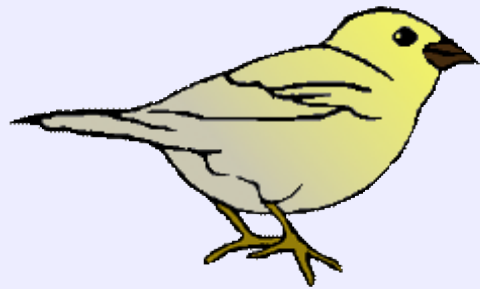
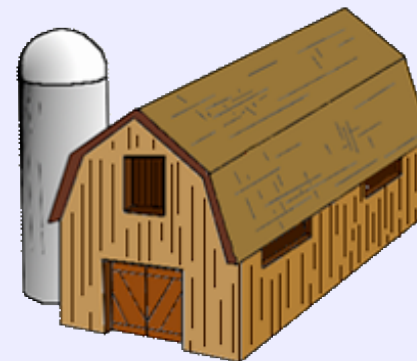
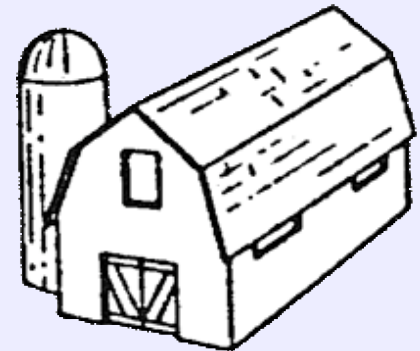


Revisiting Snodgrass and Vanderwart's object databank: color and texture improve object recognition



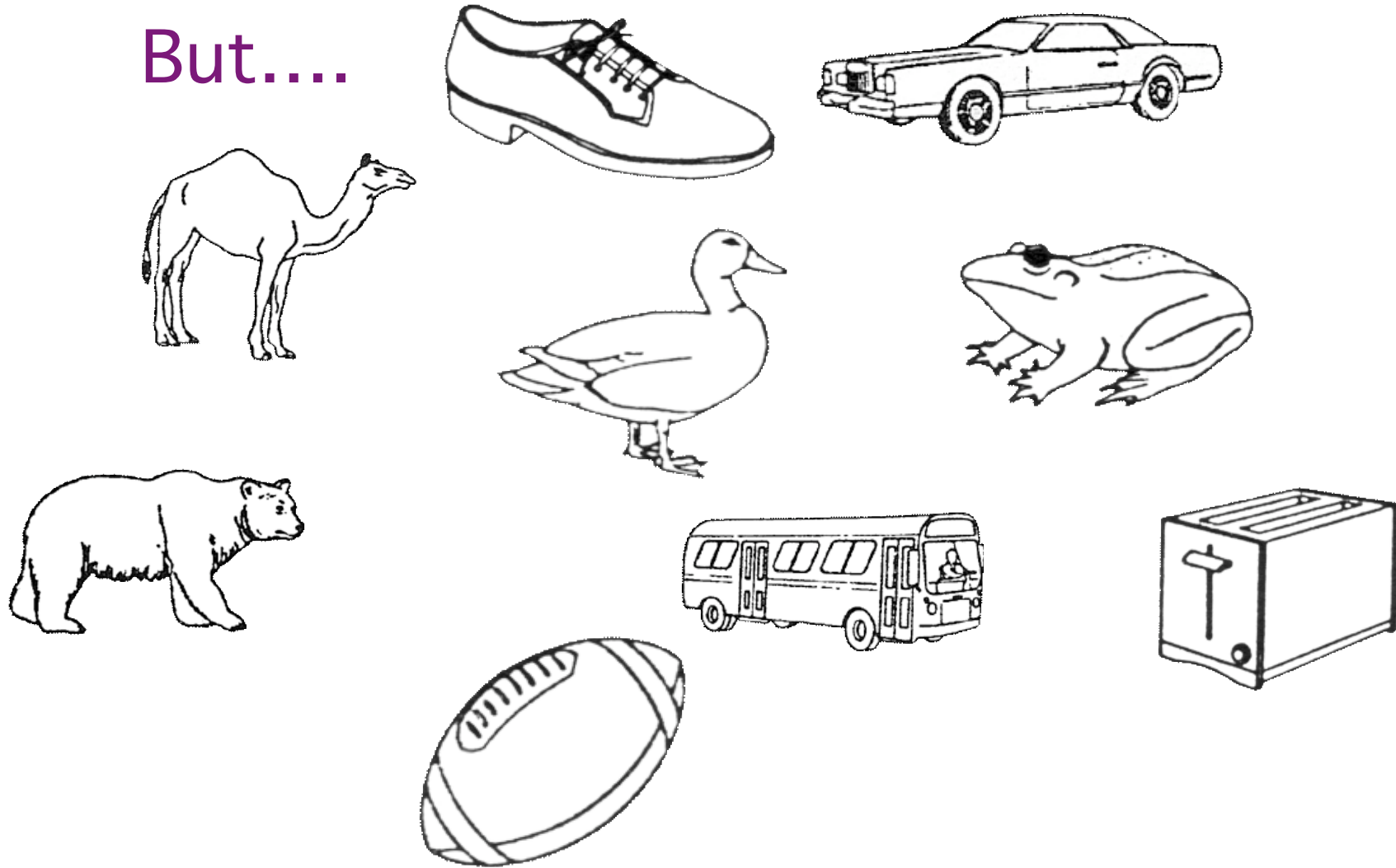
Bruno Rossion (Brown University)
Gilles Pourtois (Tilburg University, The
Netherlands)

