

# What Makes a Great Picture?

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© Robert Doisneau, 1955

*With many slides from Yan Ke,  
as annotated by Tamara Berg*

15-463: Computational Photography  
Alexei Efros, CMU, Fall 2011

# Photography 101

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- Composition
  - Framing
  - Rule of Thirds
  - Leading Lines
  - Textures and Patterns
- Lighting
  - Direction
  - Color coordination / balance
  - “Golden Hour”

# Framing

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*“Photography is all about framing. We see a subject -- and we put a frame around it. Essentially, that is photography when all is said and done.”*

-- from [photo.blorge.com](http://photo.blorge.com)



# Frame serves several purposes:

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1. It gives the image depth
2. Use correctly, framing can draw the eye of the viewer of an interest to a particular part of the scene.
3. Framing can bring a sense of organization or containment to an image.
4. Framing can add context to a shot.

<http://digital-photography-school.com/blog/frame-your-images/>

# Examples of nice framing

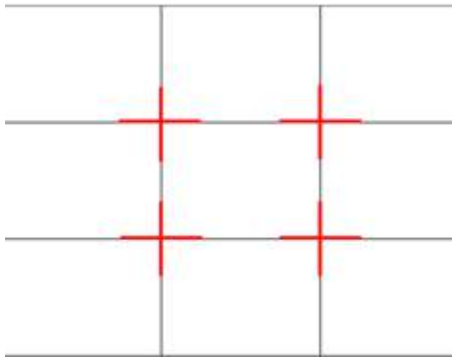
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<http://flickr.com/photos/paulosacramento/226545698/>  
<http://flickr.com/photos/chrisbeach/13868545/>  
<http://flickr.com/photos/74531485@N00/929270814/>  
<http://flickr.com/photos/freakdog/223117229/>  
<http://flickr.com/photos/cdm/253805482/>

# Rules of Thirds

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# Other examples

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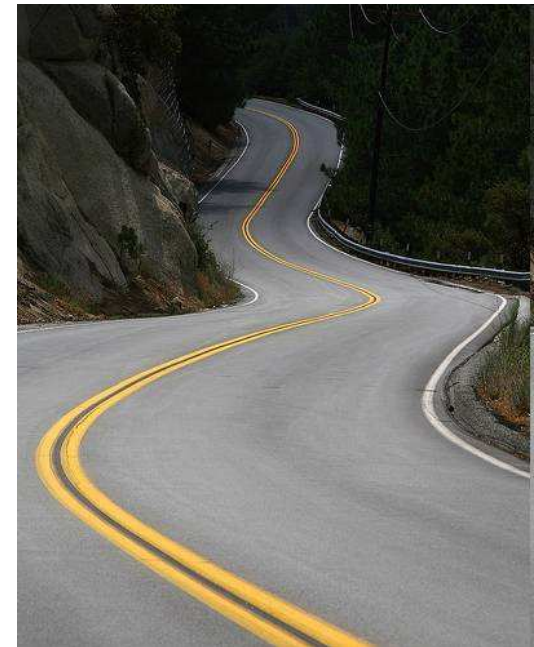
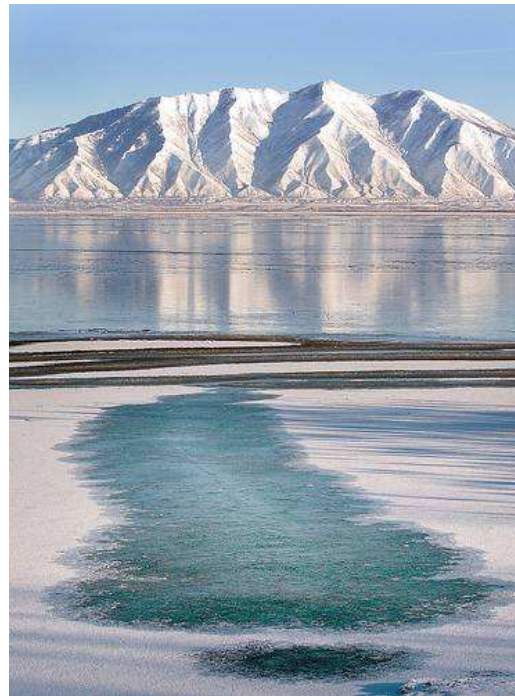
# Leading Lines





# More examples

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# Textures and Patterns



# Color Coordination



Complementary colors (of opposite hue on color wheel)

# Front Lighting



# Side Lighting



# Back Lighting



# Photography 101

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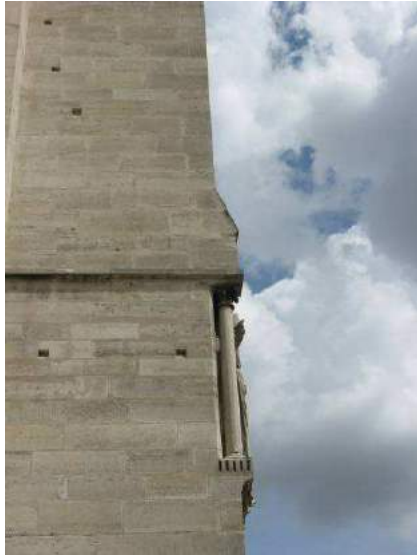
Anyone can take great pictures...





# I am a sucky photographer...

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...but I am a pretty good photo critic!

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<http://flickr.com/photos/aaefros/>

# of my Paris photos on Flickr: 32

Total # of my Paris photos: ~1250

~2%

# The Postmodern Photographer

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## The Old Days: a pre-process

- Load film
- Find subject
- Position camera
- Set all the settings “just right”
- Take a deep breath...
- ...Press button!

