

# Art, Vision, Probability

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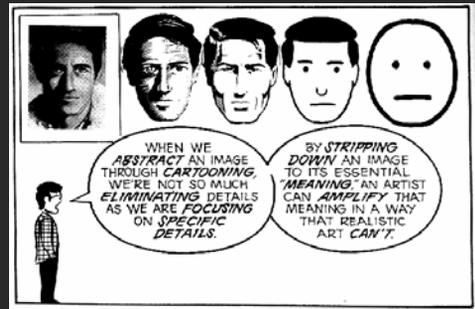
## What is this?



Why does the paint stroke look so much like a leg?  
Why does it look good?



How can a photograph that's so abstract be so expressive?



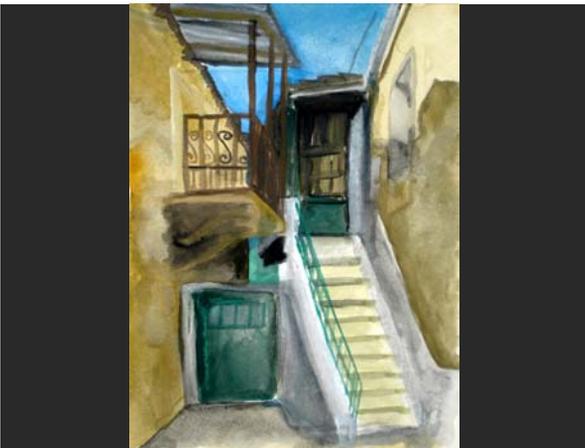
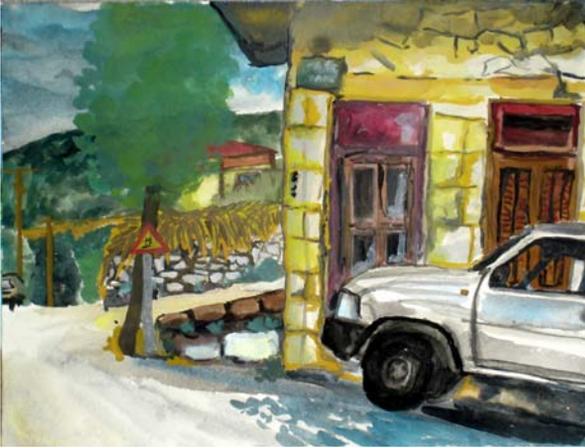
S. McCloud, *Understanding Comics*



## The abstraction paradox

The HVS is designed to interpret the world  
We give it something totally unlike the world  
Yet we perceive it as *more* expressive ... why?





## Questions

1. Why does abstraction work?
2. What makes an artwork successful?
3. How can we create a computational model for art and photography?

# 1. Vision and probability

## “The vision problem”

Given light measurements, infer properties of the world

Problems:

- huge set of unknowns
- limited measurements
- indirect link to unknowns

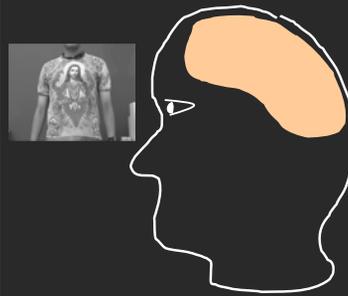
## Reconstruction from images



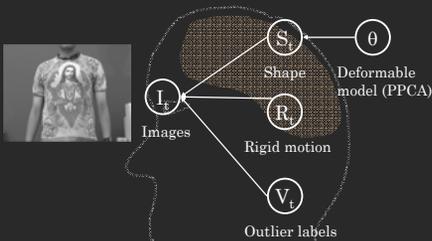
Input video

with: L. Torresani

## Non-rigid structure-from-motion

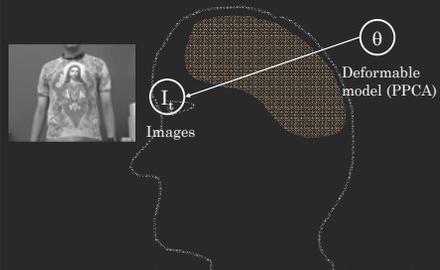


## Non-rigid structure-from-motion



Problem: Given images, solve for everything else  
... underconstrained: determine both  $\theta$  and  $S_t$   
humans can do it from dots alone (Johansson 1973)