University of Louvain, Belgium



Snodgrass, J.G. & Vanderwart, M. (1980). A standardized set of 260 pictures: norms for name agreement, image agreement, familiarity, and visual complexity. *JEP:HPP*, 6, 174-215.

But....



Only line drawings available, without any texture and color information !

Why care?

- No advantage of color over black and white photographs in object classification and semantic tasks (Oostergaard & Davidoff, 1985; Davidoff & Oostergaard, 1988)
- No difference in correct naming latencies for simple line drawings and colorized photographs (Biederman & Ju, 1988)

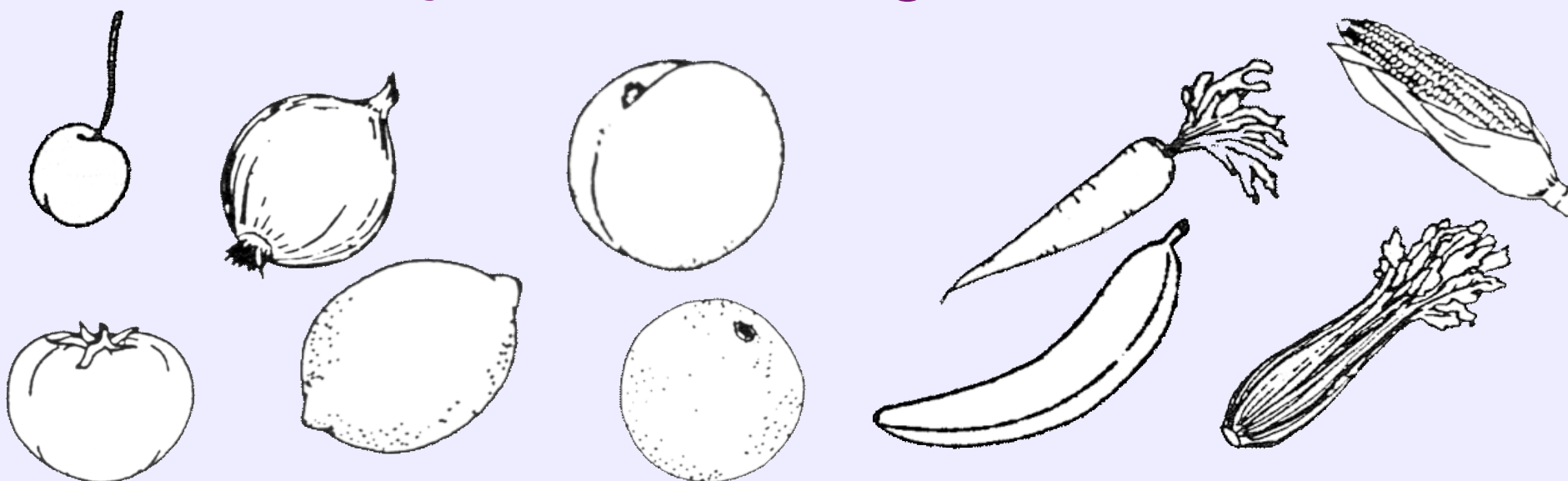
→ **Only shape matters**, color and other surface characteristics are not part of an object representation (e.g. Biederman, 1987)

However...

- Object naming is facilitated by **congruent surface color** and **photographic detail** as compared to line drawings (Price & Humphreys, 1989)

This holds particularly for...

