

Correlated Predictors,
words
and scenes
or

What should we mean by “scene”?

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What are scenes?

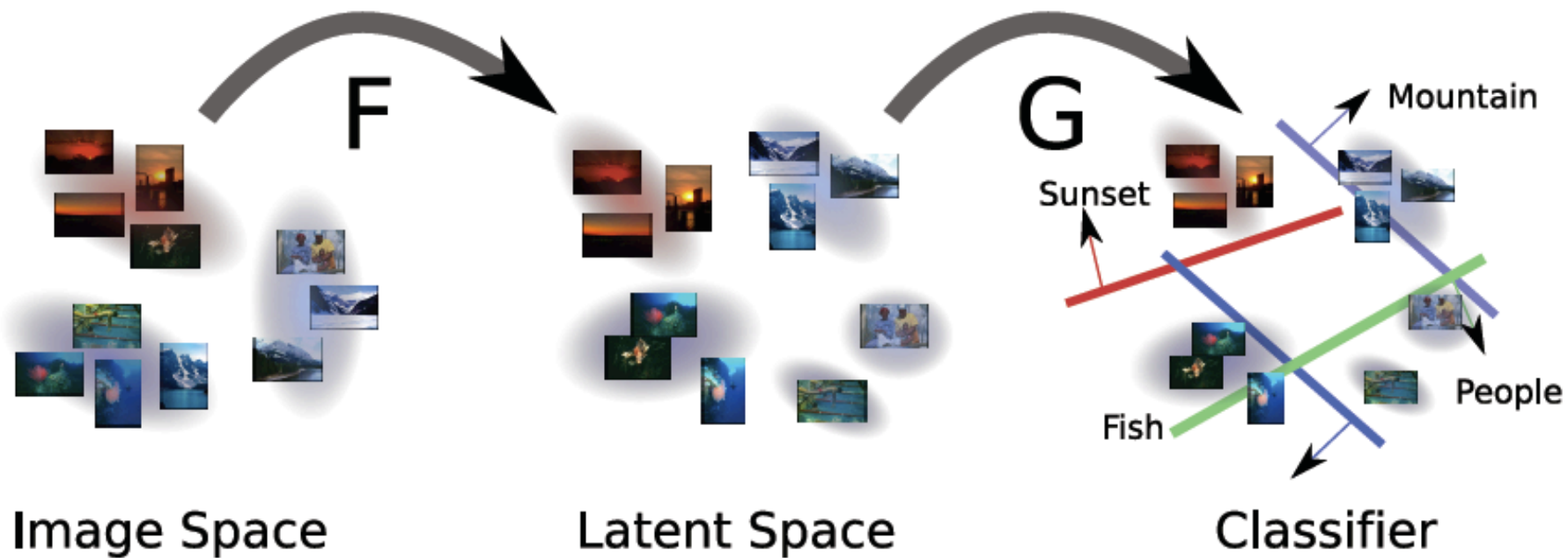
- A convenient bag in which helpful context lives
 - Objects that tend to want to appear together
 - Geometric “stages” on which objects appear
 - Illumination fields in which objects are immersed

Scenes as object bags

- We could build collections of labelled scene images
 - useful, but..
 - kitchen, bathroom, outdoor, and then?
- We could collect images of similar appearance
 - but..
 - might not really have similar objects in them
- Unsupervised bag discovery
 - Pictures of the same scene tend to contain similar objects
 - i.e. tend to attract the same image annotations

Word prediction

- Simple method:
 - rack up some features, build a bunch of linear classifiers one per word
 - works poorly
 - few examples per word
 - many features, only some are stable
- Idea
 - some features are not helpful
 - a low dimensional subspace is good at predicting most things (Ando +Zhang,)
 - We can find this space by penalizing rank in the matrix of linear classifiers

