

# Concepts in context: Evidence from a feature-norming study

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## Abstract

Concepts are typically conceived as context-free knowledge structures. Recently, a different view has emerged according to which subjects produce situation-specific conceptualizations, thereby raising important questions about the level of contextual dependency in conceptual representations. In this paper, we present a feature-norming study in which subjects are asked to generate properties of concepts presented in context. Collected data are analysed to investigate the actual amount of conceptual variation induced by contexts and the effect of context modality.

**Keywords:** Semantic feature norms; property generation; context.

## Concepts and contexts

Both in classical and in post-classical models, concepts have been conceived as substantially context-free knowledge structures. Regardless of the particular theory (e.g. exemplar, prototype, and connectionist), it is normally assumed that concepts result from abstracting critical information about an entity *per se* (such as shape, colour, etc.), leaving behind background situations (i.e. the contexts) in which these entities are experienced. Concepts thus become invariant to different contexts of use. Accordingly, the same representation of an apple is used both when categorizing an entity on a tree, and when categorizing the same entity in a supermarket.

Recently, this view has been overtly criticized. For instance, Yeh and Barsalou (2006) argue that concepts not only contain a large array of situational information about the physical settings, events, and subjective perspectives of agents, but they also produce different conceptualizations in different contexts. That is, the supermarket situation would activate context-specific information concerning an apple, different from that activated by a different context, such as a tree in a garden. These two claims directly follow from the perceptual simulation model adopted by the authors, but more in general they raise important questions about the level of contextual dependency in our conceptual representations. Wu and Barsalou (2009) use a property generation task to investigate the situated nature of concepts, and report that approximately 26% of the features produced by subjects were indeed situation-related. Subjects generated properties (semantic feature norms) provide interesting evidence about conceptual representations, but one intrinsic limit of the study in Wu and Barsalou (2009) is that stimuli were presented *out of context*, as is customary in semantic norming. This way, it be-

comes impossible to address and test the more specific and crucial issue concerning the relation between concepts and context, that is the actual effect of the context in modulating and biasing conceptual representations.

In this paper, we present a feature-norming study in which subjects are asked to generate properties of concepts presented *in context*. To the best of our knowledge this is the first property generation task with this type of design. While we do not commit ourselves to any specific model of conceptual representation, collected data allow us to directly address three key issues concerning the effects of contexts on concepts: i.) the actual amount of conceptual variation induced by contexts, and ii.) the property types that are more subject to contextual variation, and iii.) the effect of the context modality. In particular, we will investigate the effect of both linguistic contexts (i.e. a sentence in which the context noun appears) and extralinguistic contexts (i.e. an image of a situation in which an entity can be experienced).

## Semantic Feature Norms

Nowadays there is a strong consensus on the fact that it is possible to describe the internal structure of a concept in terms of a set of semantic properties (Garrard, Lambon Ralph, Hodges, & Patterson, 2001; Baroni & Lenci, 2008). A traditional way to access and study the structure of conceptual knowledge is the use of semantic feature norms. These are lists of properties that participants produce describing and defining a specific concept; moreover they include several measures and statistics calculated according to feature production frequencies.

As suggested by McRae and colleagues (McRae, Cree, Seidenberg, & McNorgan, 2005) these lists do not provide a static and definitive representation of concepts but they are the most direct way to study the dynamics associated with the online process that subjects carry out when they have to process a specific concept.

Different researchers used these lists to investigate various aspects of human cognition. They have been used to test the psychological validity of hypothesis (Wu & Barsalou, 2009), and as stimuli for different experiments such as semantic similarity (McRae, Sa, & Seidenberg, 1997) and property verification tasks (Cree, McNorgan, & McRae, 2006).

One of the most widely used norms is the collection of McRae et al. (2005). This is the largest set of semantic prop-

erties freely available: it includes properties for 541 leaving and non-leaving English nouns. Another smaller example is represented by the collection of Vinson and Vigliocco (2008). In this case the authors extended their analysis to the domain of actions and events. They collected norms for 167 living and non-living objects and for 287 events and actions.