## The New Digital Days: a post-process

- Get a 2 GB memory cartridge
- Take pictures like there is no tomorrow!!!
- ...
- Back home, spend hours of agony trying to find 1-2 good ones

# How to recognize the good photos?



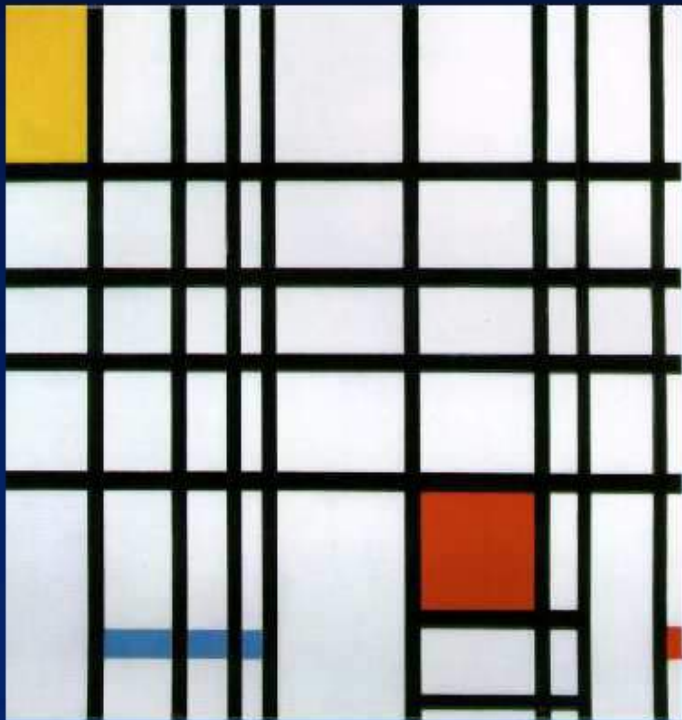
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# Outline

- Photography 101
- Recognition (CVPR '06)
  - What makes one photo better than another?
  - What features can we extract?
  - How can we measure our performance?

[Y. Ke, X. Tang, and F. Jing. \*The Design of High-Level Features for Photo Quality Assessment\*. CVPR 2006.](#)

# Not Critiquing Art



Piet Modrian



Lothar Wolleh

Not considering semantic measures of what makes a photo good (subject matter, humor, etc).  
Professional = those you would frame, snapshot = those that would stay in photo album.

# Applications

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Image search for improved quality along with relevance.

Automatically select the best photos from a set of vacation pictures to choose the best ones to show.

See if computer can perform well on a traditionally human task.

# What makes one photo better than another?

- Simplicity
- Realism
- Basic photographic techniques



# Simplicity

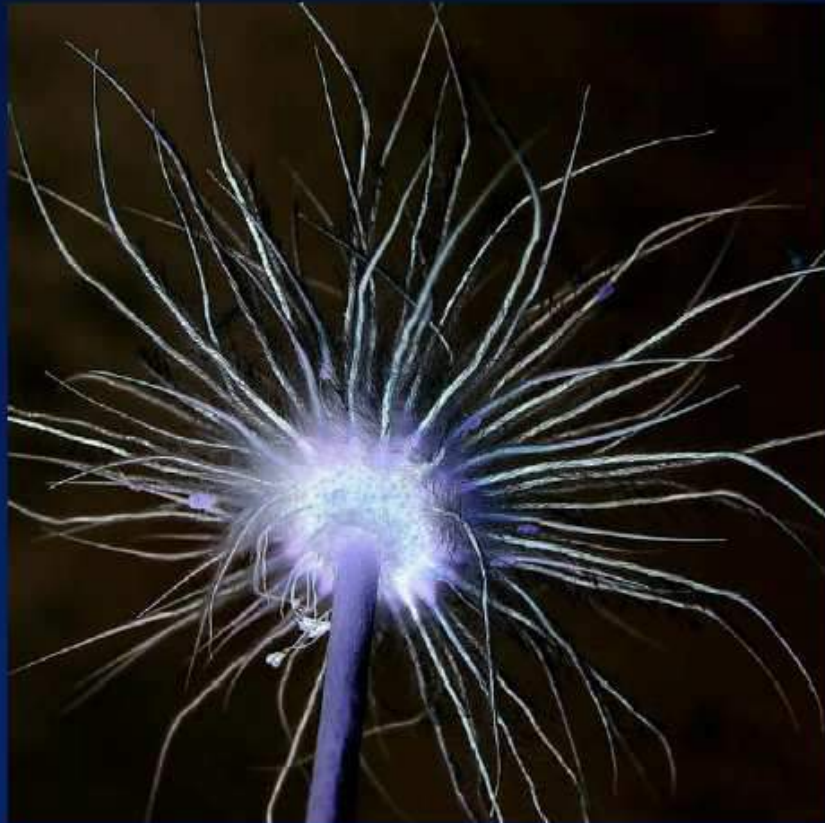


“Look Into” by Josh Brown @ Flickr



Prof - Obvious what one should be looking at  
ie easy to separate subject from the  
background. Snap – unstructured, busy, filled  
with clutter.

# Simplicity



“alien flower” by Josef F. Stuefer @ Flickr

# Simplicity



“Waiting in line!” by Imapix @ Flickr

# Basic techniques

- **Blur** - Snaps – entire photo blurry indicates poor technique. Prof - background out of focus by widening the lens aperture, but foreground in sharp focus.
- **Contrast and brightness** Make the subject pop out by choosing complementary colors for subject & background. Isolate the subject by increasing lighting contrast between subject & background.

Abstract concepts - “Good composition, color & lighting”

# (Sur) Realism

Snaps look real, while  
prof photos look surreal.