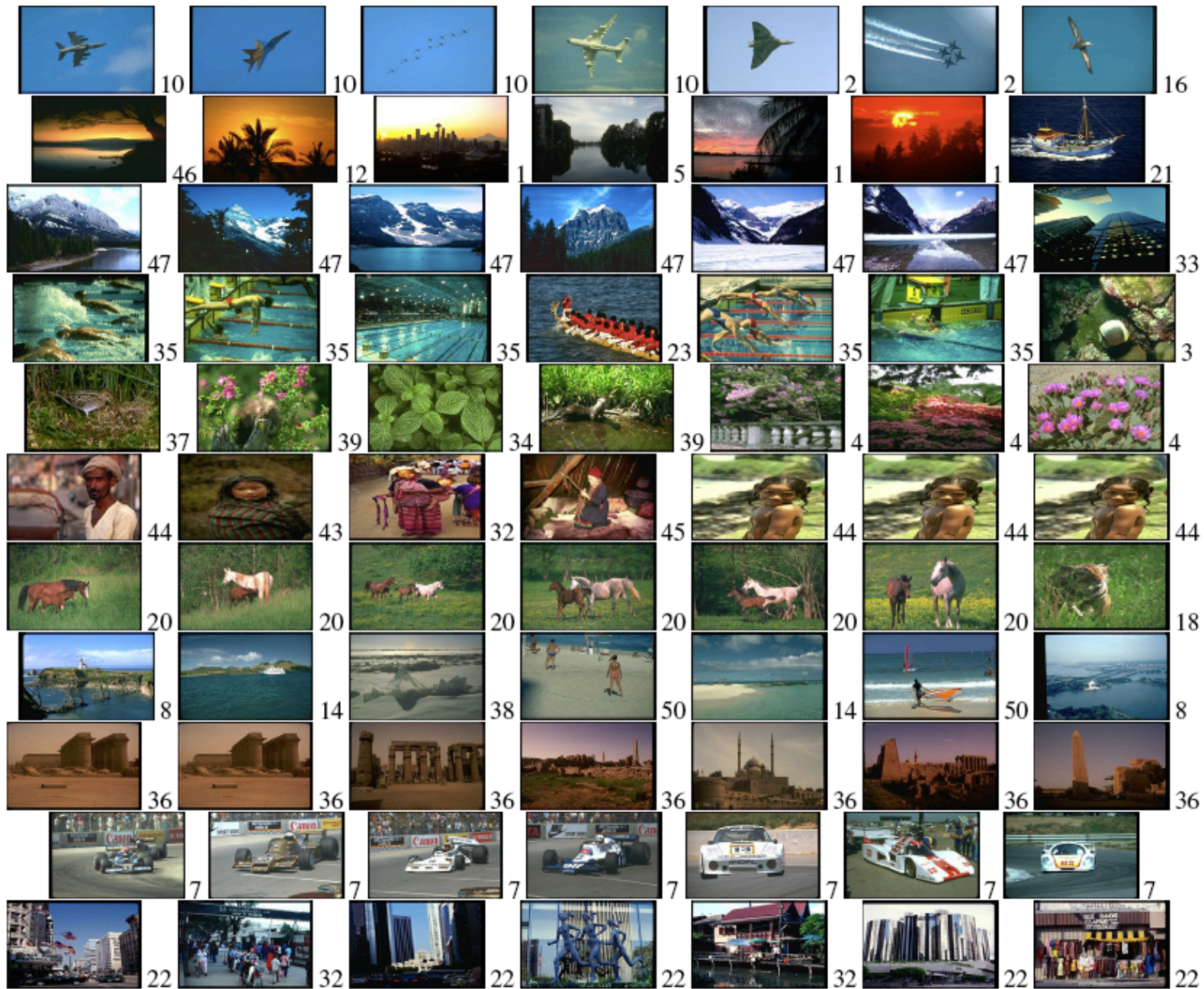


Cluster points in the latent space

Close together if they predict similar words

This means we get clusters of things that look similar in ways that support word prediction

Scene



CD #
(rough proxy)

How should we describe scenes?

- We probably don't know the names of each type
 - Can't just
 - build a kitchen detector, bedroom detector, etc.
 - Like objects
- Want to:
 - make useful statements about scenes whose names aren't known
 - say how a scene of known type is different from others of that type

Attributes

- Desiderata suggest a fluid notion of scene category
 - in terms of useful properties rather than names
 - new scene types are new collections of properties
 - not all required to match (exemplar theory originated here)



Has Beak, Has Eye, Has foot, Has Feather



'is 3D Boxy'

'is Vert Cylinder'

'has Window' ~~'has Screen'~~

'has Row Wind' ~~'has Headlight'~~

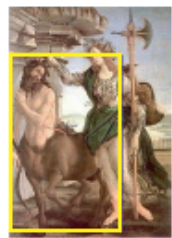


'has Hand'

'has Arm'