The pros of these collections are quite straightforward, however they manifest also different limitations (McRae et al., 2005). Traditionally, they are restricted to a relative small number of concepts (few hundreds) and the process of collection, normalization and classification is very long and expensive. Moreover, the linguistic nature of the task favours the information which is easily verbalized, penalising spatial and temporal relations between entities. During the classification phase, the annotators have to reinterpret the intents of the subjects and they cannot always preserve the original information. Finally, as we said above, all existing norms were collected by presenting words in isolation and not associated with a specific context.

### Collecting context-sensitive feature norms

The main goal of this work is to describe the collection of semantic feature norms for 8 concrete concepts and to analyse the effects that contextual variability exerts on the number and types of properties produced.<sup>1</sup>

### Design

The collection was performed on-line using a website interface. To reduce any biasing effect on the participants we adopted a very simple layout and a sans serif font.

**Stimuli** The 8 normed concepts correspond to the following English nouns: *apple*, *banana*, *bear*, *horse*, *bike*, *car*, *hammer*, and *knife*. Nouns were selected in order to have an equal number of animate and inanimate concepts belonging to the semantic classes traditionally used in this kind of studies, that are fruits, animals, vehicles, and tools.

For each concept we identified two alternative situations frequently associated with the correspondent object. We downloaded from the Web 16 colour pictures depicting the two contexts for each concept and we downsized them (288\*320 pixels). The pictures do not include the target object unless it is strictly necessary for the correct interpretation of the context (e.g. a showroom without some cars inside would not be representative). This way, participants are not biased in their descriptions by a specific instance of the concept appearing in the picture. A native English speaker produced 16 sentences describing the context depicted in the correspondent picture. Unlike the visual contexts, the sentences include the target concept noun (written in capital letters).

For every trial, the target concept noun (in capitals) appears on the top of the screen and it is followed by 10 blank lines that participants have to fill in with concept properties. In

the case of the linguistic context, the sentence containing the target word appears instead of the target word. For the visual context, the picture appears on the left of the blank lines.

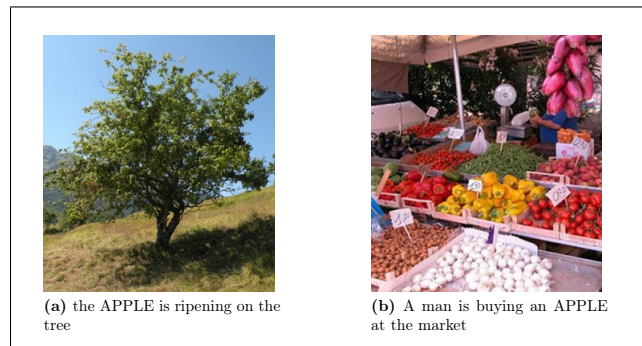


Figure 1: *apple*: visual and linguistic contexts.

Figure 1 shows the visual and linguistic contexts for *apple*.

**Procedure** The experiment included 40 trials: 8 concepts in 5 different contexts (2 visual, 2 linguistics, and 1 no context). 5 lists of 8 items each were created: the distribution of the items ensured that every list comprised each of the 8 concepts in a specific context. The order of the trials was semi-random: all the out of context trials appeared before the linguistic ones and those before the visual ones. In this way, the complexity of the stimulus was increasing during the experiment. 125 native English speakers performed the experiment. 100 of them were recruited posting the request on different mailing lists, the remaining 25 using Amazon Mechanical Turk. The experiment started after written instructions of the task and 3 examples. The task required to read the word (or the sentence) at the top of the screen and to produce a maximum of 10 properties per concept describing different aspects of it: colours (e.g. CHERRY *red*)<sup>2</sup>, tastes (e.g. ICE CREAM *good*), shapes (e.g. BALL *round*), functions (TRAIN *transportation*), typical locations (SHARK *ocean*), emotions (CHRISTMAS *excitement*), evaluations (SOUP *hate*), ...