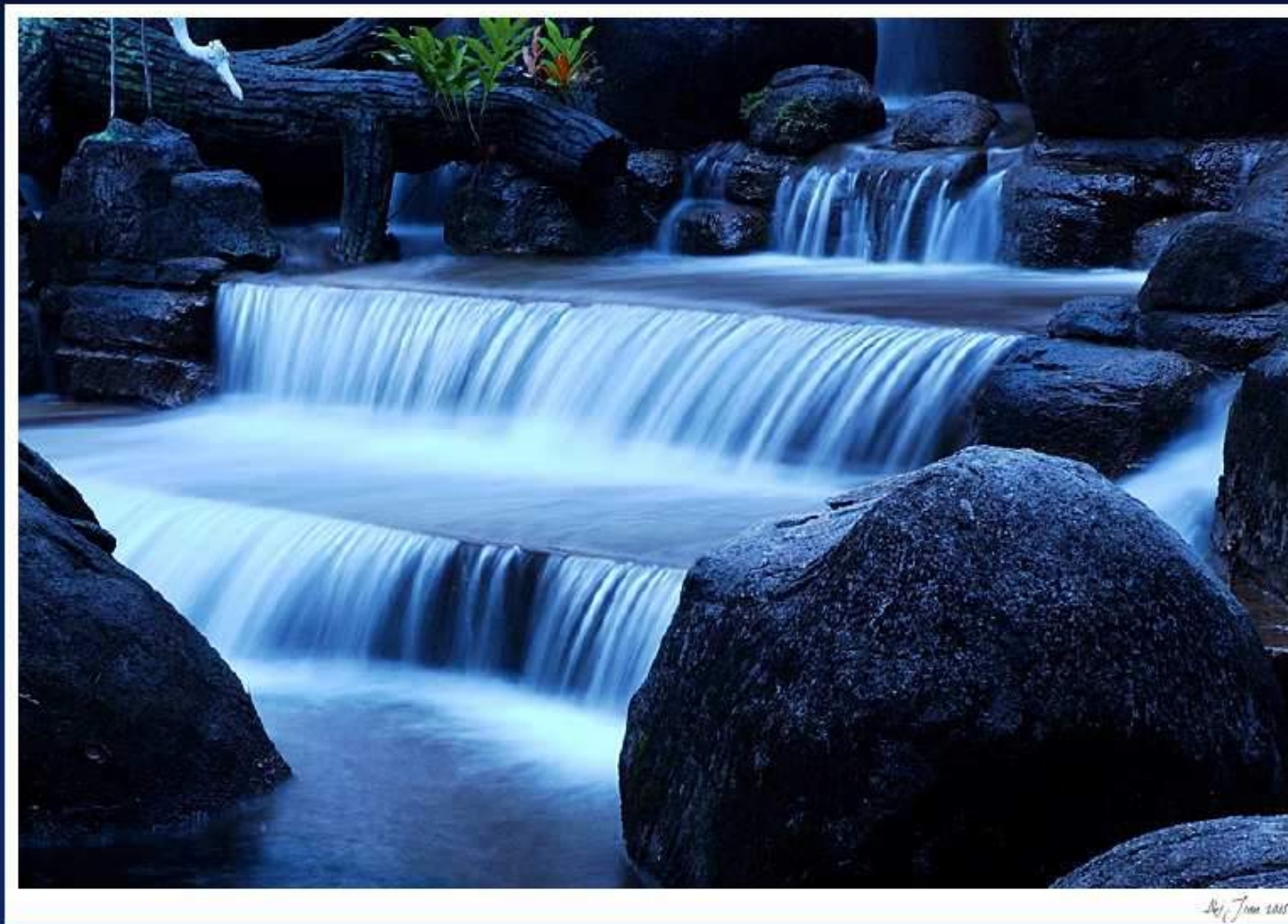


“Golden Gate Bridge at Sunset” by Buzz Andersen @ Flickr



“Golden Gate 3” by Justin Burns @ Flickr

# (Sur) Realism



“Somewhere Only We Know Prt2 (sic)” by Aki Jinn @ Flickr

# Techniques

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Lighting conditions – time of day (morning, dusk), colored filters to adjust color balance (make sky bluer, sunset more brilliant), careful color selection of scene

Camera settings – adjust settings like focal length, aperture, shutter speeds to modify mood, perspective. Eg might use long shutter speed to capture waterfall and give a misty look

Subject matter – ordinary objects in unusual poses or settings (challenging since would need obj rec first)

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# Features – Spatial Distribution of Edges



More edges  
near border  
due to  
background  
clutter



More edges  
near center  
of img

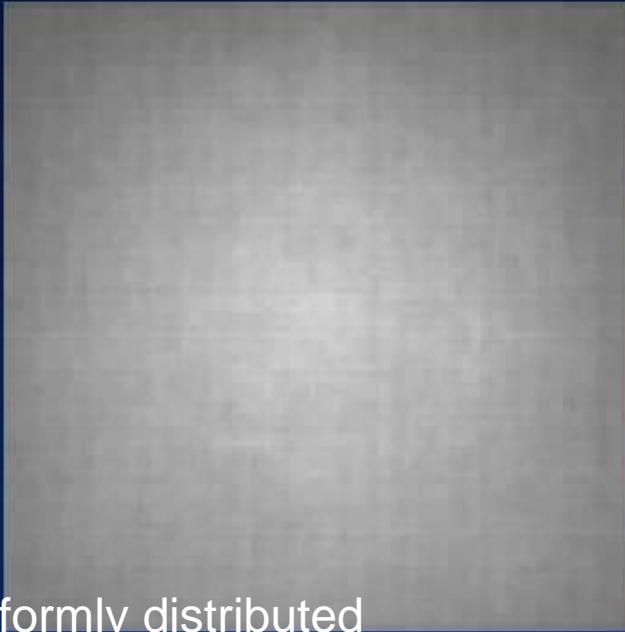


“Picture of a picture...” by Ted Johnson @ Flickr

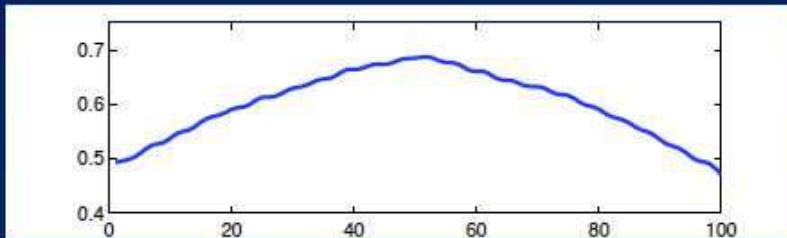
Trying to capture a photo’s “simplicity”

