

Computer Vision

Working to help machines see...

What do you do in computer vision?

- In computer vision, we try to make machines/robots “see” like humans



You know what one great vision system is?

YOU

Ready to use your
visual wisdom and worldly knowledge
to help a robot “see” like you
?

Challenge 1:
help a robot
recognize a chair

Materials? Colors?



Parts? Shape?



What's in common?



What's in common?




Stand on
ground?







Forget the properties,
describe the affordances instead



what can a chair
be used for?

Surface, at a certain height, used for sitting...

Surface, at a certain height, used for sitting...

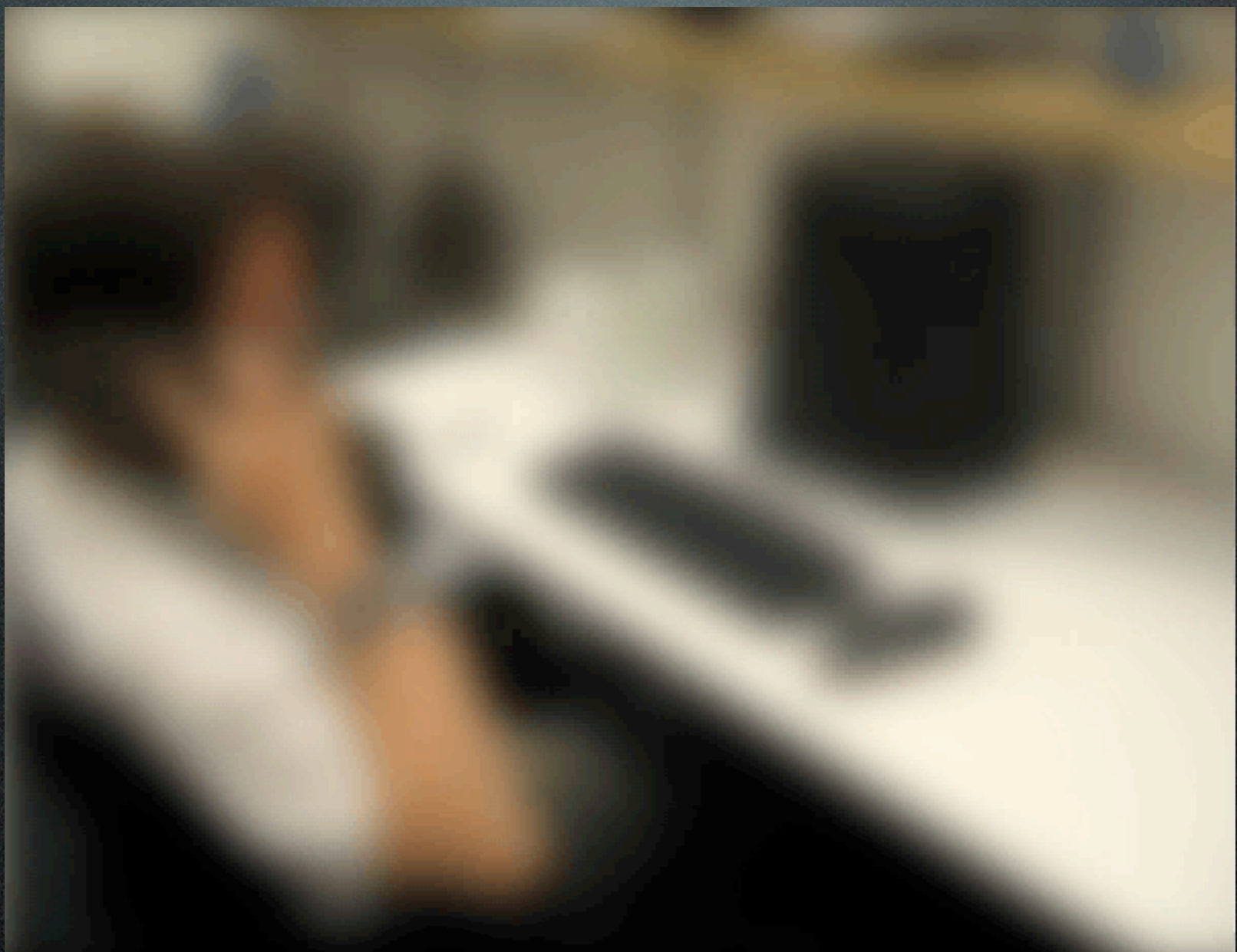


Surface, at a certain height, used for sitting...



Hopeless!

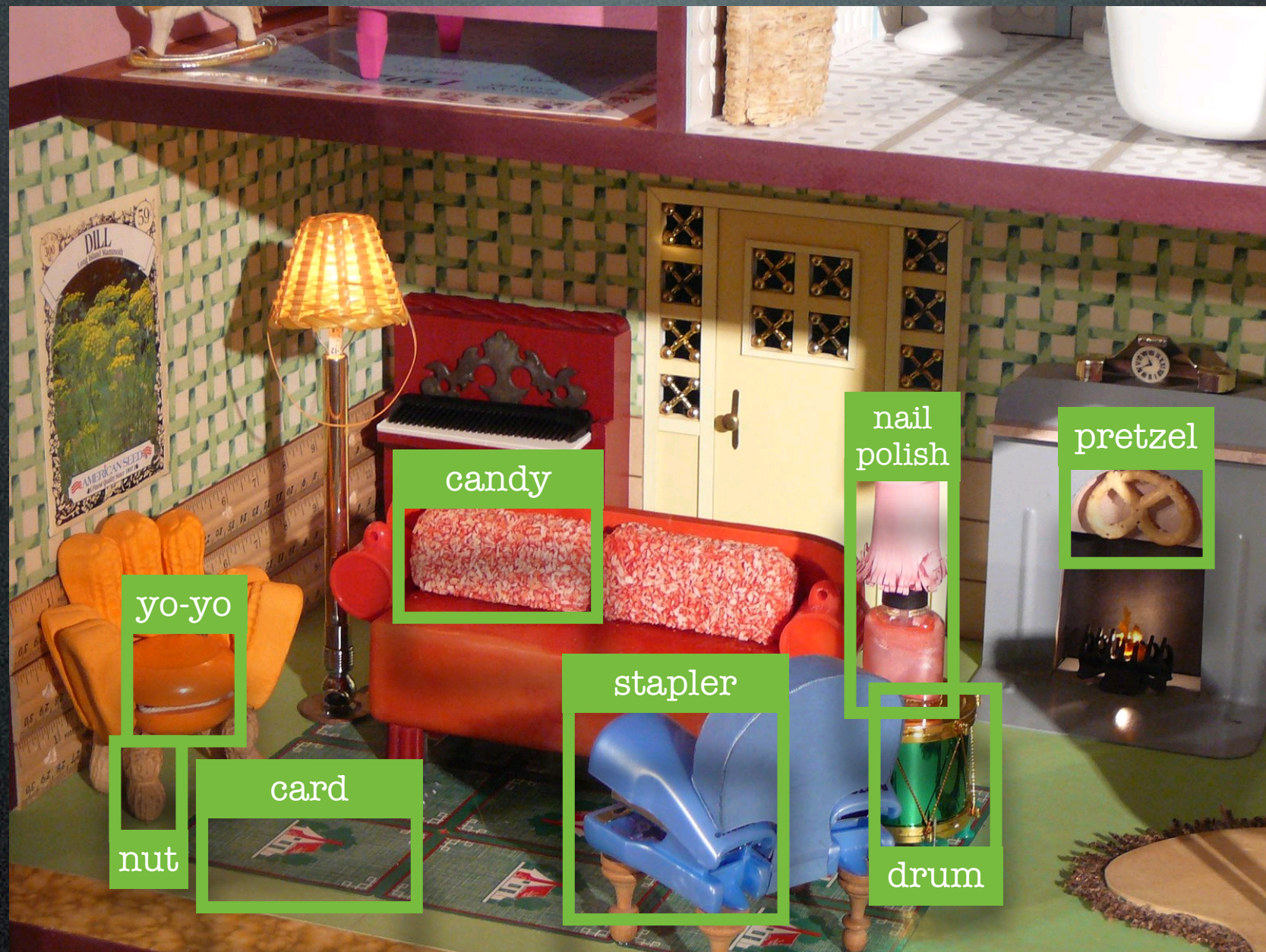
Challenge 2:
help a robot
understand a scene





