields. After consultation with the artist, some items were dropped from the list because they were difficult to express in line drawings, and an extra set of nearly 200 items was added. The artist then started the creation of the pictorial set. She created the drawings with a freehand computer application (using a digital tablet and pen set) that was optimal for making digital, two-dimensional graphics. The artist used the same space for the whole set, so that in its original form each picture occupied a space of 15x15 centimeters (DPI=300 pixels/centimeter). In addition, the artist was asked to use the same graphic style for all the drawings, using strokes of similar width. Finally, we asked for coherence in the coloring of the drawings. Two external informants (research assistants from the BCBL) assessed the quality of the drawings and improvements were made when needed, resulting in the final set of 750 drawings used in the normative studies.

**Procedure.** Seven norming studies were carried out, one per language or dialectal variation. Data acquisition was lab-based for all studies, with the exception of the Dutch study (Netherlands), which was carried out over the Internet. The English, Italian, German, and Spanish studies were carried out using Experiment Builder (SR-Research, Ontario, Canada). The French study was implemented in OpenSesame 2.8.3 (Mathôt, Schreij, & Theeuwes, 2012). Finally, the Dutch studies (Netherlands and Belgium) were run with Internet presentation using software developed at the Max Planck Institute for Psycholinguistics in Nijmegen.
Participants were tested individually. The 750 drawings were presented one by one in the center of a computer screen. Participants were asked (a) to type in the name that they felt best described the picture, and (b) to rate the visual complexity of the drawing on a scale from 1 (very simple) to 5 (very complex). They were specifically asked to provide a single name for each picture, avoiding the use of longer noun phrases. After both answers had been provided, the next picture was shown. If the participants did not know the object or its name, they were asked to use a specific language string (e.g., “DK” in English, corresponding to “don’t know”). For the visual complexity rating, participants were explicitly told not to rate the visual complexity of the object in real life, but the complexity of the drawing, as has been done in previous studies.

The drawings were presented in a different random order to each participant. The experimental session was self-paced and lasted about 90 minutes.

**Data preprocessing.** First, a native speaker of each language checked the answers for obvious spelling errors and corrected them. Together with this, an initial response recoding was done for each language by merging basic variants of the same names (e.g., hyphenated or pluralized forms) and discarding parts of speech that were not nouns or verbs (e.g., determiners and adjectives).

Second, trials where participants did not know the name of the concept (2.4% of the data across languages) and idiosyncratic responses (i.e., responses provided by a single participant, corresponding to 2.7% of the data) were excluded (see Table 1 for details). In line with the liberal criterion tested by Snodgrass and Yuditsky (1996), we excluded the responses given by single participants from further analyses, in order not to jeopardize the final outcome by outliers. We felt safe to do so because our stimuli
were normed by 60 or 100 participants (rather than the much smaller sample sizes used in other normative studies). Researchers who are particularly interested in idiosyncratic responses are referred to a file with the raw data.\(^1\) The trials with non-idiosyncratic responses constitute the core of MultiPic and those data were submitted to further analyses. These data ranged from 93.1% of the trials for Belgium Dutch to 97.3% of the trials for Spanish.

Third, name agreement was computed for each of the 750 drawings in each dataset calculating the $H$ index and the percentage of modal names. The $H$ index is an information statistic that reflects the level of agreement across participants (Shannon, 1949). When all participants give the same name to a given drawing the $H$ index is 0, and the $H$ value increases as a function of the divergence of responses. We also calculated the percentage of participants who gave the most frequent name (i.e., the modal name). As stated above, these indices were based on the agreement for each modal noun regardless of the modifiers (see O'Sullivan, Lepage, Bouras, Montreuil, & Brodeur, 2012). Nonetheless, it is worth mentioning that only 0.81% of the unfiltered responses included more than one word. Visual complexity scores were obtained by averaging the numerical responses to each drawing across participants in each sample and by computing the corresponding means (see Table 1).

### Results

\(^1\) Almost all researchers exclude the “don’t know” responses; others exclude tip-of-the-tongue responses, or responses given by single participants. To assess the impact of our exclusions, we calculated the $H$ indices for each item in each language 1) including the trials where respondents did not know the name and the trials with idiosyncratic responses, and 2) including the trials with idiosyncratic responses but excluding the unknown items. The correlations between the $H$ indices calculated in these ways and the $H$ indices calculated from the filtered data ranged from $r=.94$ to $r=.98$ across languages, demonstrating that the filtered data agree with the unfiltered data to a great extent. For those researchers interested in the idiosyncratic responses, we have stored the unfiltered data on a publicly available website [https://figshare.com/s/bf4a360ab3ec9053729f](https://figshare.com/s/bf4a360ab3ec9053729f), so that colleagues can calculate the exact metric they are interested in.
Name agreement. As shown in Figure 1, many drawings were named consistently by most participants (in all languages, more than 40% of the items yielded $H$ indices below 0.5). The mean $H$ index in the different languages ranged from 0.50 (Spanish) to 0.86 (German), and the mean percentage of modal names ranged from 78.5% (German) to 87.6% (Spanish). Across languages, many items yielded a single response (i.e., $H$ index of 0). These numbers ranged from 139 in German to 280 items in Spanish. Similarly, the percentages of the most frequent responses for each item were reassuringly high, with a substantial number of drawings yielding values higher than 70% (ranging from 512 in Belgian Dutch to 628 in Spanish).