'has Plastic' ~~'has Saddle'~~

'is Shiny'



'has Head'

'has Hair'

'has Face'

~~'has Saddle'~~

'has Skin' ~~'has Wood'~~



'has Head'

'has Torso'

'has Arm'

'has Leg'

~~'has Wood'~~



'has Head'

'has Ear'

'has Snout'

'has Nose'

'has Mouth'



'has Head' ~~'has Furniture Back'~~

'has Ear' ~~'has Horn'~~

'has Snout' ~~'s Screen'~~

'has Mouth' ~~'has Plastic'~~

'has Leg' ~~'is Shiny'~~



~~'has Furniture Back'~~

~~'has Horn'~~

~~'s Screen'~~

~~'has Plastic'~~

~~'is Shiny'~~



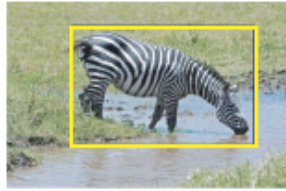
'is 3D Boxy'

'has Wheel'

'has Window'

'is Round'

'has Torso'



'has Tail'

'has Snout'

'has Leg'

~~'has Text'~~

~~'has Plastic'~~



'has Head'

'has Ear'

'has Snout'

'has Leg'

'has Cloth'



'is Horizontal Cylinder'

~~'has Beak'~~

~~'has Wing'~~

~~'has Side mirror'~~

'has Metal'



'has Head'

'has Snout'

'has Horn'

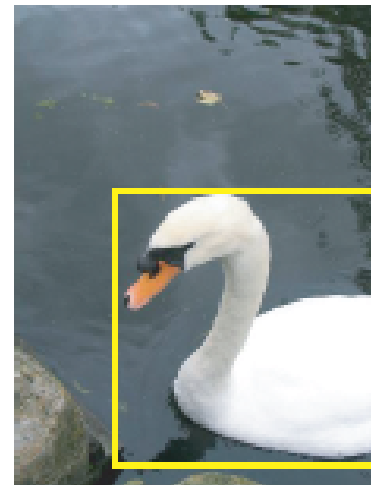
'has Torso'

~~'has Arm'~~

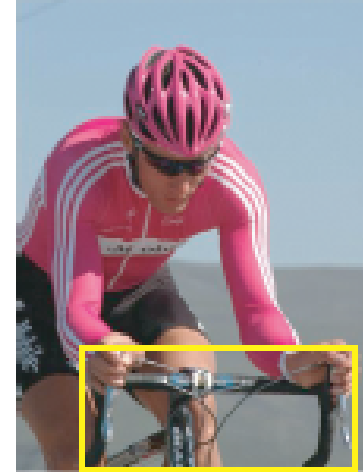
Missing attributes



Boat
No "sail"



Bird
No "tail"



Bicycle
No "wheel"

Extra attributes



Bird
"Leaf"

But what should the properties be?

- Things that are good at helping predict where objects are
- And what they are like