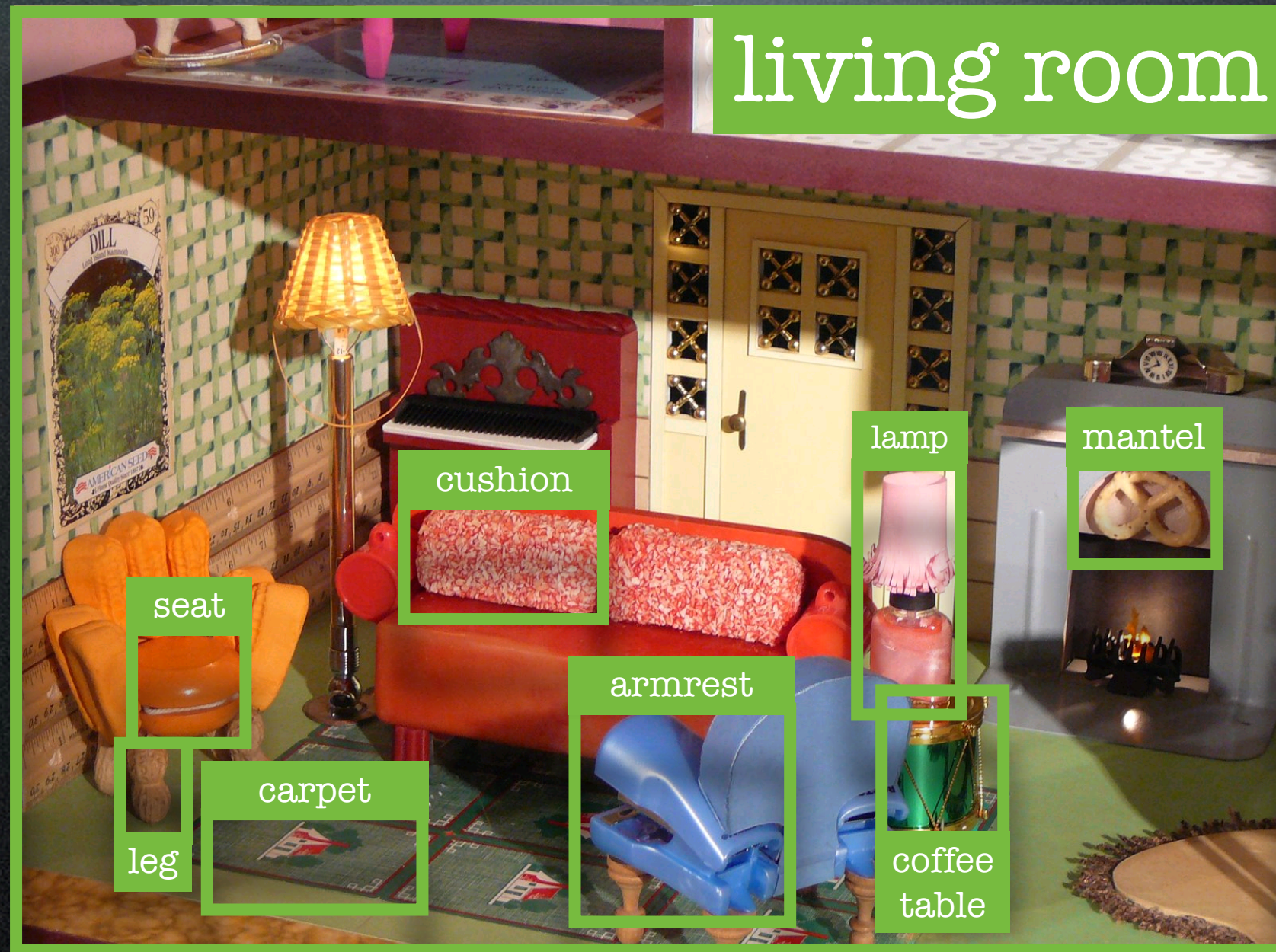


Joan Steiner



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Is the identity of objects context-dependent?




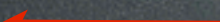


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Challenge 3:
help a robot
find animals

Find Wald...animals



Some of the challenges in computer vision

- classification/categorization  “what is it?”
(e.g. recognize a chair)
- detection  “find it”
(e.g. find animals in picture)
- scene/image understanding  “what is happening here?”
(e.g. understand a scene, describe the elements, understand the relationships and actions)
- ... 
 - reconstruct a 3D scene from 2D images
 - recognize the emotion on the face
 - split the images into its components
 - etc.

But it's not so hopeless
after all!

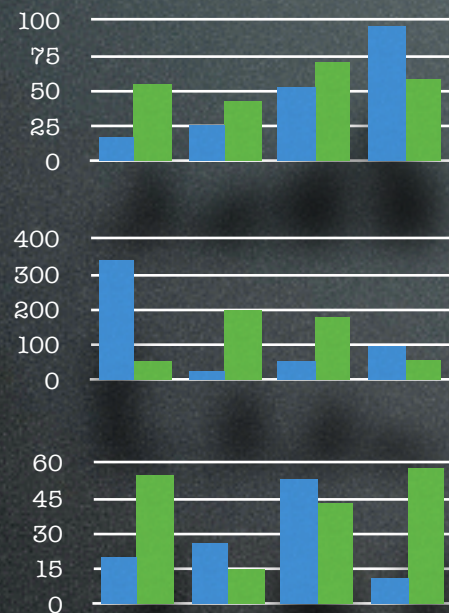
How do you do it?