- Structurally similar objects
- Objects with diagnostic color

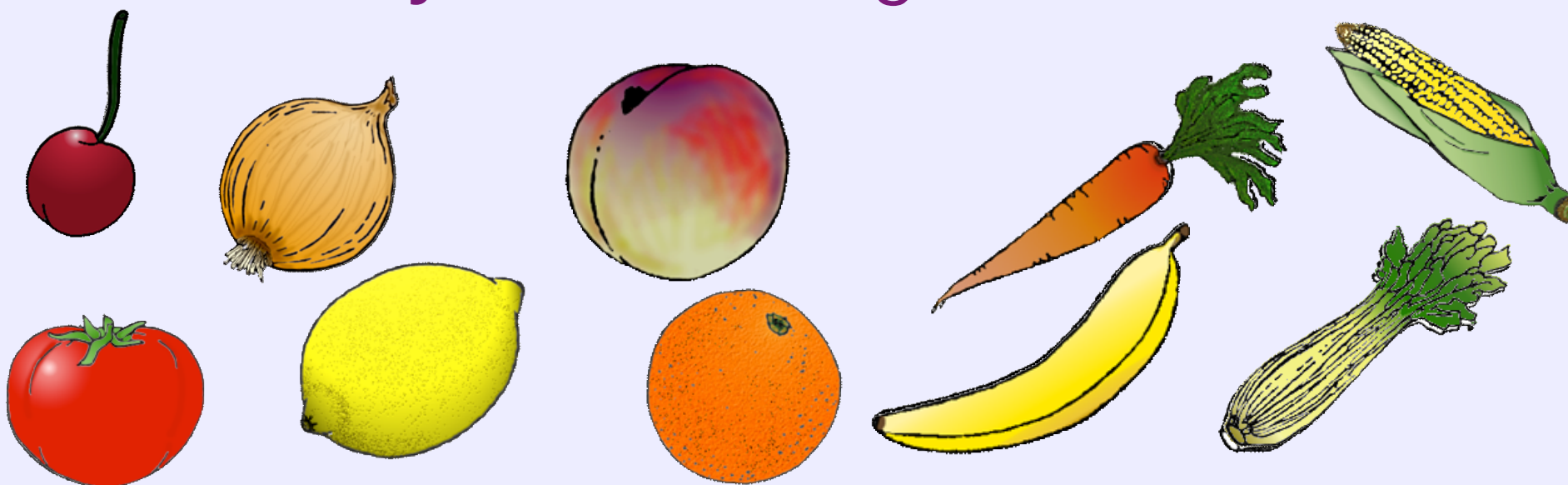


However...

- Object naming is facilitated by congruent surface color and photographic detail as compared to line drawings (Price & Humphreys, 1989)

This holds particularly for...

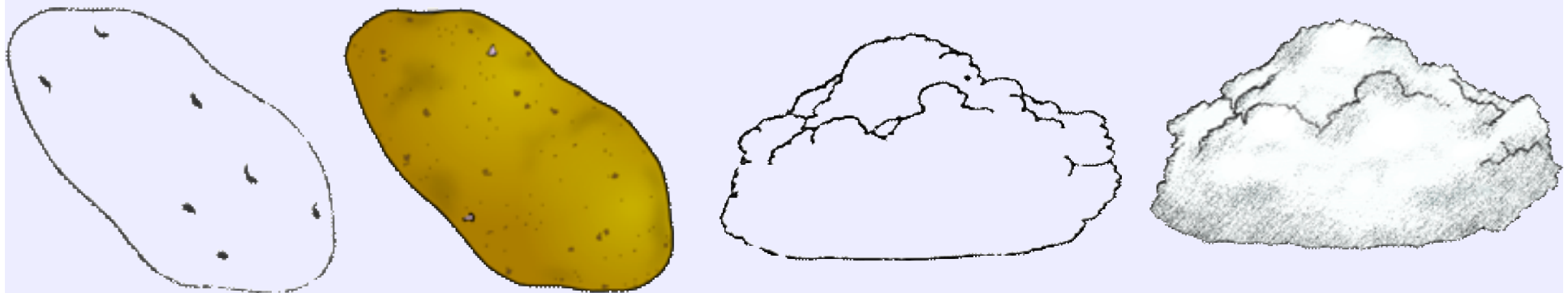
- Structurally similar objects
- Objects with diagnostic color



Object recognition is also facilitated by texture and color...