# Spatial Distribution of Edges

Mean Laplacian of snapshots

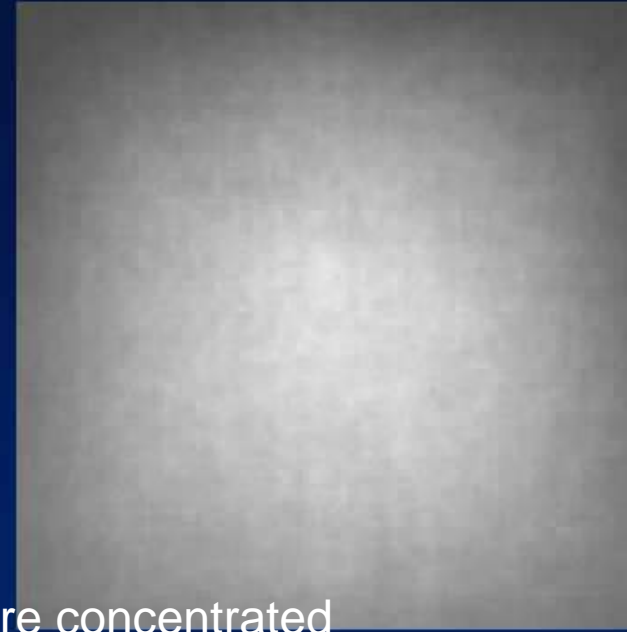


More uniformly distributed

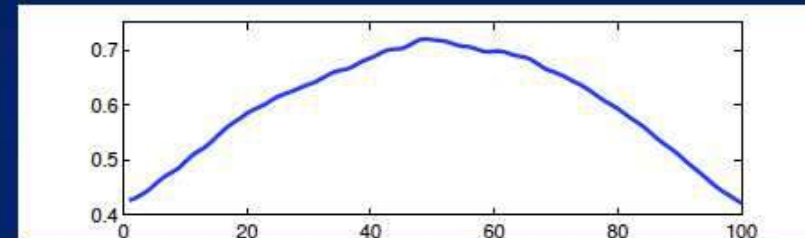


Low quality photos

Mean Laplacian of professional



More concentrated



High quality photos

Expect high quality photos to have high spatial frequency edges nearer to center than snapshots

# Edge width

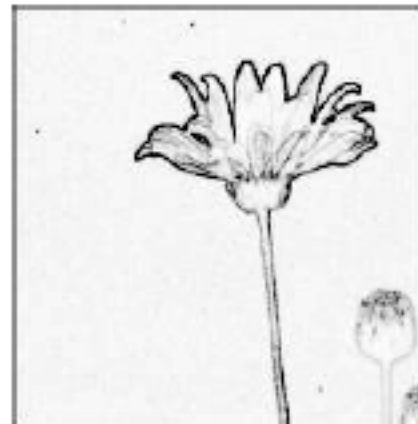
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Calculate area that edges occupy – width of bounding box covering 96% of edge energy

Cluttered regions should tend to produce a larger bounding box, and well defined subjects should produce a smaller one.



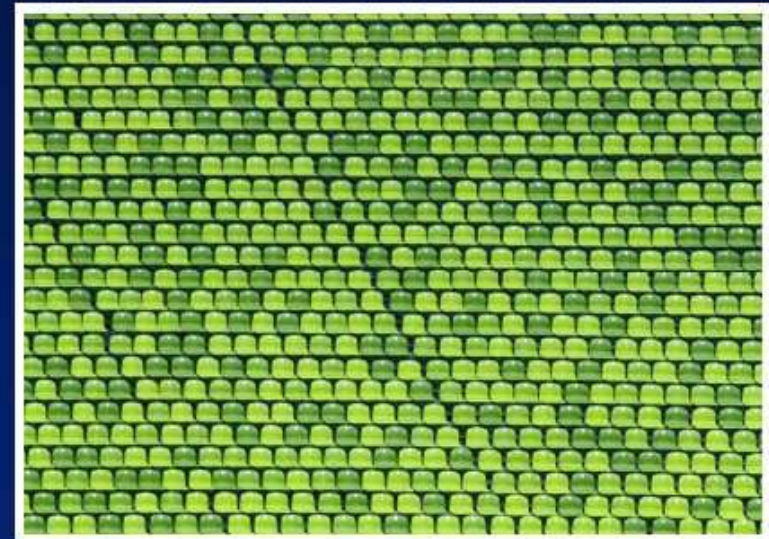
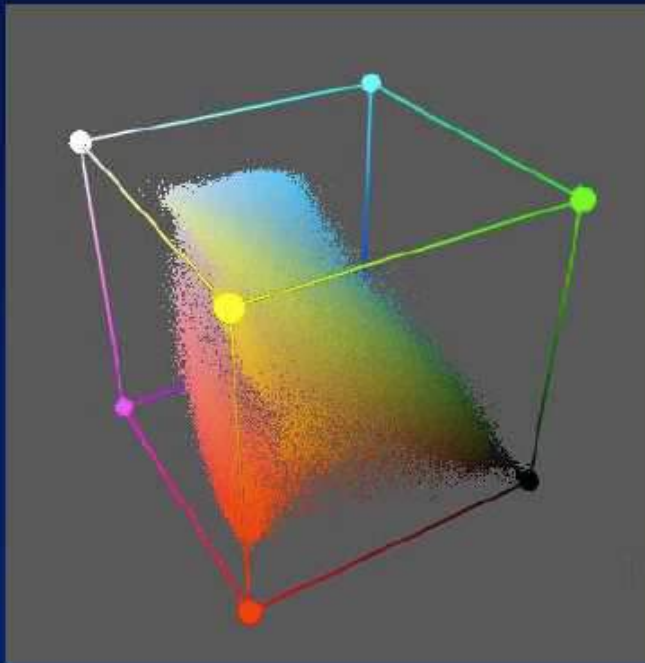
.94