Let's recognize a chair:

input: image



feature extraction



a bit of magic



output: label



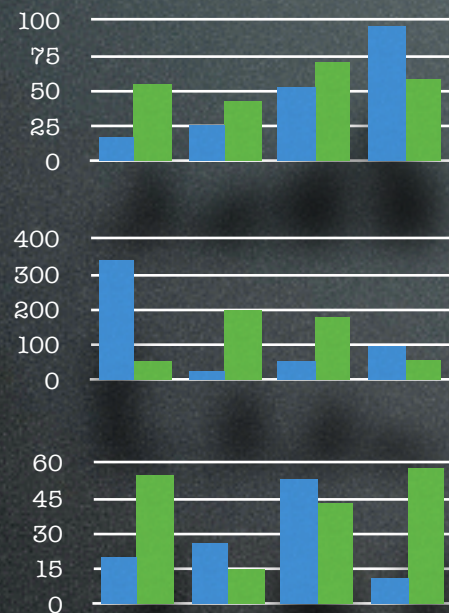
a typical computer vision pipeline

Let's recognize a chair:

input: image



feature extraction



a bit of magic



output: label



+ ASSUMPTIONS

As you saw before, the general computer vision task is too hard

So, let's assume our task is not to recognize chairs in the general sense, but to tell apart wooden chairs from glass wine bottles

Class discrimination



can we find a simple way
to tell them apart?