In normals, when objects have to be discriminated at the **subordinate level** or shapes are degraded through **occlusion** (Price & Humphreys, 1989; Wurm et al., 1993; Tanaka & Presnell, 1999)

In patients with **low-level vision** (Wurm et al., 1993) or **visual agnosia** (Mapelli & Behrmann, 1997; Chainay & Humphreys, 2001)

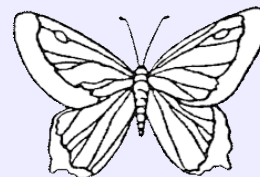


The present study:

- Testing the role of **surface information** in object recognition, on the largest set of common objects used in the literature
- Dissociating **the role of texture and color** in object recognition
- Providing **new sets of stimuli** for object recognition studies in normal and patients, with comparative **normative data** on these stimuli

Stimuli

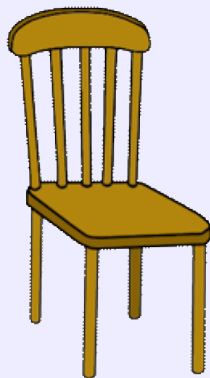
- **Scanning** of the 260 original drawings at high resolution (600 dpi) and “**cleaning**” of the images, all saved in 72 dpi on a white background (281 x 197 pixels).
- Careful **colouring** and **texture addition** by a professional graphist, using Adobe Photoshop 5.0 and color information from encyclopaedic books
- Two formats: 600 dpi for **high resolution prints out** and 72 dpi for **screen presentations**.
- All computerized images available in 3 types: **line drawings, grey levels, color.**



Subjects and tasks

- 240 students (age range 18-22)
- 60 subjects for each of the 4 tasks: naming, familiarity and complexity judgements, image agreement.
- 20 subjects in each condition (line drawings, grey level, color)
- Each task tested similarly as Snodgrass & Vanderwart (1980)

e.g.:



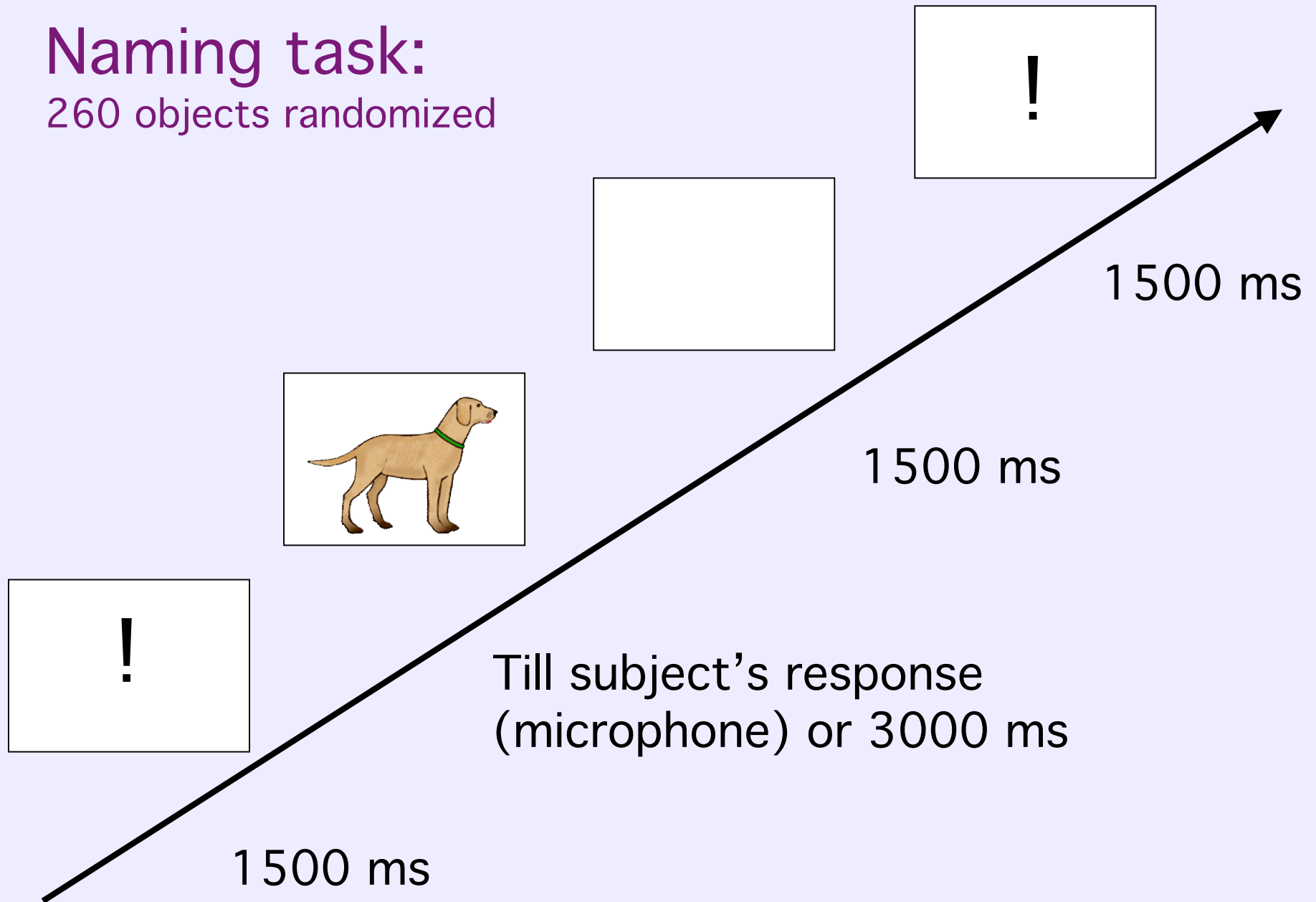
Unfamiliar



Familiar

Naming task:

260 objects randomized



Results: naming task

1. Accuracy rates (based on the most common name given): 88% (LD), 89.3% (gray), 90.7% (color), $p=0.001$)

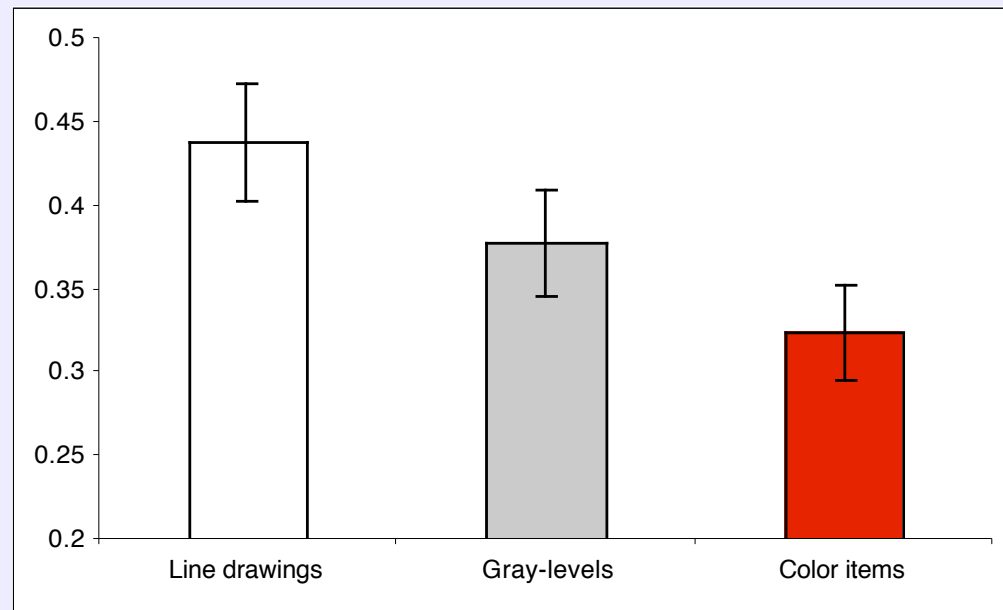
2. Overall agreement of subjects for the item names (H statistic, from Snodgrass and Vanderwart, 1980):

Lower values = higher agreement (H = 0, all subjects giving the same name)