.56

# Color Distribution

- K-NN on color histogram

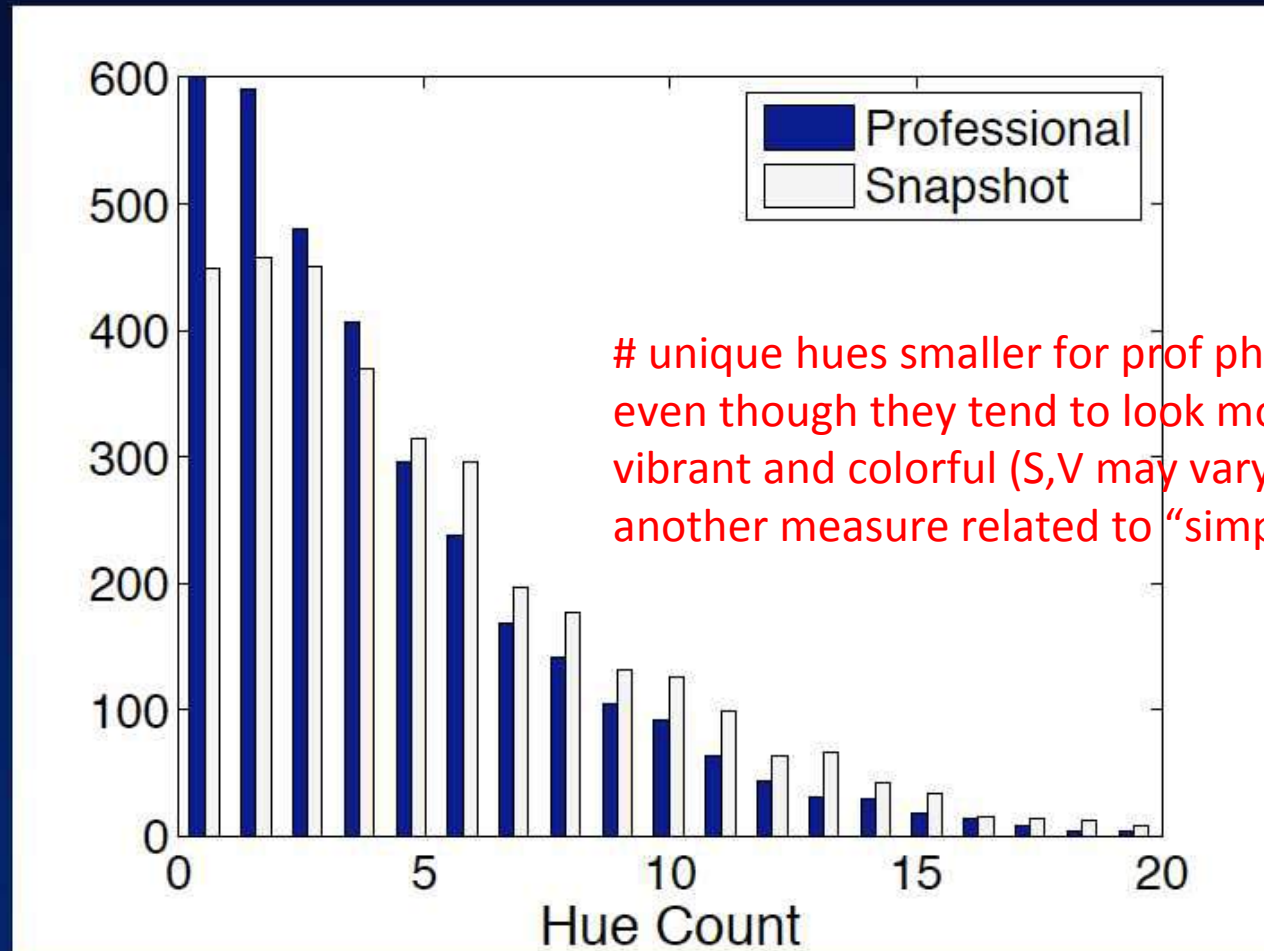


$$q_{cd} = \# \textit{professional\_neighbors}$$

For query image find k nearest neighbors in training set.  
Quality = number of prof neighbors in top 5.

20 bin histogram defining possible unique hues

# Hue Count



# unique hues smaller for prof photos even though they tend to look more vibrant and colorful (S,V may vary more) – another measure related to “simplicity”



$$q_h = 20 - (\# \text{ hues } > \text{ threshold})$$

# Most unlikely colors...

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From Lalonde and Efros, ICCV'2007

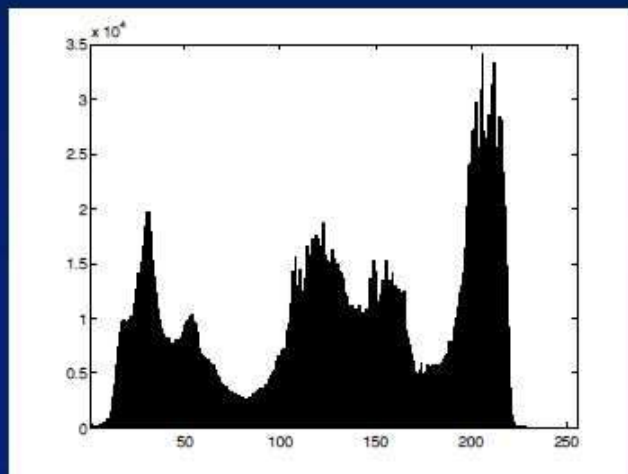
# Blur

- Look at frequency distribution.
- Measure the amount of blur in the sharpest object, instead of the *average* blur.