Feature extraction

example: color



■ chair

■ bottle

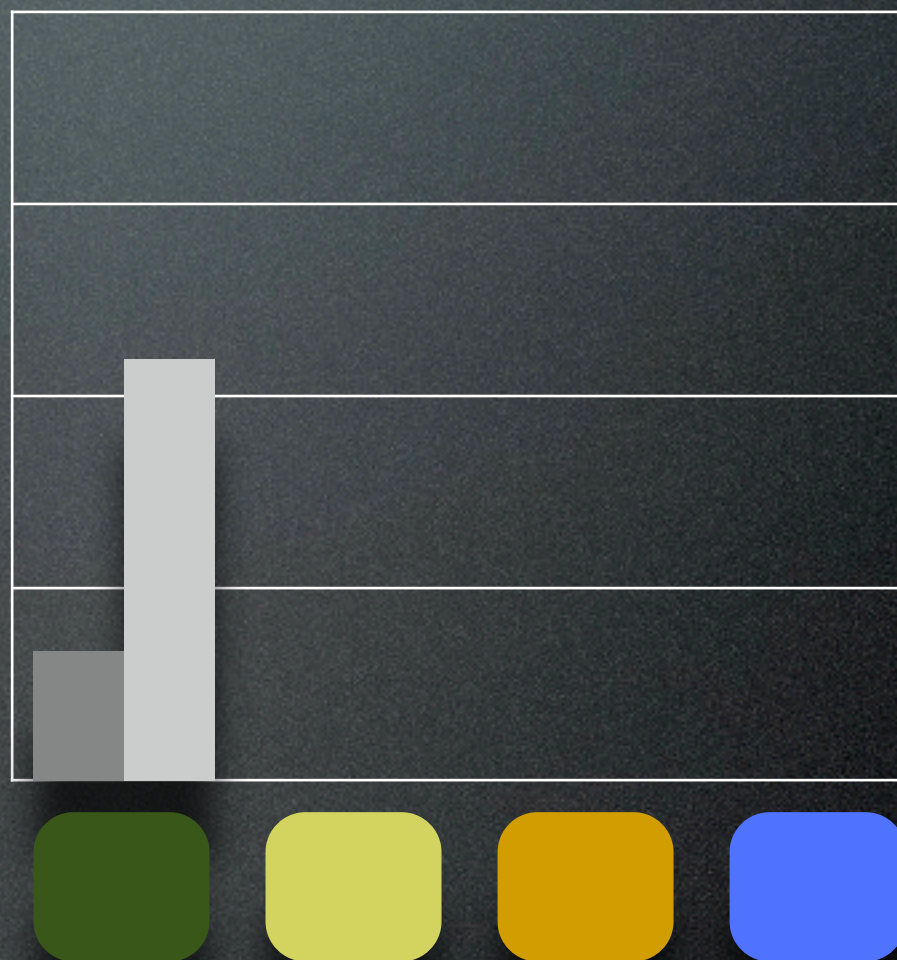


Feature extraction example: color



chair

bottle



Feature extraction example: color



chair

bottle



Feature extraction example: color



chair

bottle



Feature extraction example: color



Collect more data

wooden chairs



glass wine bottles

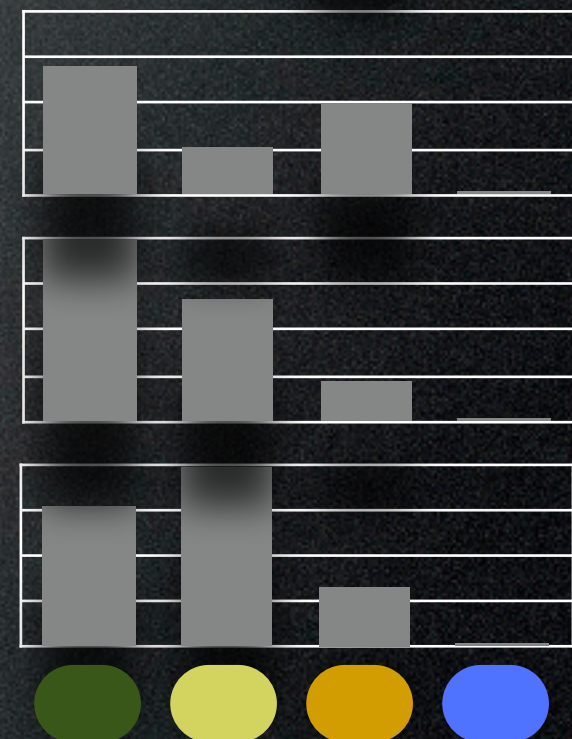


Compute features

wooden chairs



glass wine bottles

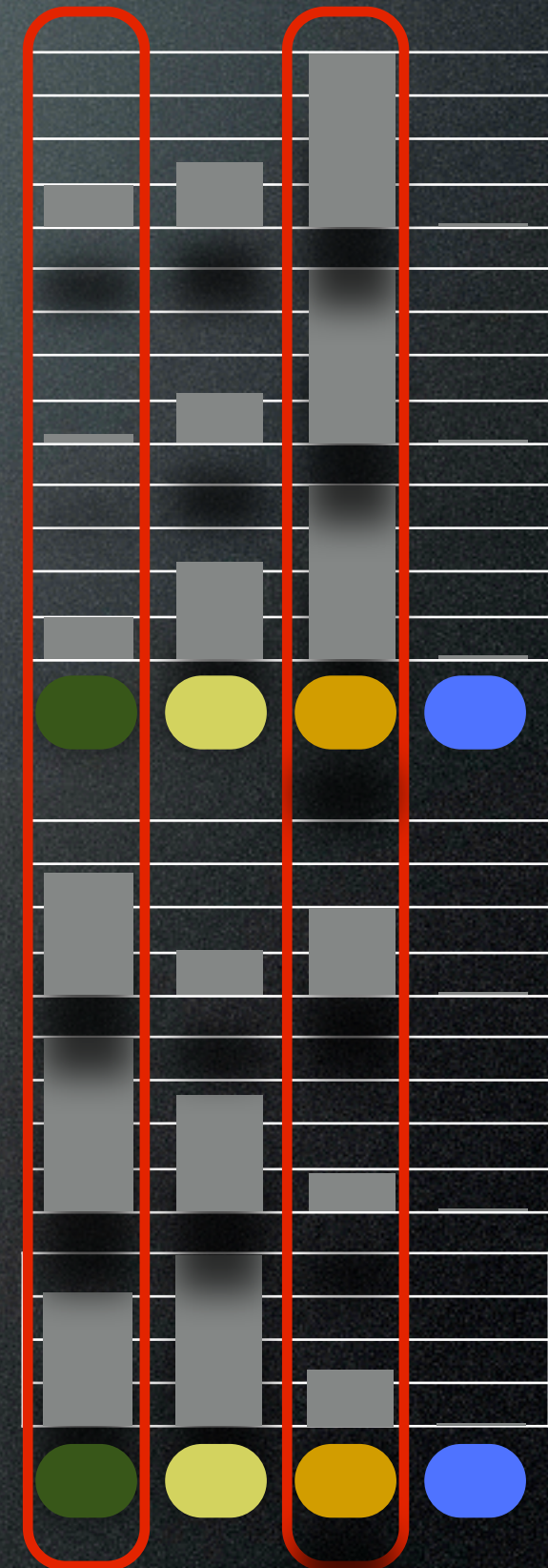


Discover discriminative features

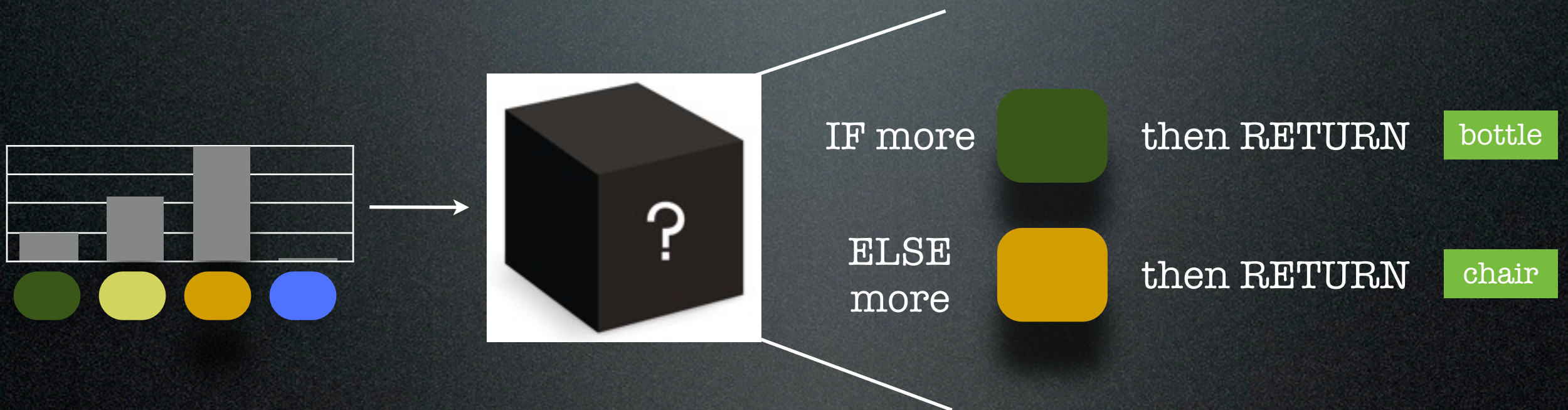
wooden chairs



glass wine bottles

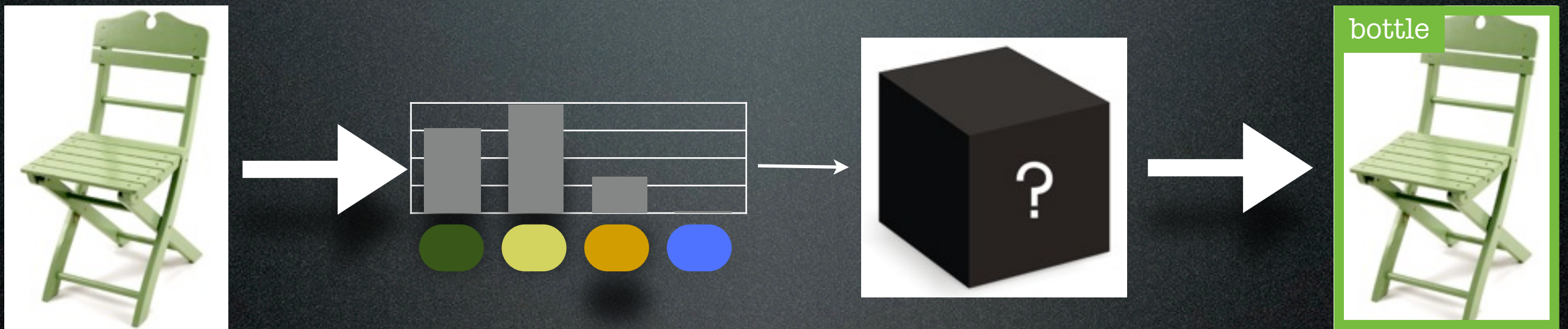


Peek inside the black box:



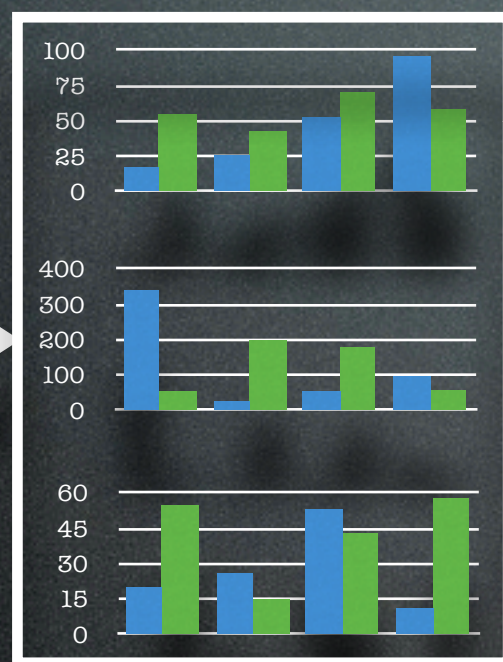
Let's test our system...

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Back to the drawing board...

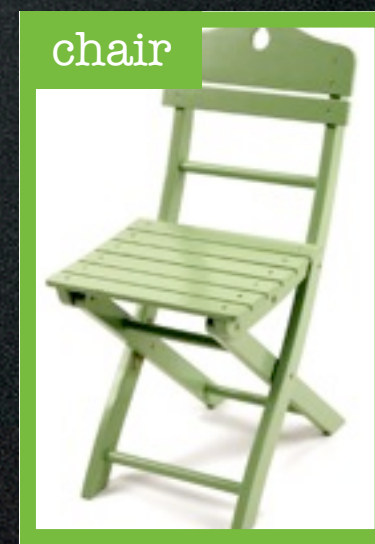
More features!