Gray level vs. Line drawings: $p=0.01$

Colorized vs. Gray level: $p=0.01$

Colorized vs. Line drawings: $p<0.0001$

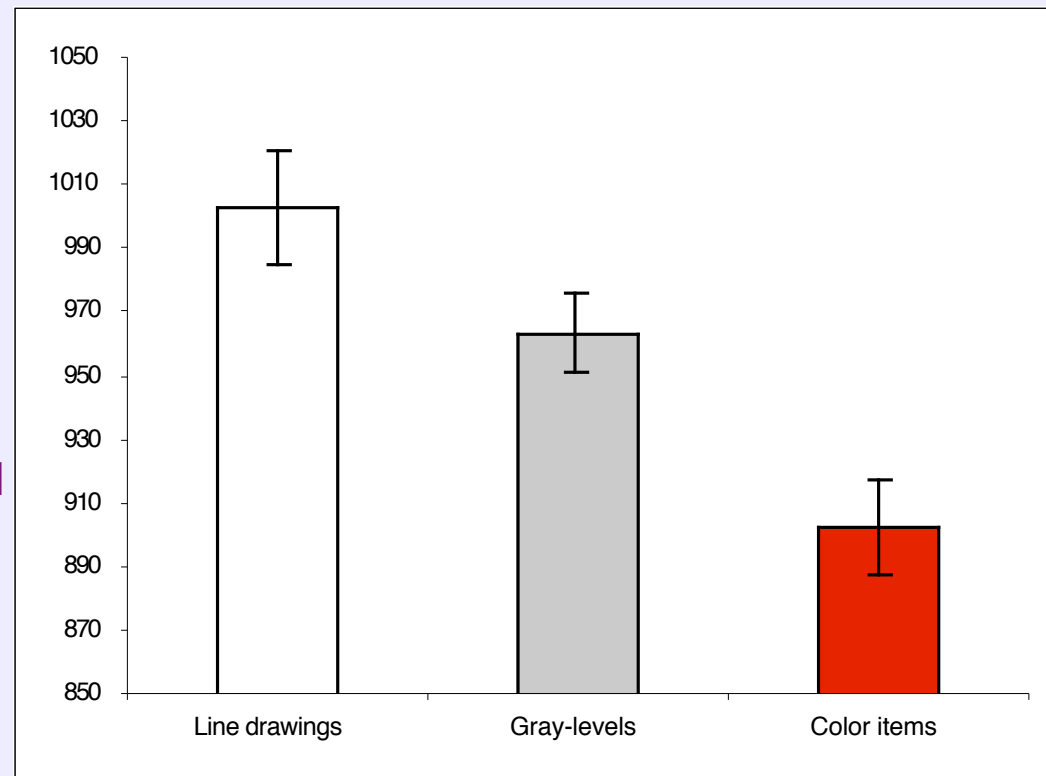


3. Correct response times: significantly different among the three conditions ($F_{2,259}=46.5$, $p<0.0001$)

Gray level vs. Line drawings: $p<0.001$

Colorized vs. Gray level: $p<0.0001$

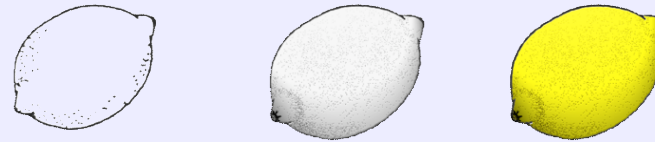
Colorized vs. Line drawings: $p<0.0001$



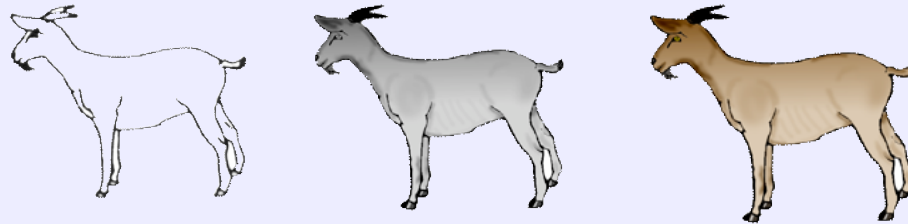
+ color, but not texture alone, further reduces naming RTs for the 40 items **named fastest as line drawings** only (color vs. line drawings: $p<0.01$).

4. Naming task: Analyses by categories

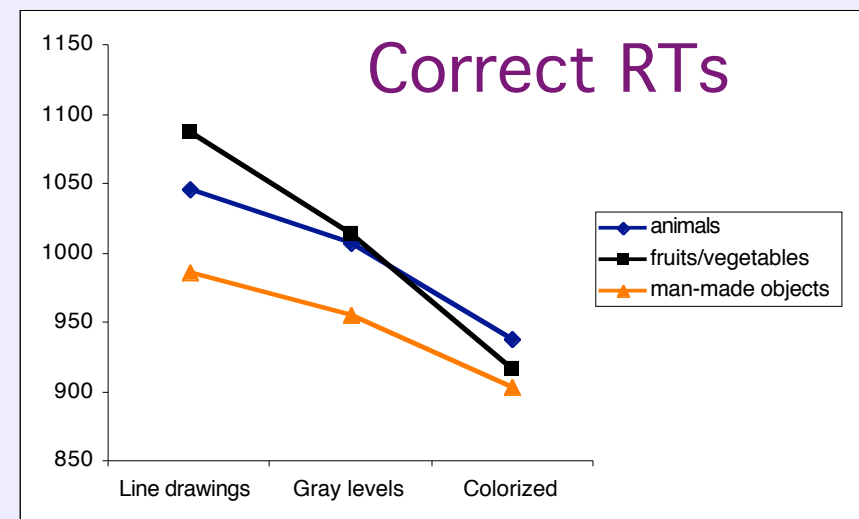
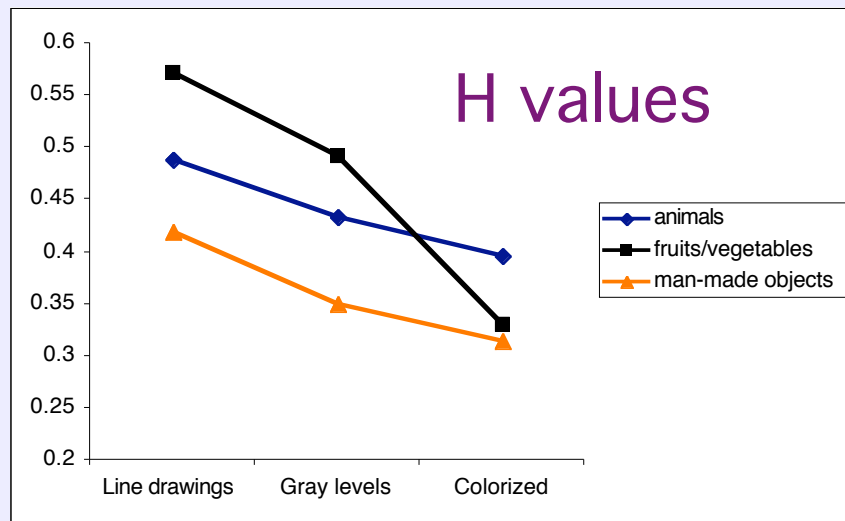
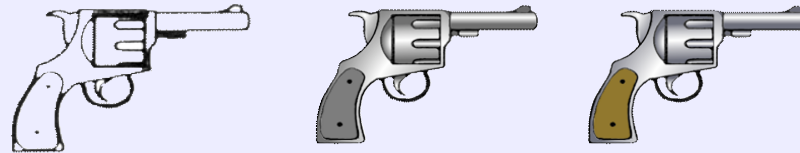
Fruits/vegetables (N=28)