Scenes as stages

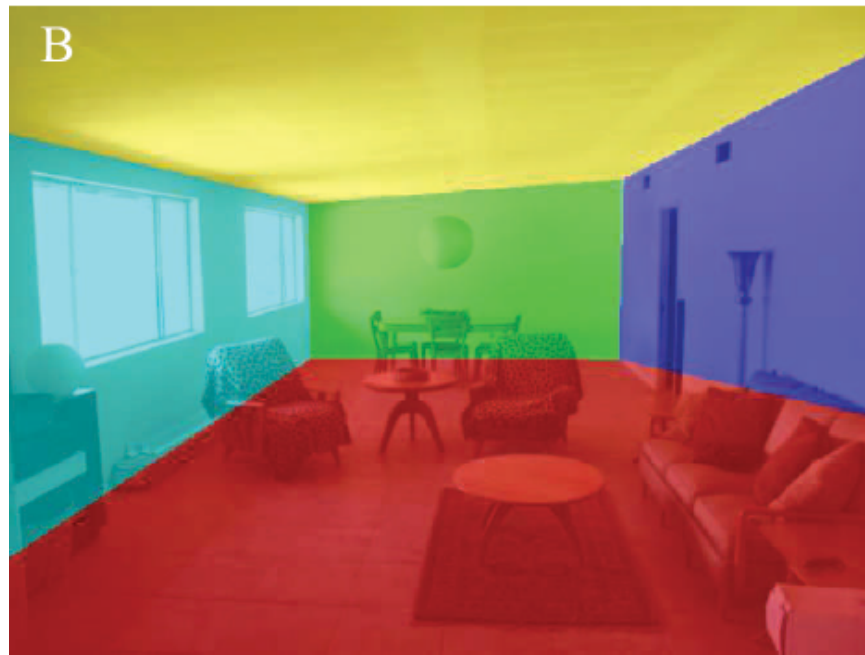


Hedau et al, in review

Estimating the box - I

- Use vanishing points to estimate rotation
 - cf Coughlan Yuille 99, 03; Rother, 02; Yu et al '08; Deutscher et al, '02
- Use lines through vanishing points to get translation
 - box corner
 - using search, cost criterion learned using structure learning
 - yields first estimate of face appearance

The “stage”



Hedau et al, in review

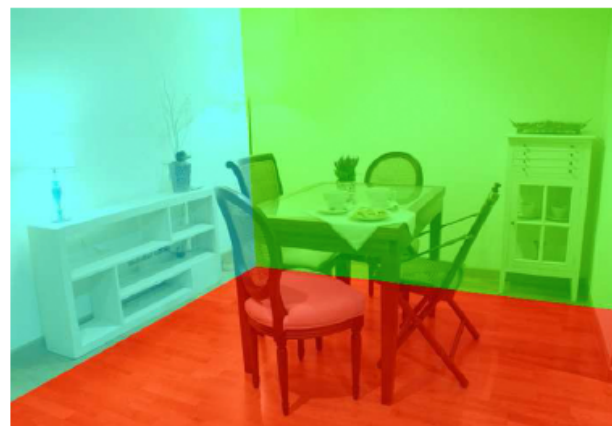
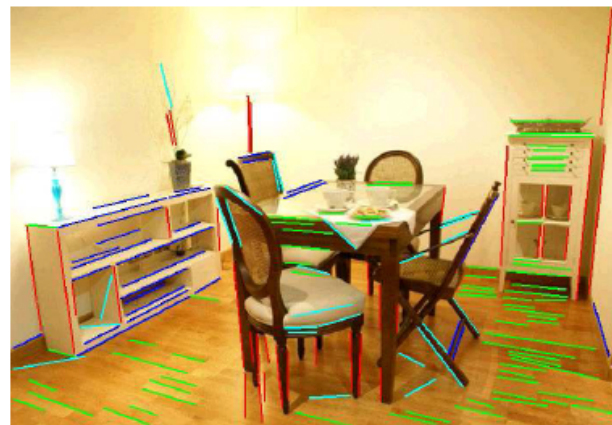
The “players”



Hedau et al, in review

Process - II

- Refine translation estimate by
 - using “non-face” info to rule out some lines, features
 - reestimate with search, criterion learned via structure learning

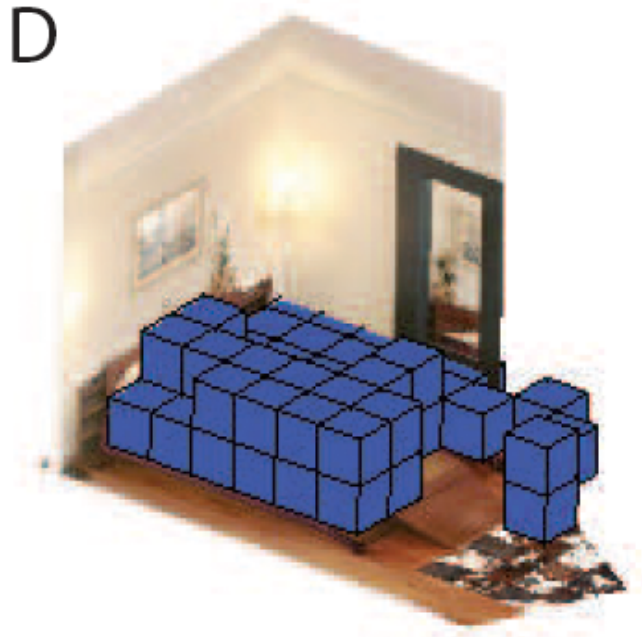


Stage and Players



Now, what can we do...

- Free space estimate
 - using standard SFM construction for camera given manhattan world
 - couple appearance model of objects to approximate geometric models?