Figure 1. Density plots of the $H$ index in each language.

Native speakers individually inspected the 15,695 different names constituting MultiPic to classify them by their correctness (i.e., whether or not they could be

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Interestingly, the resulting name agreement values were better (i.e., lower $H$ indices and higher percentages of modal names) in Spanish than in the other languages. This is likely due to the facts that the initial list of concepts was created from a Spanish lexical database and that the artist was Spanish. Likewise, it seems reasonable to assume that the Snodgrass and Vanderwart stimulus set has some North-American bias.
considered as “reasonable responses”\textsuperscript{3}. As in preceding studies, this classification was based on the identity of the depicted concept (see O’Sullivan et al., 2012). Responses were classified as correct when the word appropriately reflected the concept presented, and this category included synonyms, near-synonyms, hypernyms, responses including addition or deletion of some of the words, specifications of the function or part, brands, breeds, types and other forms of hyponymy. In contrast, responses were classified as incorrect when the words corresponded to unrelated concepts, semantically related but distinct concepts, visually similar but semantically unrelated identities, and to incorrect specifications (as in O’Sullivan et al., 2012).

According to this manual coding, only 7.1\% of the responses across languages were classified as incorrect (SD=3.42; range: 0.74-9.93 across languages). At the same time, it became clear that the decision whether a response is acceptable or not is far from trivial. We looked at the inter-rater reliability (IRR) by asking three trained coders to rate the responses to a random subset of 150 drawings from the Spanish dataset. Kappa was computed for each coder pair and then averaged to provide a unitary index of IRR (see Light, 1971). The resulting kappa indicated partial disagreement between raters as to which responses could be considered correct (κ=.579; see Landis & Koch, 1977). Readers who want to use the stimuli in a particular language, can find all the unfiltered data and the correctness classification as supplementary materials by referring to https://figshare.com/s/bf4a360ab3c90537291f.

\textbf{Table 2.} Correlation matrix for the main indices across languages with the corresponding Pearson’s r coefficients. All bivariate correlations resulted significant after correcting for multiple comparisons, except for those marked with an asterisk. H corresponds to the \textit{H} index. MN corresponds to the percentage of participants giving the modal name as a response. VC corresponds to the visual complexity rating.

\textsuperscript{3} The authors thank an anonymous reviewer for pointing them to the need of this analysis.
Cross-language comparisons. The mean name agreement indices for the current dataset (an $H$ index of 0.74 and 80% modal name responses across languages) are in line with those reported in preceding studies (e.g., Alario & Ferrand, 1999: 0.36 and 85%; Bonin et al., 2003: 0.67 and 77%; Brodeur et al., 2014: 1.71 and 60%; Dimitrakopoulou et al., 2009: 0.55 and 87%; Liu, Hao, Li, & Shu, 2011: 1.32 and 66%; Manoiloff et al., 2010: 0.68 and 81%; Nishimoto et al., 2005, liberal criterion: 0.74 and 91%; Rossion & Pourtois, 2004, colorized: 0.32 and 90%; Sanfeliú & Fernández, 1996: 0.27 and 82%; Snodgrass & Vanderwart, 1980: 0.56 and 87%; Tsaparina, Bonin, & Méot, 2011: 0.82 and 81%). In general, name agreement is higher for smaller (often more carefully selected) stimulus sets than for larger stimulus sets (Brodeur et al., 2014).

Because the MultiPic database was tested in the same way across languages, it provides opportunities to look at language commonalities and differences, and it deals with the problem mentioned by Dell’Acqua, Lotto and Job (2000) that any language
difference in previous studies could be due “to a pure cultural difference in the subjects’ population or to a structural difference in the pictures submitted to the subjects’ judgment” (p.592). Because the MultiPic database includes name agreement data and visual complexity ratings for the same input in different languages, a within-study cross-cultural comparison can be performed.

To do so, a series of correlation analyses were run in order to verify the cross-linguistic commonalities of the measures obtained. These correlations are presented in Table 2, where Pearson’s $r$ is reported ($\alpha=.05$ after Holm’s correction for multiple comparisons). All cross-language correlations between related indices were high, demonstrating the inter-lingual convergence of the results. Not surprisingly, the strongest (close to perfect) correlations were obtained for the visual complexity scores, with $r$ values between .90 and .97.