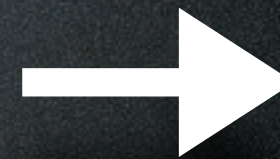
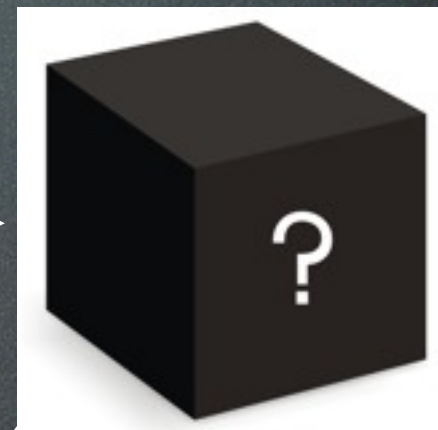
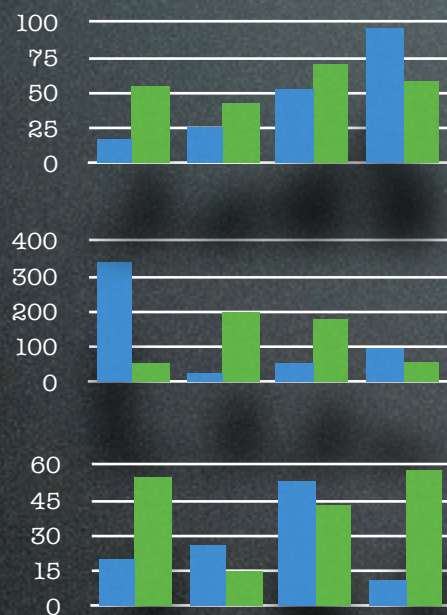
materials
textures
gradients
corners
surfaces

...

...



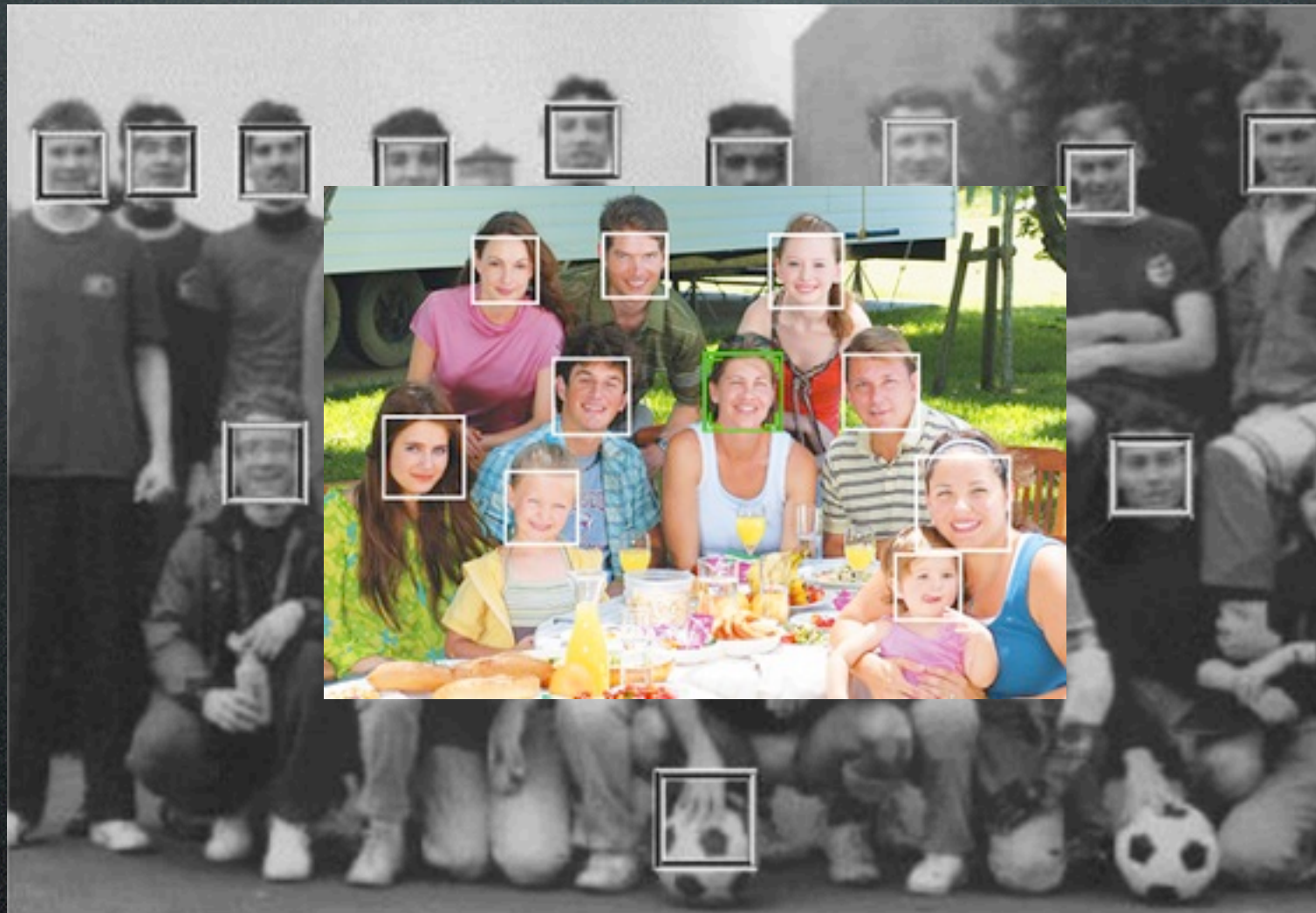
More data!



Computer vision is hard
but possible!