Prof photos  
should  
have some  
part of  
photo in  
sharp focus

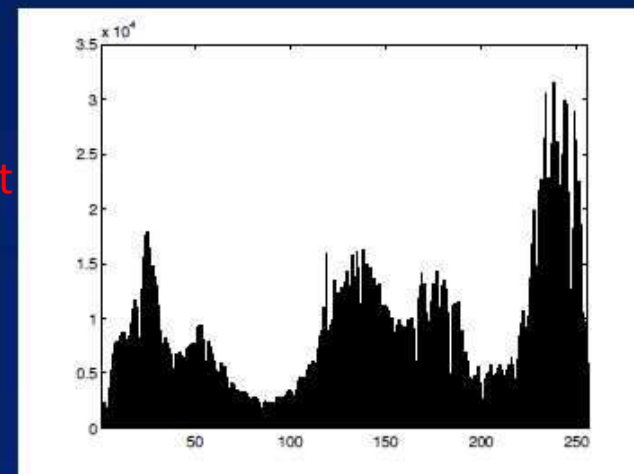


# Low Level Features - Contrast



Prof photos  
usually have  
higher contrast

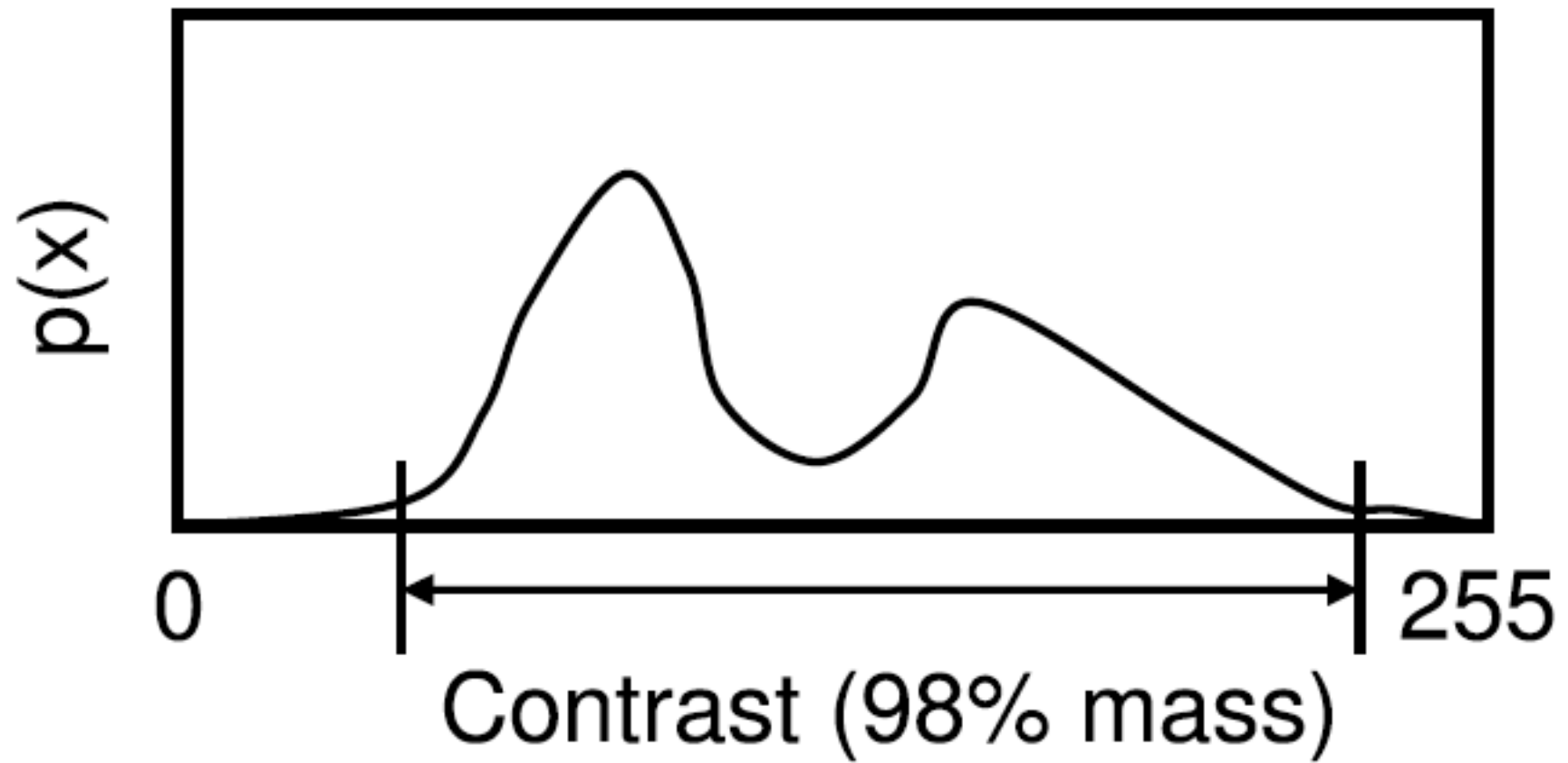
Contrast =  
width of  
middle 98%  
mass of hist





# Contrast

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## Low Level Features – Avg. Brightness



Professional photographers may adjust exposure to be correct on subject only so subj pops from bkd. Cameras tend to adjust brightness to average at 50% gray, but prof photos might deviate significantly. Use ave brightness as feature.

# Classifier

- Naives Bayes
- We assume independence of the features
- We achieve better results with added features even though they are not independent.

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# Dataset – DPChallenge.com

Use photos average rating as ground truth quality measure

Use only top 10%, bottom 10% as dataset.

Use half for training/half for testing.



Statistics	Voting Breakdown
<b>Place:</b> 1 out of 829	1   0
<b>Avg (all users):</b> 7.987	2   1
<b>Avg (commenters):</b> 8.805	3   2
<b>Avg (camera):</b> 7.998	4   5
<b>Avg (no camera):</b> 6.333	5   24
<b>Views since voting:</b> 6597	6   54
<b>Views during voting:</b> 1003	7   89
<b>Votes:</b> 478	8   109
<b>Comments:</b> 190	9   89
<b>Favorites:</b> 133 (view)	10   105


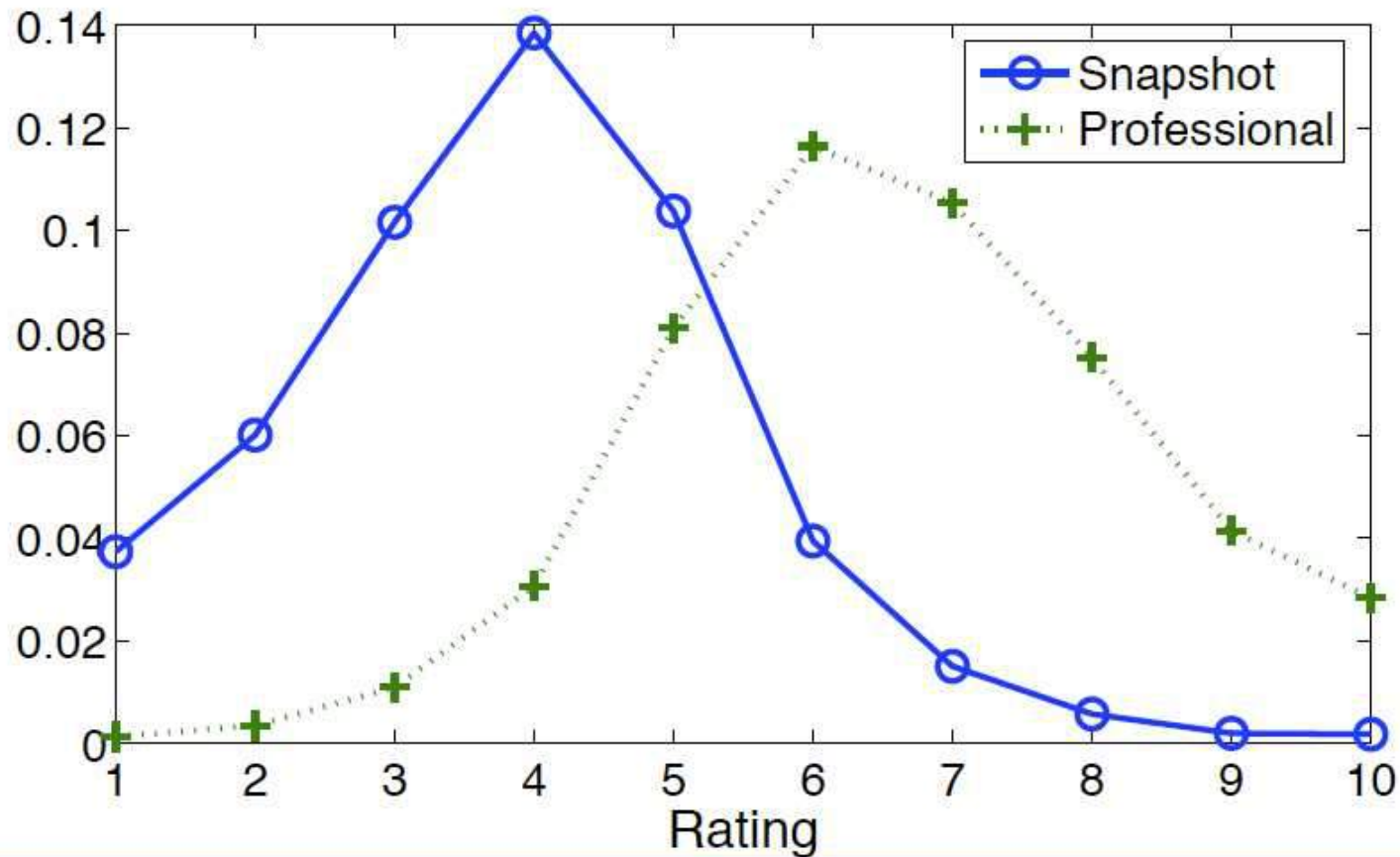
 [Add this photograph to your favorites!](#)