Hedau et al, in review

Stage lighting



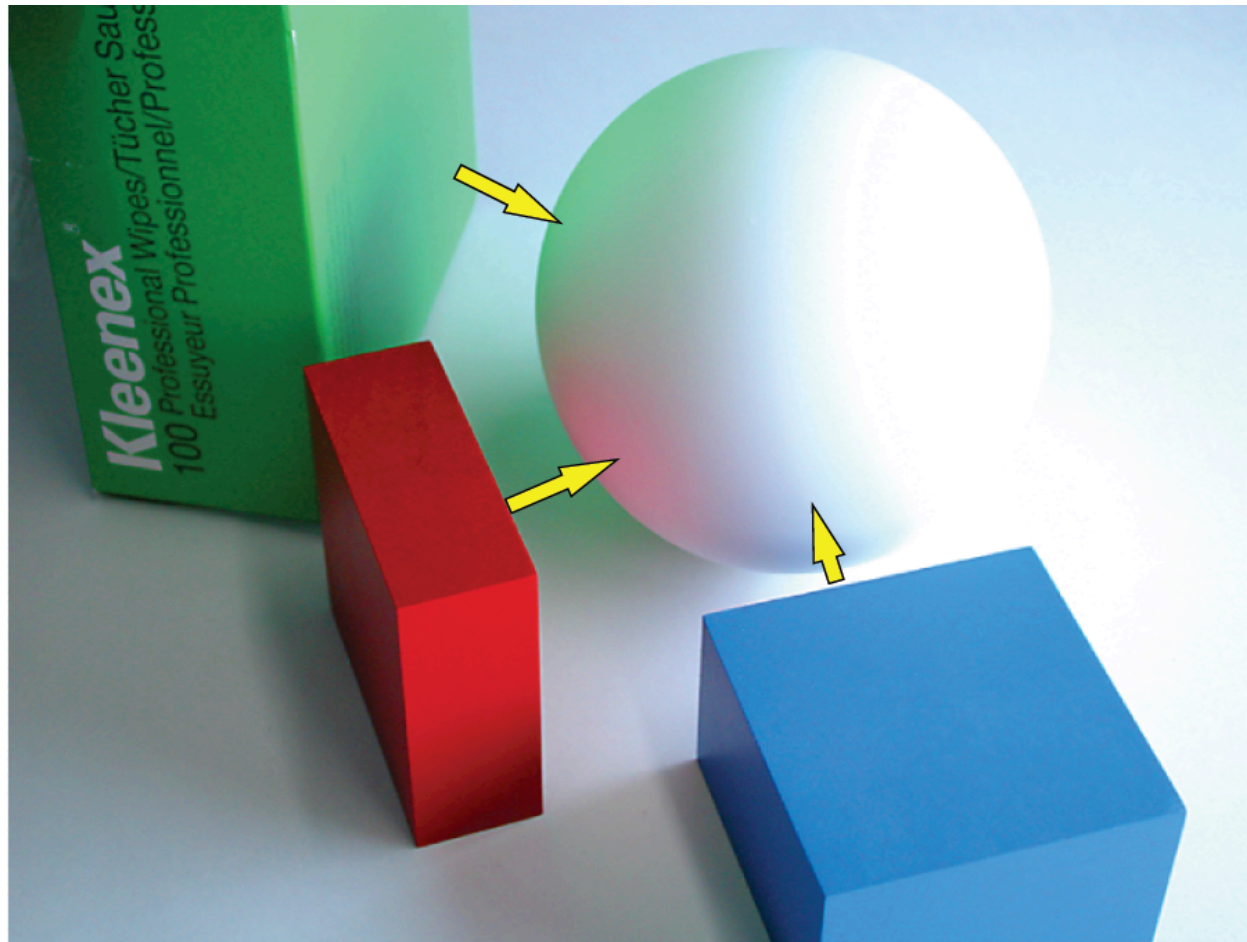
From Koenderink slides on image
texture and the flow of light



Stage lighting could tell us...

- Material properties
 - roughness
 - shininess
 - lightness
 - surface color
- Shape (!?!)