## Non-rigid structure-from-motion



Approach: Optimize *marginal* distribution of  $\theta$

## Reconstruction from images



Reference frame

Raw video

Lucas-Kanade

Robust algorithm

3D reconstruction

## Moral for computer vision

Compute marginals to factor out uncertainty

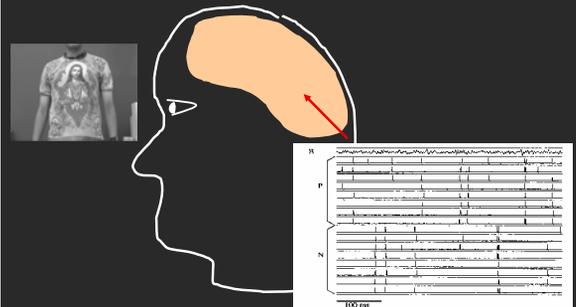
Recognition and reconstruction must be done jointly!

c.f. W.T. Freeman, A. Yuille, S.C. Zhu, ...

## 2. Human visual processing

### Visual encoding

Goal: scene representation to guide actions



## Inference must be *robust*



## Robustness to uncertainty



## Robustness to limited cues

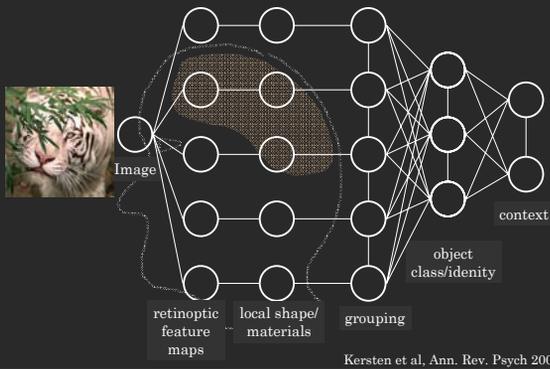


## Neural processing is (probably) probabilistic

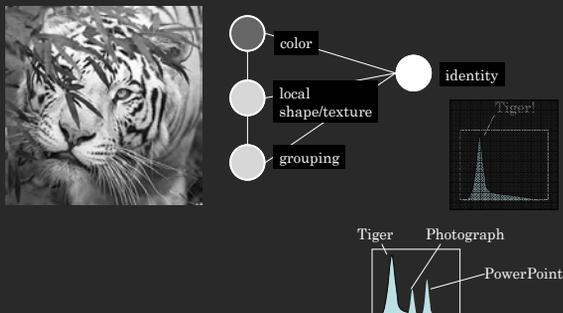
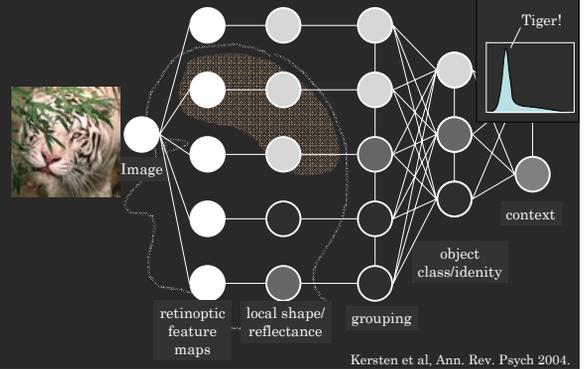
See, for example:

- Rao and Olshausen (eds), *Probabilistic Models of the Brain*, 2002
- Hinton, "What kind of graphical model is the brain?" IJCAI 05

## Visual processing as Bayesian inference



## Visual processing as Bayesian inference



## 3. Art, vision, probability



### Visual ambiguity “illusions”

object identity

Face People

local surface features

### Representation with repeated primitives

object identity

Face Blobs

local surface features



### Creating a visual language

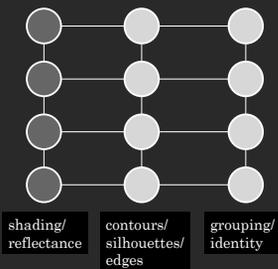
Each interpretation has a compact *code*  
 Efficient models have high probability in model selection [MacKay]



## Composition



## Why does line drawing work?



shading/  
reflectance

contours/  
silhouettes/  
edges

grouping/  
identity

## Segmentation and grouping



## Summary

- The HVS infers scene properties by probabilistic inference
- Art works convey a scene by providing sufficient cues to the visual system
- Ambiguity, economy, and visual languages all correspond to specific probabilistic interpretations
- General-purpose computer vision may require a similar approach

## Open questions

1. Can we optimize an image (painting or photo) for a desired effect on the HVS?
2. What kinds of internal representations are useful for modeling art and HVS?
3. What does this have to do with goals of art?
4. What kinds of data do we need from our capture devices and vision algorithms?