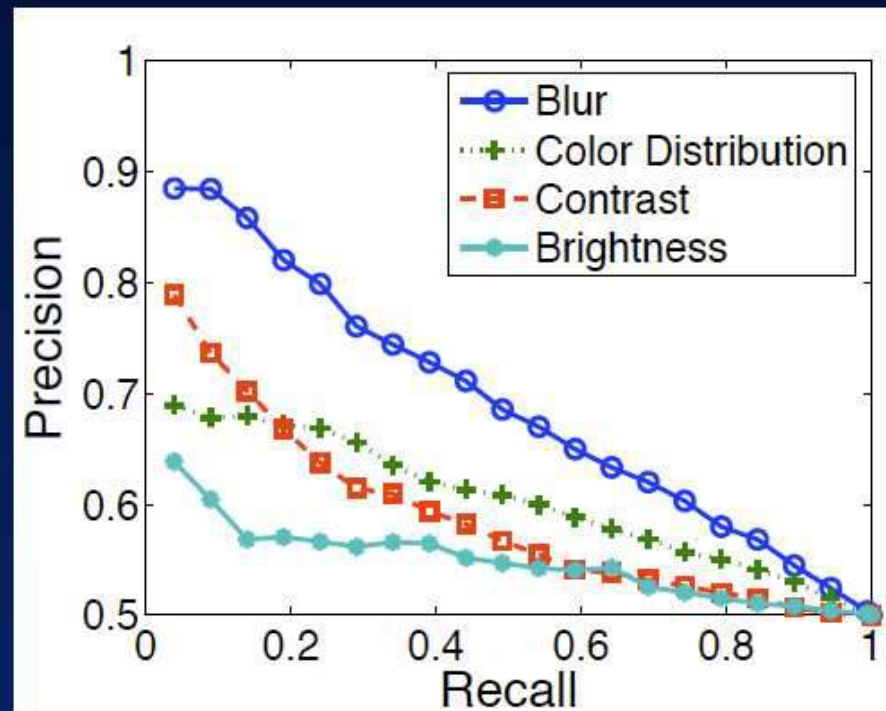
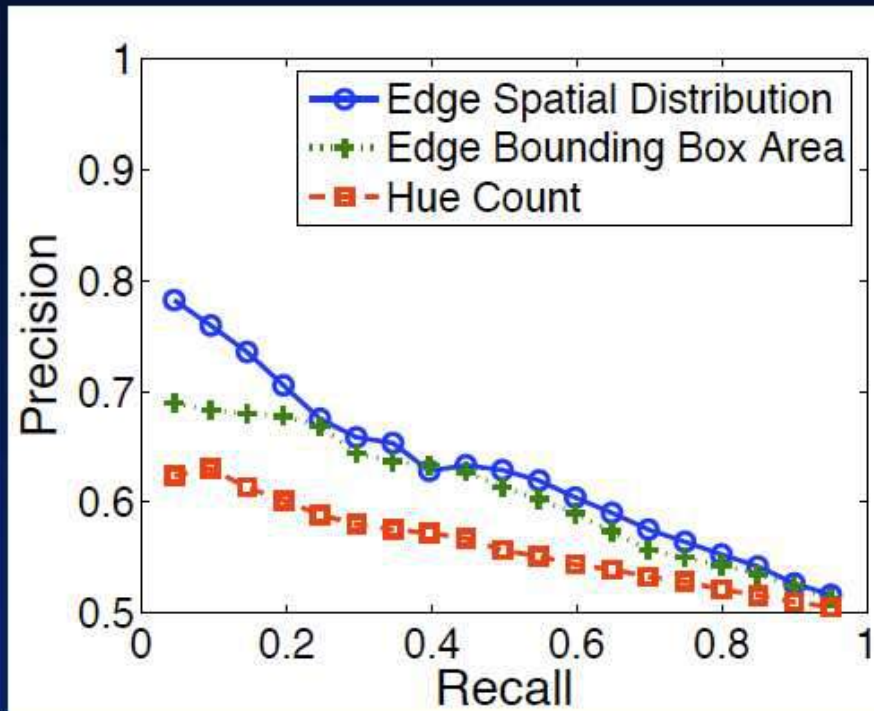
Photo contest website,  
user rated

60K photos  
40K photographers  
10/90 percentile

# Difficulty of Dataset



# Results



$$\text{recall} = \frac{\# \text{ professional photos above threshold}}{\text{total } \# \text{ professional photos}}$$

$$\text{precision} = \frac{\# \text{ professional photos above threshold}}{\# \text{ photos above threshold}}$$