The entire experiment lasts around 15 minutes.

### Post Processing

**Data Codification** At the end of the collection a process of filtering and refinement took place. We identified all the synonym properties and we normalized to the same property (e.g. “bike” and “bicycle” were coded as *a bike*). All the coordinative or disjunctive features providing more than one piece of information were split into single elements. For example, “is red, green or yellow” became *is red*, *is green*, and *is yellow*. We removed all the quantifiers (e.g. “can be”, “generally”, “usually”) and the material not relevant to our analysis (e.g. miscellanea, incoherences, and free associations). The resulting properties were coded according to a specific set of patterns (e.g. *beh - eats honey*).

<sup>1</sup>The collection is freely available at <http://sesia.humnet.unipi.it/norms>

<sup>2</sup>In this work, we use caps to indicate CONCEPTS and italics to indicate the *properties produced*.

**Coding Scheme** The properties were classified according to a reviewed and reduced version of the coding scheme proposed by Wu and Barsalou (2009). The scheme proposes 4 main categories to classify the properties.

- *Taxonomic properties*: properties describing taxonomic relations (hyperonyms, hyponyms, synonyms, and coordinates).
- *Entity properties*: properties describing the entity per se (e.g. internal and external properties and elements, prototypical behaviours).
- *Situation properties*: properties associated with the contextual background (e.g. locations, time, participants, functions).
- *Introspective properties*: properties describing feelings and mental states (e.g. evaluations, contingencies).

## Results

Table 1: Average and Standard Error of the number of features produced by each subject for each concept grouped according to broad property class and modality.

Class	No Context		Visual		Linguistic	
	Mean	SE	Mean	SE	Mean	SE
TAX	1.20	0.09	1.34	0.15	1.24	0.11
ENT	3.82	0.40	3.64	0.36	3.86	0.41
SIT	2.34	0.26	2.66	0.31	2.15	0.25
INT	1.61	0.19	1.59	0.19	1.64	0.19

Participants produced 6922 properties in total: 3619 entity properties, 2025 situation properties, 644 introspective properties, and 634 taxonomic properties. Table 1 reports the averaged number of features (and Standard Error) produced by every subject for each item and grouped by property class and modality. There are no strong differences among modalities (visual, linguistic, and no context): this suggests that different contexts are not exerting an appreciable effect on the number of properties generated. The differences arise analysing the property classes. The properties describing different aspects of the target concept (ENT) are the features produced more frequently (52% of the total). Properties providing contextual information related to the target concept (SIT) are produced the 29% of the time (interestingly, this figure is close to the one reported by Wu and Barsalou (2009)). Properties describing mental states (INT) and taxonomic relations (TAX) are less frequent (around 9%).

### Analysis

**Model** We analysed our data adopting the framework of the linear-mixed effect models with a Poisson linking function (Baayen, Davidson, & Bates, 2008). The dependent variable was the property frequency. Table 2 presents the

coefficients and p-values of the mixed model. To analyse the effects exerted by contextual variability we included two factors in contrast coding: the factor *Modality* for the effects of visual (+.5) and linguistic context (-.5), and the factor *Context* comparing the effects produced in the out-of-context (-.5) and in the in-context (+.25) conditions. We also analysed the effects associated with the type of feature produced: the factor *Property* compares the object related properties such as entity and taxonomic properties (-.25) and the context related properties such as situation and introspective properties (+.25); the factor *ObjectProp* codes the effects of taxonomic (-.5) and entity (+.5) properties; and the factor *SituationProp* the effects of situation (-.5) and introspective (+.5) properties. The random factors are: *Concept* and *Subject* representing the model intercepts, and the random slopes for all the main effects (*Modality*, *Context*, *Property*, *ObjectProp*, and *SituationProp*).

Table 2: The coefficients for the mixed effects model.