Animals (N=53)

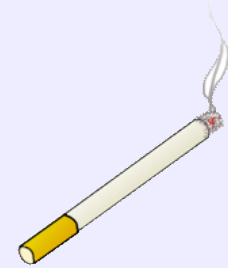
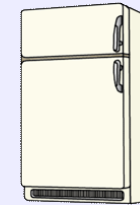


Man-made objects (N=158)

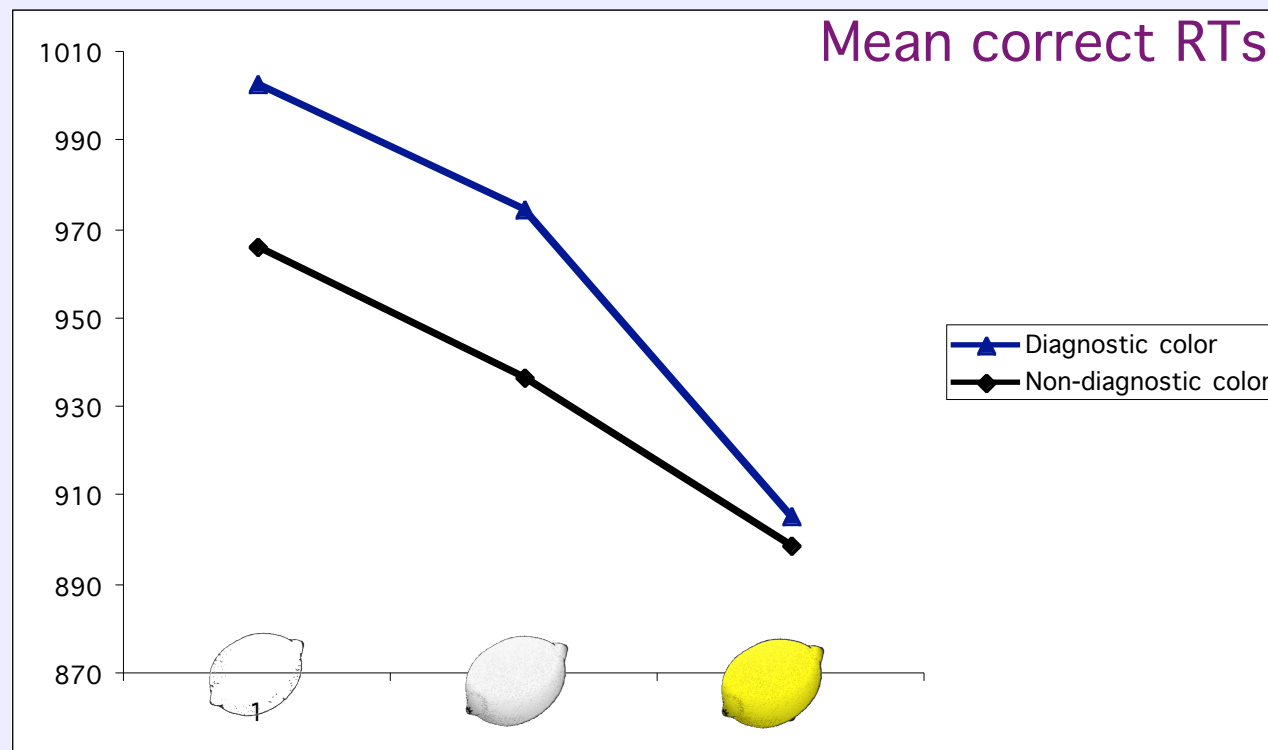
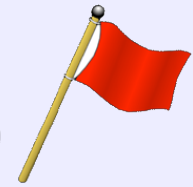
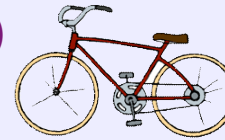