Stage lighting is hard



From Koenderink slides on image texture and the flow of light

Stage lighting is hard

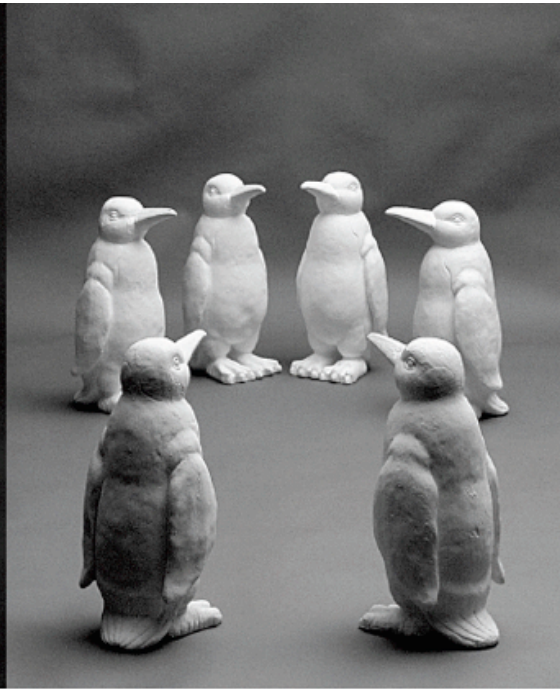


Adolph von Menzel:
Das Balkonzimmer, 1845

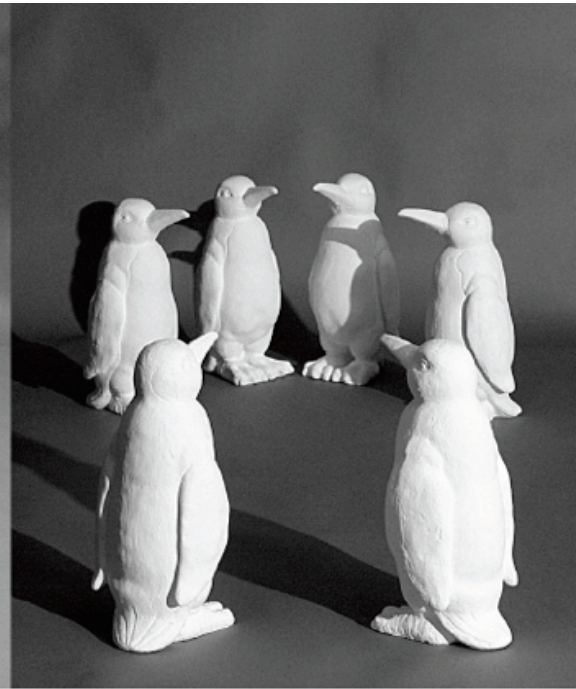
From Koenderink slides on image texture and the flow of light



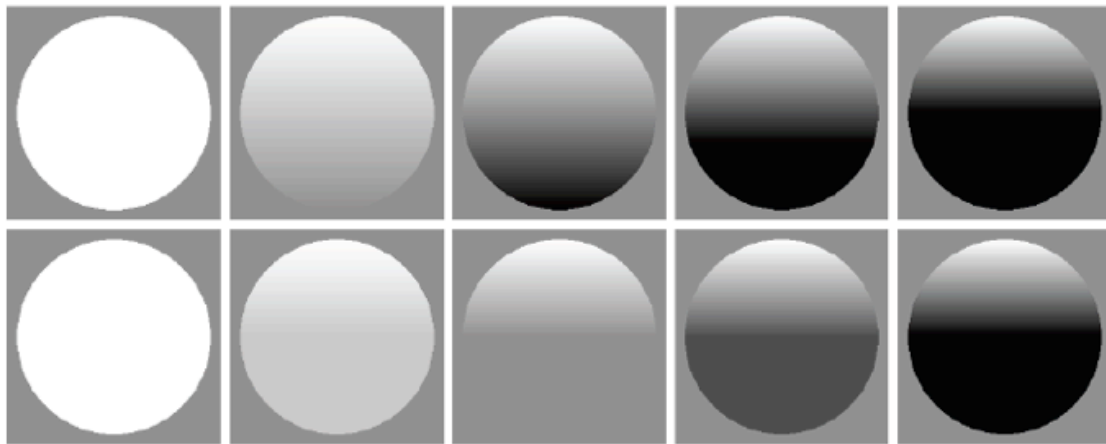
“nativity scene”



“rainy day”