Predictor	Coefficient	Signif.
(Intercept)	0.66	***
Property	-0.28	
ObjectProp	1.10	***
SituationProp	-0.43	***
Context	0.01	
Modality	0.04	
Context:Property	-0.09	
Context:Object	-0.17	
Context:Situation	-0.06	
Modality:Property	0.14	
Modality:Object	-0.13	
Modality:Situation	-0.29	**

\* $p < .05$ , \*\* $p < .01$ , \*\*\* $p < .001$

**Property Types** The factor *Property* compares the global mean of entity and taxonomic properties (properties more associated with the object) with the global mean of situation and introspective properties (properties that are more associated with the context). We find a slightly significant effect ( $p < .1$ ) in favour of the first group of properties: object related properties are produced more frequently than context related properties. The factor *ObjectProp* shows a significant positive main effect for the entity properties compared to the taxonomic ones. The factor *SituationProp* reveals a positive effect for situation properties compared to the introspective ones.

**Context and Modality** As expected, there is no significant main effect for *Context* and *Modality*. We only find a not significant effect associated with the visual context. Participants produced almost the same number of properties independently of the presence and type of contextual information

they are exposed to.

**Interactions** The interactions reported in table 2 do not report significant effects. The only significant effect is described by the interaction `Modality:SituationProp`: situation properties are positively biased by visual contexts, while introspective properties are more biased by linguistic properties.

## Qualitative Analysis of Feature Types

Table 3: Percentage of the features that are context dependent (associated with only one context) and context independent (associated with both contexts).

Property	Dependent	Independent
C-super (.78)	0.08	0.92
C-subord (.19)	0.57	0.43
C-coord (.02)	1	0
C-syn (.02)	0.5	0.5
<b>Taxonomic</b>	<b>0.2</b>	<b>0.8</b>
E-exsurf (.27)	0.15	0.85
E-excomp (.24)	0.12	0.88
E-sys (.21)	0.28	0.72
E-beh (.07)	0.39	0.61
E-incomp (.06)	0.24	0.76
E-insurf (.06)	0.17	0.83
E-mat (.06)	0.07	0.93
E-quant (.02)	0.53	0.47
E-whole (.01)	0.21	0.79
<b>Entity</b>	<b>0.2</b>	<b>0.8</b>
I-cont (.68)	0.62	0.38
I-eval (.30)	0.38	0.62
I-emot (.02)	0.5	0.5
<b>Introspective</b>	<b>0.54</b>	<b>0.46</b>
S-func (.47)	0.25	0.75
S-assoc (.15)	0.58	0.42
S-loc (.15)	0.44	0.56
S-action (.08)	0.33	0.67
S-particip (.08)	0.48	0.52
S-origin (.06)	0.1	0.9
S-time (.01)	0.59	0.41
S-socart (<.01)	1	0
<b>Situation</b>	<b>0.35</b>	<b>0.65</b>

In this section we perform a qualitative analysis of the data to identify the main effects exerted by contextual information on specific property types.

For each concept, we divided the features produced in both contexts from those associated only with a specific context. We did the same procedure for visual and linguistic contexts individually. The aim of this analysis is to determine which property types are more dependent from a specific context (context dependent) and which are produced in

both contexts (context independent). We are interested in a general evaluation of this effect without taking into account inter-conceptual variability, for this reason we summed together the results obtained for each concept. After a preliminary analysis, we discovered that property types show an almost coherent trend both in a visual and linguistic context. We analysed the effects exerted by the two modalities using a linear model and it emerged that they are not significant both considering the main effect ( $\beta_{visual} = 3.90, p = 0.94$ ) and the interaction with the contextual-dependency variable ( $\beta_{visual:ContextInd} = 4.54, p = 0.88$ ). Table 3 reports the results from a general point of view, without modality distinction. If there is a difference between the two modalities we will discuss it separately. We report the percentage of context dependent and context independent properties out of the total number of properties of the same type (e.g. the 92% of hyperonyms are context independent). At the end of each group of properties we also present in bold the percentage for the entire class. In brackets there is the number of properties of each type out of the total number of properties in the same class.