Color diagnosticity for man-made objects

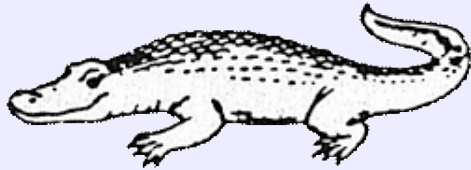
Objects with a diagnostic color (N=62)



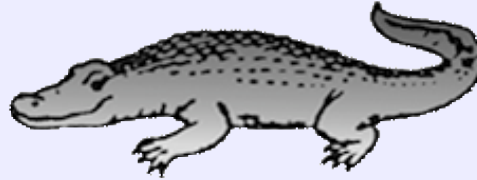
Object without diagnostic color at all (N=96)



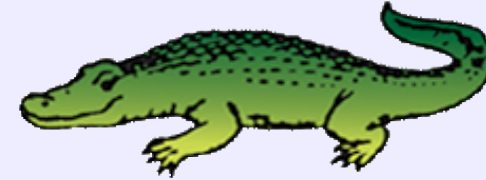
Familiarity norms (1-5 scale, “judge the object according to how usual or unusual the object is in your realm of experience”)



3.59 ± 1.11

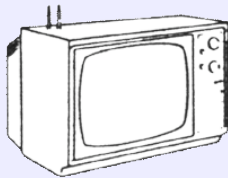


3.53 ± 0.94

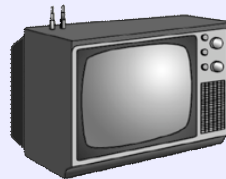


3.44 ± 0.71

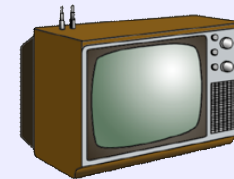
Visual complexity norms (1-5 scale, “judge the object according to the amount of detail or intricacy of line in the picture”)



2.76 ± 0.69



2.88 ± 0.87



2.70 ± 0.83

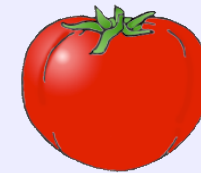
Image agreement norms (1-5 scale, object label given...then “judge the object according to how close it is to the object you imaged”)



3.73 ± 1.12



3.76 ± 0.98



3.74 ± 1.15

Conclusions

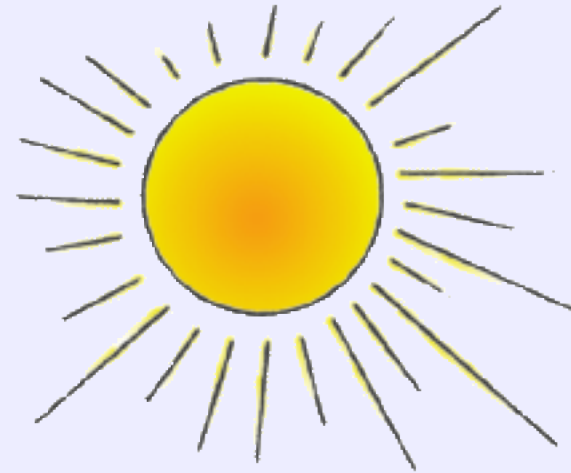
Texture and color contribute to object recognition for all categories of objects, including artefacts without any diagnostic color.

How?

- Better segmentation ? (perceptual contribution)
- Better recognition ? (“knowledge-based” contribution)

When **all** informations are available, objects are recognized at the same speed, suggesting that recognition of an object is based on **multiple cues**, with contour and surface information all part of an object representation and providing important information for recognition

There is a new set of 2D high quality pictures available for a large number of experiments involving object recognition, in both normals and patients



Special Acknowledgements to Olivier Clabots and Philippe Schynkus

Pictures available at www.md.ucl.ac.be/nefy/Face_Categorisation_Lab or www.cog.brow.edu/~tarr

Or e-mail: Bruno_Rossion@psp.ucl.ac.be