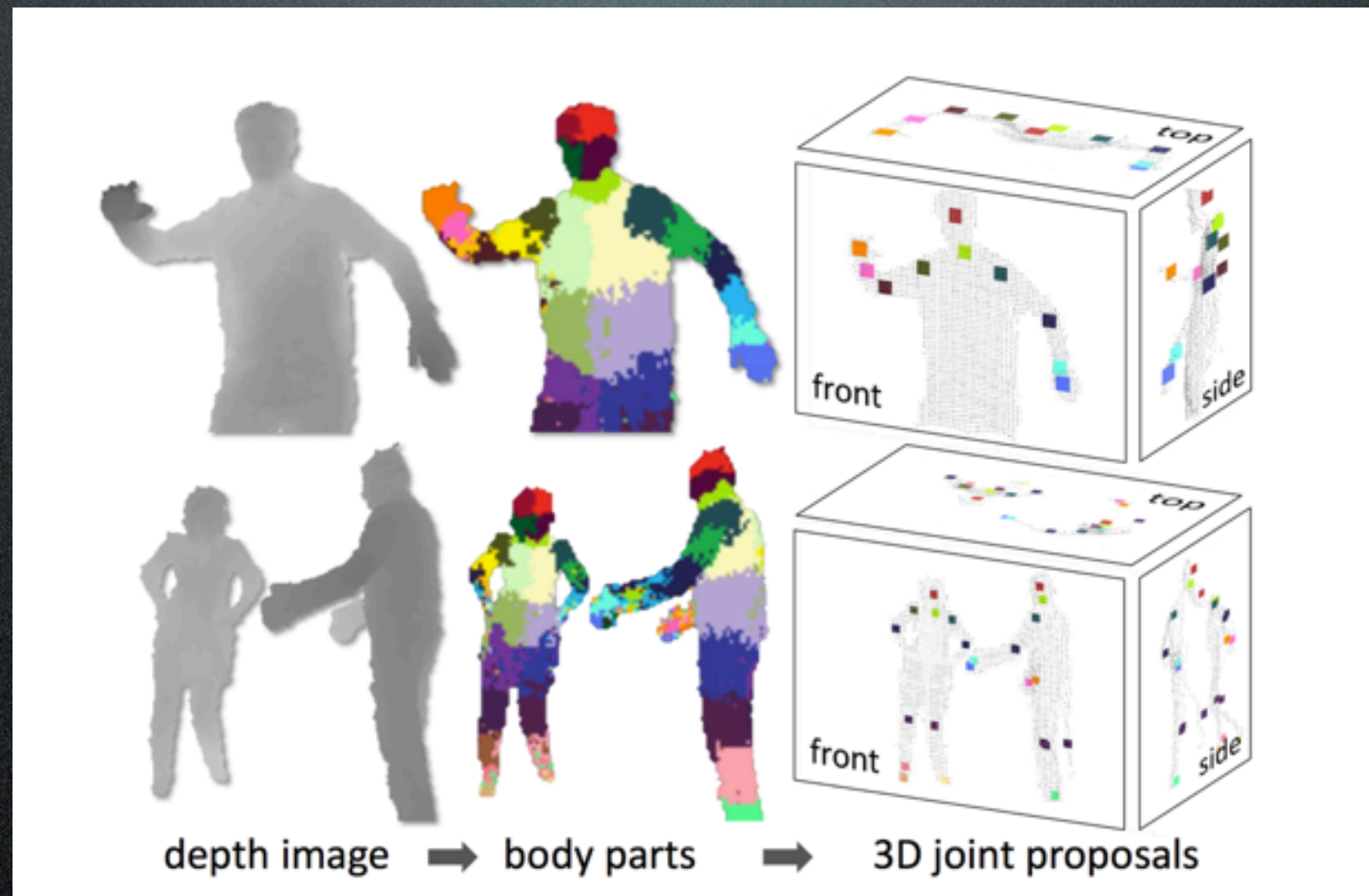
Let's look at some
successes...

Face detection



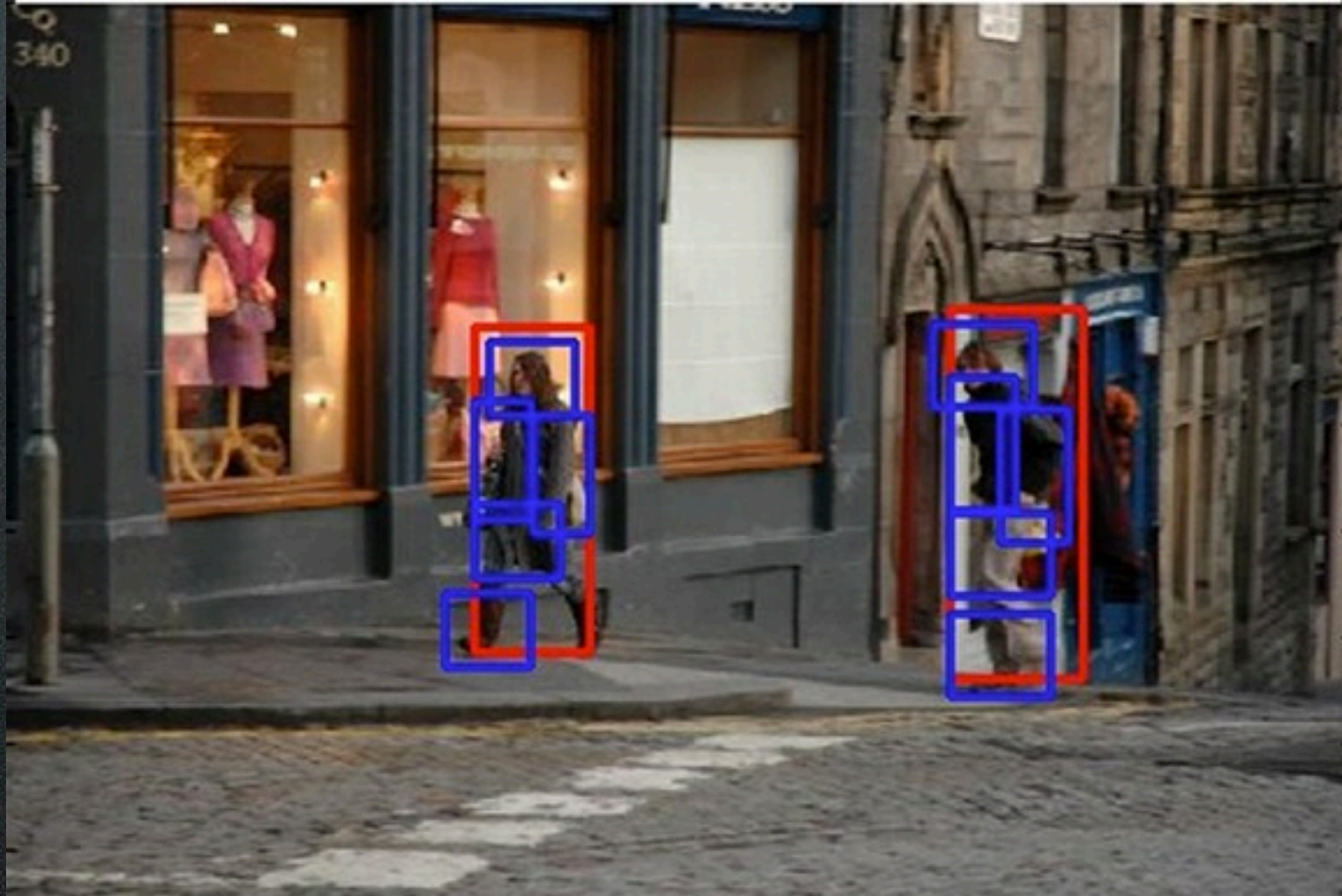
Viola & Jones (2001)