**Taxonomic Properties** Taxonomic properties describe highly stable relations among concepts. As expected, the 80% of these properties are equally produced in different contexts. The hyperonyms (C-super) are the 78% of the entire taxonomic class. These properties are represented by a small number of high frequency feature types (in total only 22 for the visual and linguistic modality) describing associations strictly language related. Hyponyms include an high number of infrequent property types (in total 48 subordinates) and are more context dependent. This can be expected, given that each concept is associated with many hyponyms, which in turn might become differently prominent depending on the context. Coordinates and synonyms are only a small group of properties.

**Entity properties** The trend of this class is consistent: all the properties describing objects' qualities are not sensitive to contextual variability (the 80% of the total). The only exception is E-quant (e.g. *APPLE different varieties*); however this group includes a very small number of features and it is valid only for the linguistic modality (65% of context dependent properties).

**Introspective properties** The most substantial group among introspective properties is represented by contingency properties (e.g. *APPLE is good with cinnamon*). These properties describe the "common sense knowledge" that we have about a specific object in specific conditions. For this reason it is not surprising to see a strong contextual effect. On the other hand, evaluations about the object (I-eval, e.g. *APPLE is delicious*) are less context dependent: participants have a personal opinion about every object that is unlikely to change in different situations. Emotions are very few cases.

**Situation Properties** The behaviour of this group of data is more various, given also the large heterogeneity of the properties in this class. Some properties are more intrinsically related to an entity, and therefore more invariant across situations: for instance, typical functions (*CAR used for transportation*), actions (*APPLE used by cooking*), origins (*APPLE grows on trees*) and locations (*BANANA grows in tropical climates*). Instead, other property types are more context-related, and therefore subject to stronger cross-situation variation, such as associations (*CAR associated with speed*). Participants (*BANANA eaten by monkeys*) are almost equally present in both sets.

### Analysis of Feature Density

The last analysis we carry on compares the distribution of specific features in terms of feature density: how many subjects produce the same feature for the same stimulus. We analysed the features divided into the context dependent and context independent sets. In the context dependent set the 85% of specific properties are produced by only one person, the 11% by two different subjects. The remaining 4% is shared by properties produced from 3 to 10 people. On the other hand, in the context independent set we have the 21% of features produced only by 2 subjects (the minimum value for having an overlap). The maximum number of subjects that produced the same feature is 47 (*KNIFE is sharp*).

### General Discussion

The experiment described in this work was aimed to test the effects exerted by contextual variability on the production of semantic properties by human beings. We produced both quantitative and qualitative evidence of these effects. Neither modality nor context variability have significant effects on the number of features produced. People list almost the same number of properties in different contexts. However, these properties are not equally distributed and the differences among them are statistically significant. As already emerged in the literature, subjects produce more entity properties than taxonomic ones and more situated properties than introspective ones. It is interesting to note that merging together the properties more object related (entity and taxonomic) and the properties more context related (introspective and situation) the difference decreases considerably with only a slightly significant effect in favour of the first group. This remarks the fact that people are including in their dynamic representation of concepts both information describing the object per se but also almost the same amount of background information. To gather more evidence, we performed also a qualitative analysis. In this case, we extracted from our collection only the properties produced in context and we identified those occurring with both contexts and those associated with only one. The results are straightforward for the taxonomic and entity properties: almost the 80% of all the properties classified in this way are produced in both contexts. We assist to an opposite effect when we move to the introspective properties.

More complex is the dynamic of situation properties: part of them are more related to the object and other to the context.

These results suggest that the context sensitivity of concepts is strongly limited to certain property types. A possible explanation can be found in Barsalou (1982). In this work the author suggests the existence of two different kinds of properties: *context independent properties* strictly associated with the object *per se*, and *context dependent properties* associated with the specific context in which the word appears. Our data point in the same direction. It is possible to identify a large group of “core” properties that are not biased by contextual variability (in particular entity and taxonomic properties, as expected) and a smaller group of more dynamic properties produced less frequently and only associated with specific contexts (introspective properties, and partially situation properties).

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