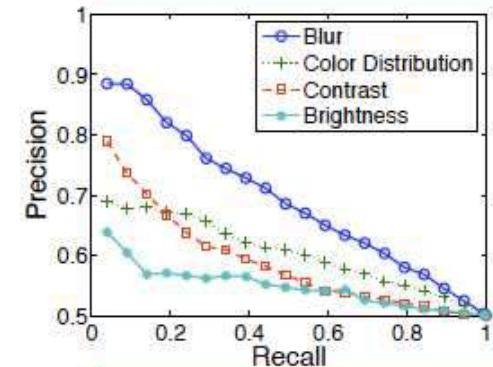
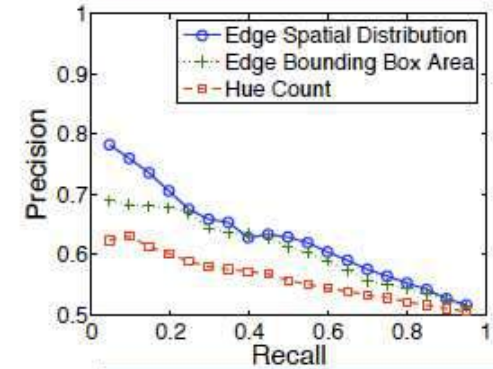
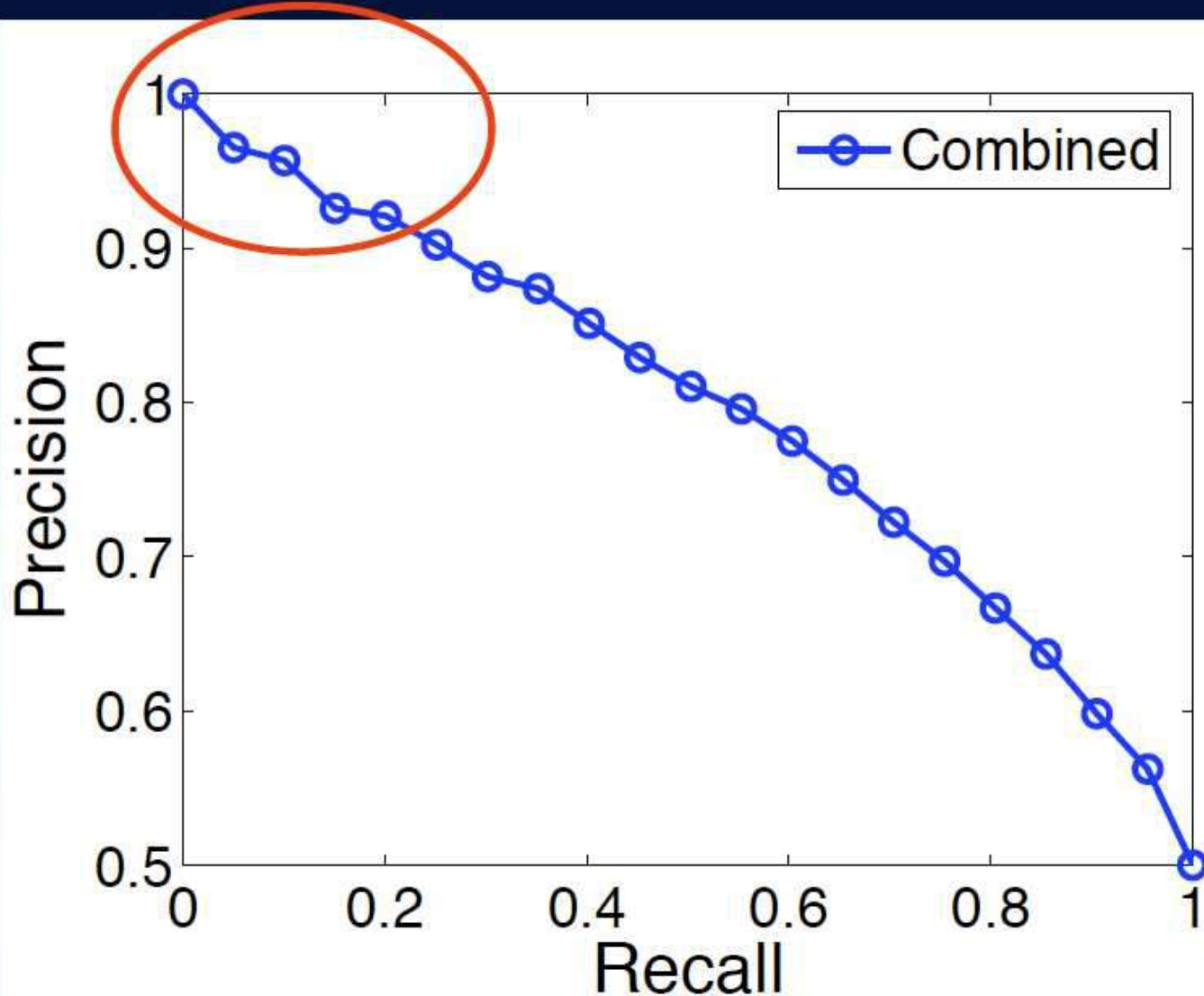
# Most Distinctive Feature: Blur

- A *badness* metric, rather than a *goodness* metric.





# Results



72% classification rate

# Web Retrieval Results



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# Web Retrieval Results



...




# Web Retrieval Results












# Oscar (Lei Yau et al, IJCV'11)

**OSCAR**



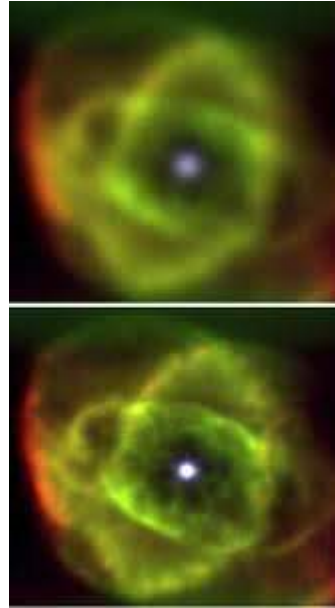
Color Confidence: 55.5%

Rating: **73** / 100

SIMPLicity	S+Composition	S+C+Aesthetics
		
		
		

# Lucky Imaging

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Juliet Bernstein, **Aseem Agarwala**, Brian Curless. "Candid Portrait Selection From Video," *ACM Transactions on Graphics (Proceedings of SIGGRAPH Asia 2011)*

[http://grail.cs.washington.edu/projects/candid\\_video\\_portraits/](http://grail.cs.washington.edu/projects/candid_video_portraits/)