Body part recognition



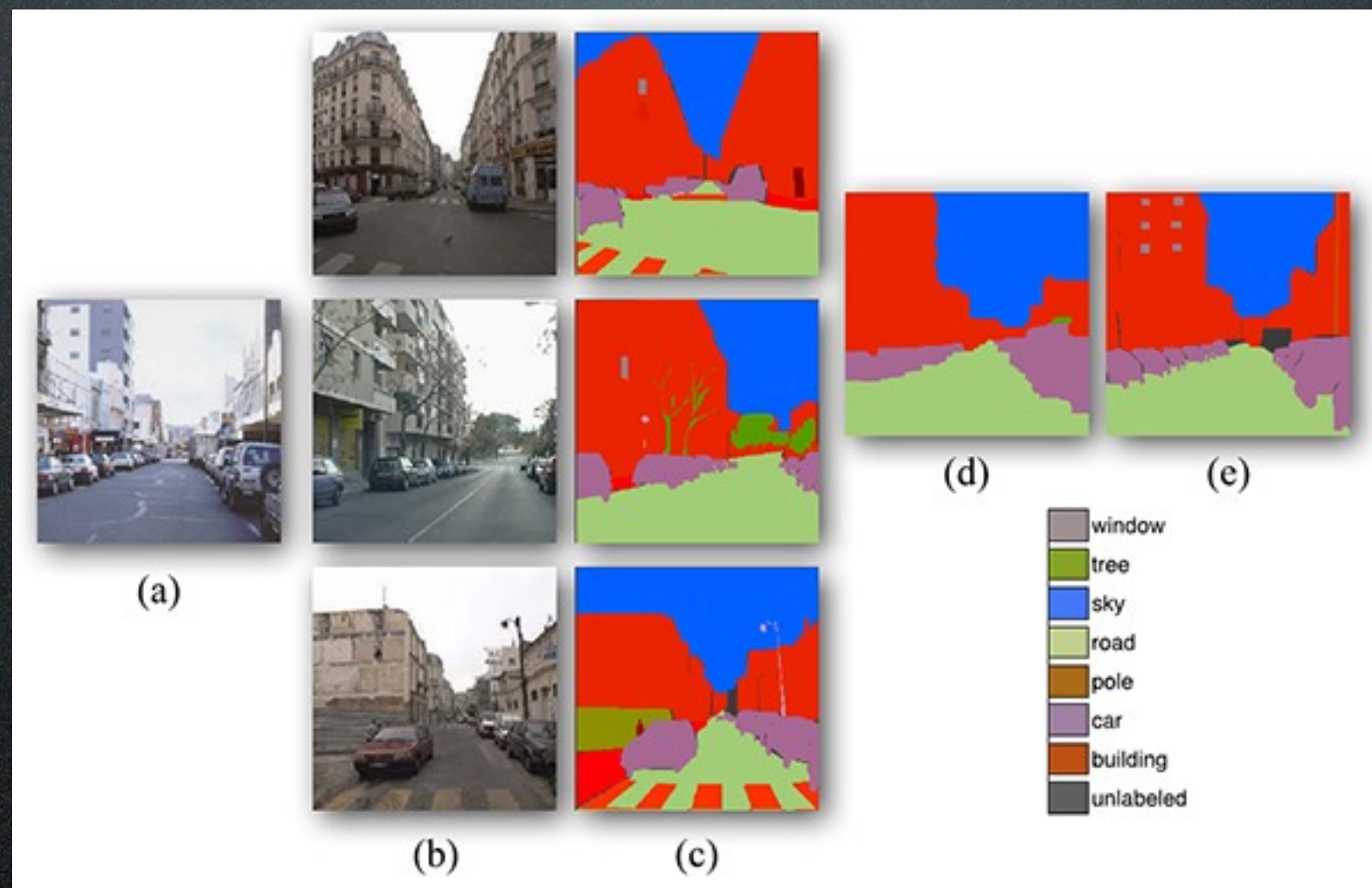
Microsoft's Kinect (2011)

Pedestrian Detection



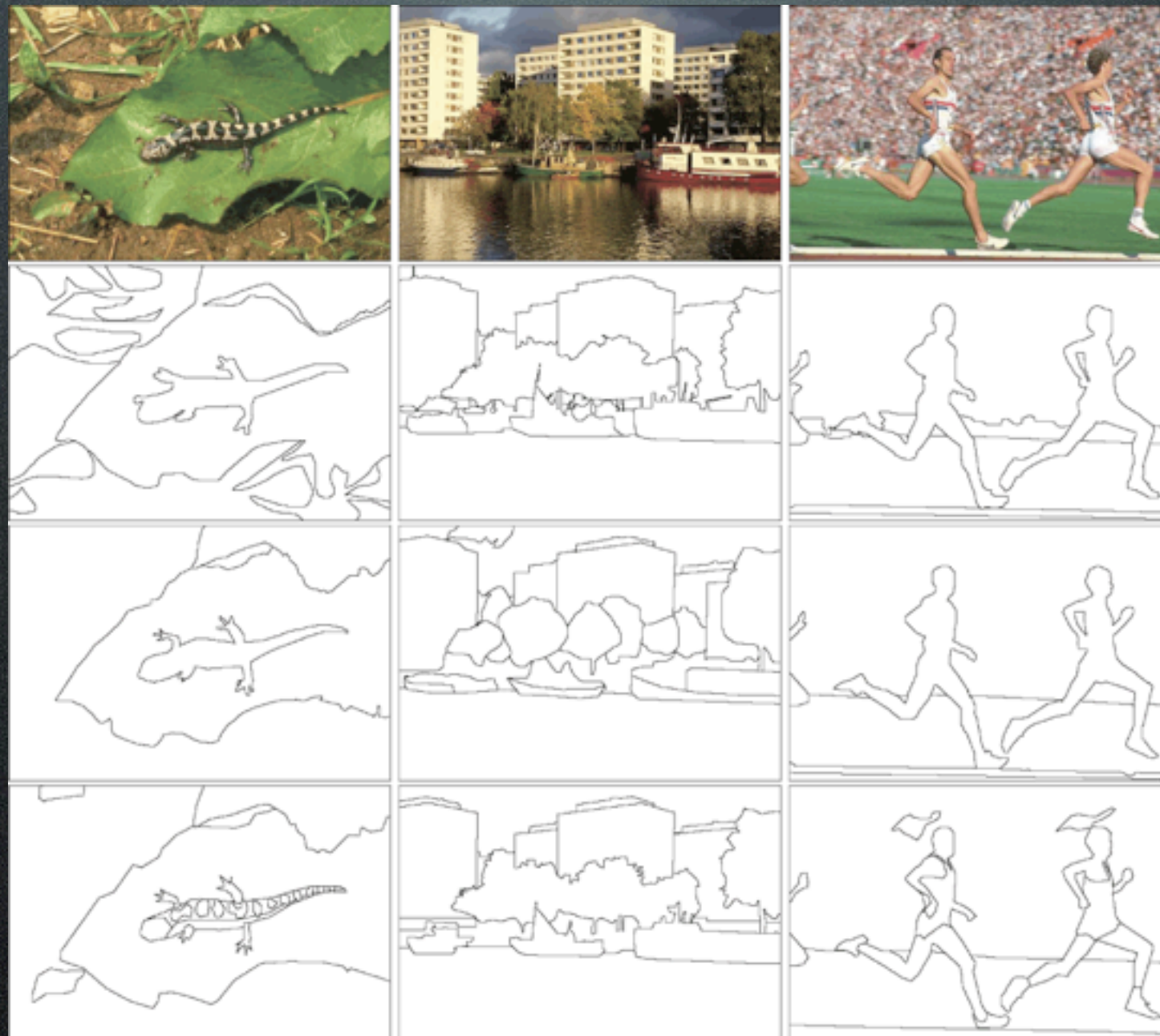
Felzenszwalb, McAllester, & Ramanan (2008)