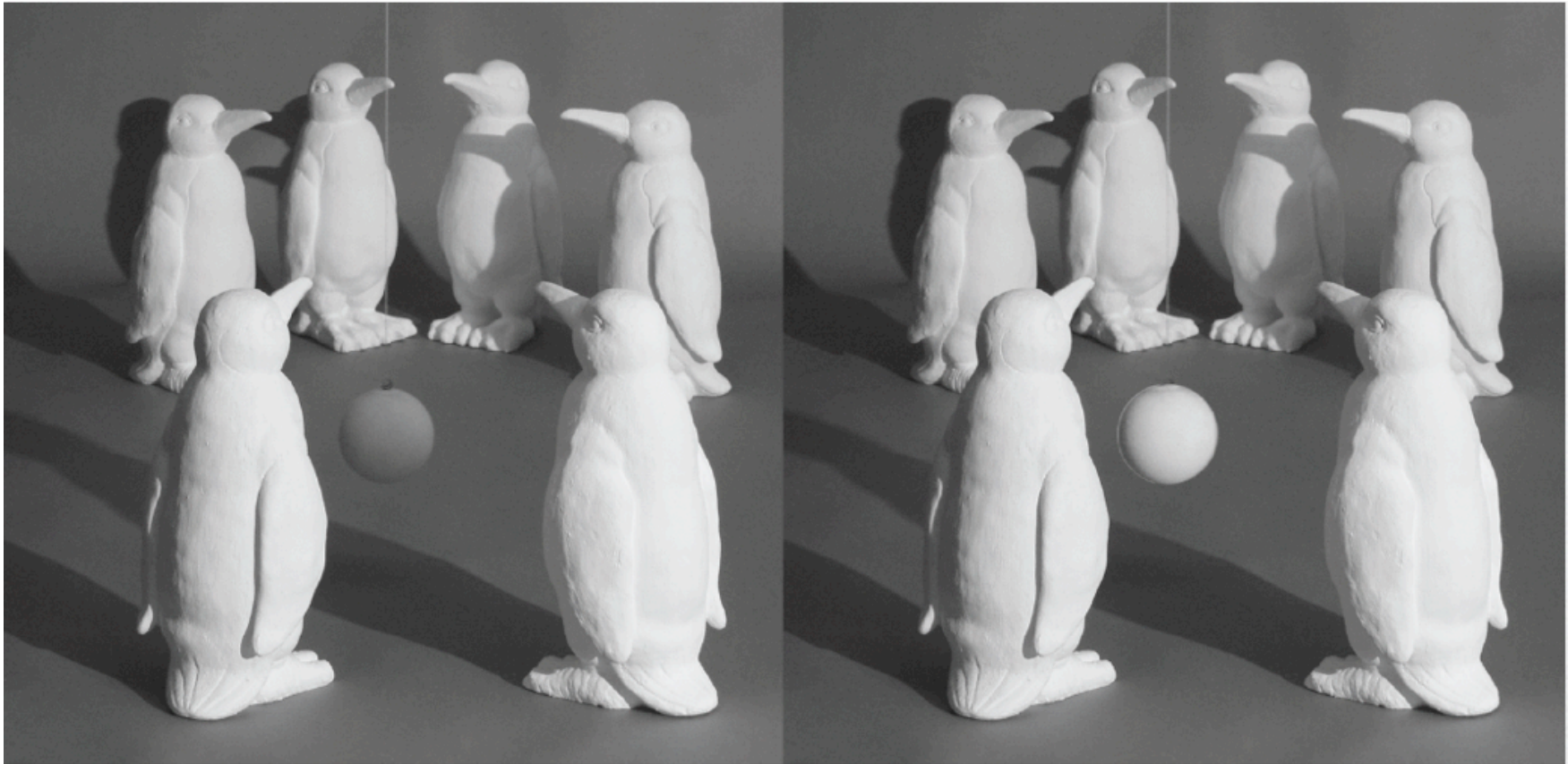
“sunny day on the beach”



scenes ↑

Psychophysics of
“light fields”.

← *probes*



Human observers turn out to be remarkably sensitive to the light field, both to direction and diffuseness.

One exception: all observers “missed” the effect of volume shadow (ground truth – *left*) and produced a non-physical setting – *right*. Cast shadow volumes are ignored.

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Summary points

- Scene categories should be fluid
 - which forces us to think about attributes rather than names
- Some attributes might be
 - what the picture looks like
 - what could be there
 - what the overall geometric stage is
 - how the stage is lit