In addition, a hierarchical cluster analysis was performed using Hoeffding’s test of independence ($30^*D$ statistic; Hoeffding, 1948). According to this measure of mutual independence ranging from -0.5 to 1, larger values of $D$ represent higher dependency between two variables. As shown in Figure 2, the visual complexity measures in all languages clustered together with high $D$ scores, being independent of the rest of the variables. The $H$ indices and the percentages of modal responses showed high interdependency scores in each language, given that they represent two alternative forms of measuring name agreement. These results showed that the data for each language clustered together, and more interestingly, that the different languages formed coherent clusters on the basis of the language families (i.e., Germanic: Dutch, German and English; Romance: French, Italian and Spanish).
Figure 2. Results of the hierarchical cluster analysis using Hoeffding’s test of independence (30*D statistic).

Discussion

In this article we presented the open-source Multilingual Picture (MultiPic) database, which includes norming data in different European languages for a large set of colored drawings from the same source created for psycholinguistic and (neuro)cognitive research. MultiPic includes 750 drawings of a variety of concepts that have been normed for name agreement and visual complexity in 6 different languages (German, Spanish, British English, French, Italian and Dutch, assessed separately for speakers in Belgium and The Netherlands).

The results revealed a high degree of convergence of the responses across speakers of the same language for each drawing, as shown by the low $H$ indices and the high percentages of participants providing modal names. Furthermore, cross-linguistic correlations showed that the name agreement scores for the individual items were relatively similar across languages, while at the same time highlighting interesting language- and culture-dependent differences (see also Brodeur et al., 2012; Dell’Acqua.
As shown in Table 2, the correlations across languages in the $H$ indices ranged between $r=.51$ and $r=.68$, except for the two samples of Dutch-speaking participants, who showed a correlation of $r=.83$. A similar result was found when comparing the percentages of participants providing the modal names as responses: All bivariate correlations resulted in $r$ values between .45 and .63, with the exception of Dutch-speaking groups that showed a much higher correlation of $r=.79$. The moderate strength of the correlations in name agreement across languages is in line with earlier evidence (see in particular Dell’Acqua et al., 2000).

The correlations found between the two Dutch-speaking groups show that a shared language boosts the concordance of the naming, but that it is not enough to get exactly the same names for all pictures. Because the two groups come from two different countries (The Netherlands and Belgium) some names diverge as a result of cultural differences between the two communities. These data can be compared to the results of Brodeur et al. (2012), who showed that norms obtained from different groups of Canadian participants speaking either French or English were highly similar, suggesting that their underlying common cultural setting lessened the impact of their linguistic differences.

MultiPic represents an alternative to the pictorial sets currently used in cross-language research (e.g., Snodgrass and Vanderwart, 1980; Rossion & Pourtois, 2004; Székely et al., 2004). It overcomes the limitations of the existing sets by including considerably more items than any of the preceding databanks and by being created by the same person in order to minimize heterogeneity in format, stroke, color codes and styles of the drawings. Most importantly, the drawings included in MultiPic have been standardized in different languages using a parallel protocol (for a similar approach, see also Bates et al., 2003; Székely et al., 2004; Viggiano, Vannucci, & Righi, 2004; Yoon...
et al., 2004). This is an important feature, because it enables researchers from different European countries to use the same materials not only for research on monolingual samples, but also for cross-linguistic research on multilingual participants.

At the same time, we acknowledge that the databank, like any other tool, has its limitations. MultiPic only provides name agreement and visual complexity ratings at the moment. The inclusion of additional subjective ratings on other characteristics of the drawings, such as image agreement, color diagnosticity or emotional valence will be of interest. Another limitation is that the MultiPic databank only includes drawings of individual concrete objects. An extension to drawings of events and actions would be desirable. Researchers should also use the statistics as intended. Given that the $H$ indices and the percentages of modal responses were calculated from the filtered data (i.e., excluding idiosyncratic responses and unknown items), it is advised to take these categories into account when selecting stimulus materials. Drawings with too many "don't know" and idiosyncratic responses can easily be excluded on the basis of the information provided in the various columns of the summary files on the MultiPic website.

Because of the limitations, researchers may find MultiPic particularly relevant in combination with other existing databases. For instance, a comparison of the names of pictures with those of BOSS (Brodeur et al., 2010, 2014) can inform researchers about language differences between speakers of British and American English. Authors can also easily check whether a stimulus used in a bilingual population has the same name agreement in both languages. Also, information about word frequency, word age-of-acquisition, word prevalence and other concept-related features can easily be added to MultiPic from other sources.
In line with current standards in the field, the materials are open-access and free from copyright restrictions for non-commercial purposes, facilitating the use of the data and extensions, for instance to further samples and other languages. Interested researchers can find the materials at http://www.bcbl.eu/databases/multipic, where colored and grey-scaled versions of the drawings are presented in different formats together with the corresponding norms for each of the languages we tested.
References