Scene Parsing



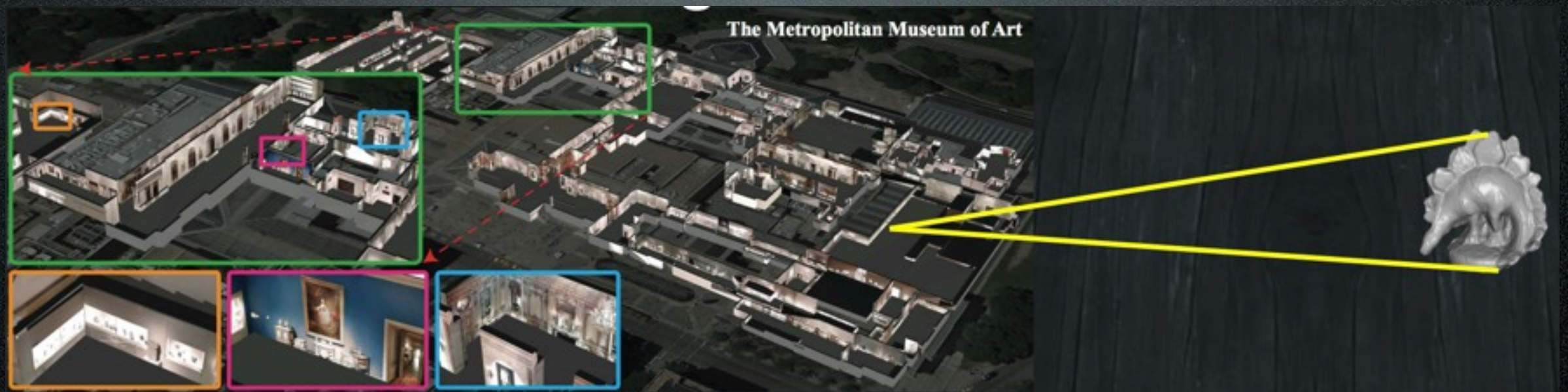
Liu, Yuen, & Torralba (2011)

Image Segmentation



Berkeley Segmentation Dataset (2001)

3D Reconstruction



Xiao & Furukawa (2012)

Predicting where people look



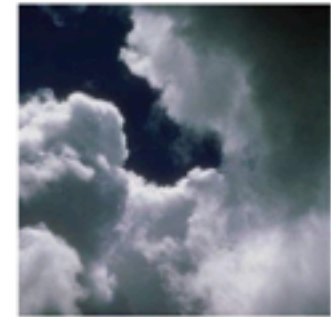
Judd, Ehinger, Torralba, & Durand (2009)

Predicting Image Memorability

Memorable



Average



Forgettable



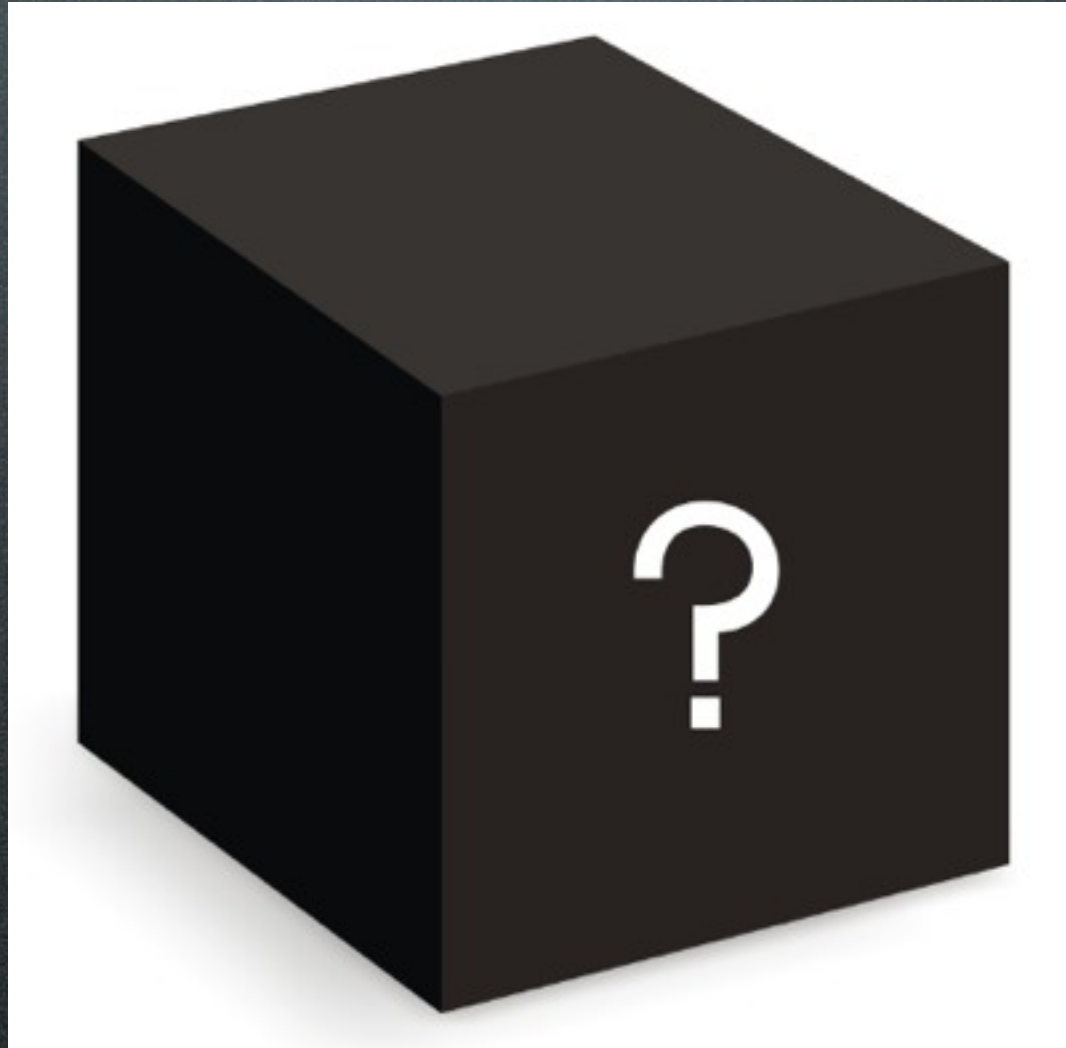
Isola, Xiao, Torralba, & Oliva (2011)

Coming soon!



Google Glass

The next great black box:



Designed by YOU

THANK YOU :)

THANK YOU :)

Questions?