



SERF:

# Enabling Optimization of Socially Sourced Images using Psychovisual Enhancements

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# Overview

- Introduction
- Social Media
- Lossy Compression
- Psychovisual Enhancements
- Experiment
- Results

# What is Psychophysics?

Psychophysics seeks to quantitatively investigate the relationship between psychological events and physical events, or more specifically, between sensations and the stimuli that produce them.

—Britannica



TRADITIONAL WEB CLIENTS



Mobile Devices



Search Engine  
Crawlers, Scrapers,  
and Bots

NON-TRADITIONAL WEB CLIENTS

Client Side



TRADITIONAL WEB CLIENTS

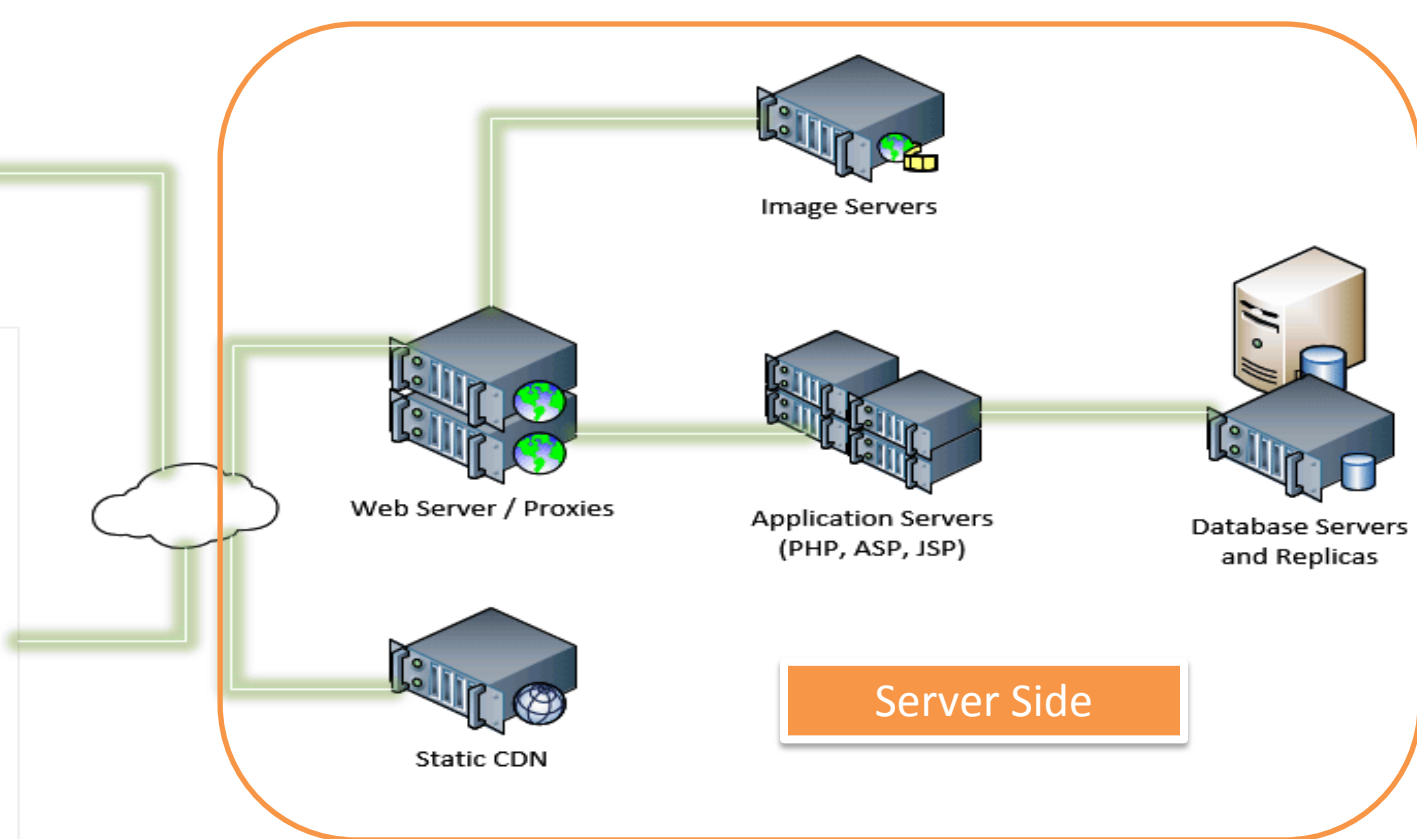


Mobile Devices



Search Engine  
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and Bots

NON-TRADITIONAL WEB CLIENTS



# Web Application Ecosystem

## Latency

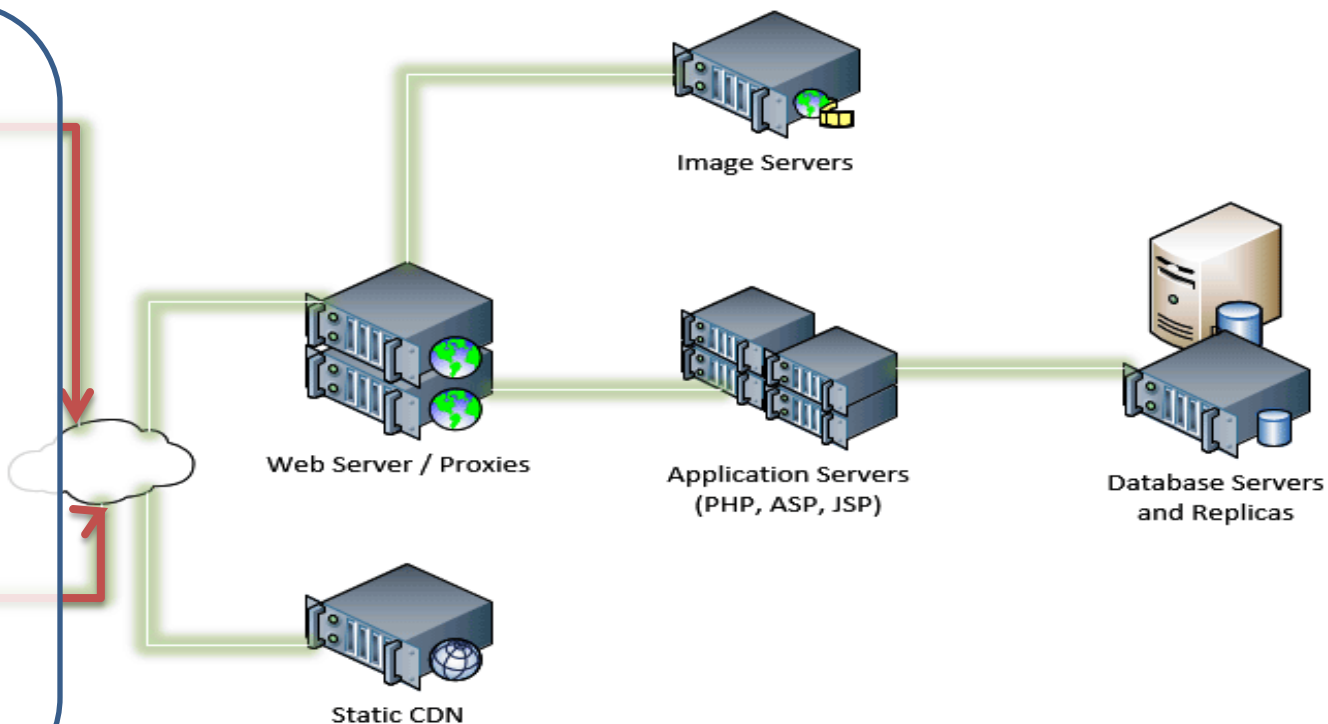
- DNS Lookups
- Multiple servers
- HTTP protocol
- TCP protocol
- Rendering
- Search Engine

## Bandwidth

- Limited
- Improving
- Video / Images
- More Devices
- Better Devices

Search Engine  
Crawlers, Scrapers,  
and Bots

NON-TRADITIONAL WEB CLIENTS



Performance and Scalability Problems in the  
**Web Application Ecosystem**

Latency

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Search Engine  
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NON-TRADITIONAL WEB CLIENTS

Reliability

Bandwidth

Spoon-feeding

Updates

Storage

Image Servers

CPU

Application Servers  
(PHP, ASP, JSP)

I/O

Database Servers  
and Replicas

Web Server / Proxies

Static CDN

Performance and Scalability Problems in the  
**Web Application Ecosystem**

Latency

- DNS Lookups
- Multiple servers
- HTTP protocol
- TCP protocol
- Rendering
- Search Engine Crawlers, Scrapers, and Bots

Bandwidth

- Limited
- Improving
- Video / Images
- More Devices
- Better Devices

Reliability

Bandwidth

Spoon-feeding

Load Balancer

Updates

Tail optimization

Storage

OS enhancements

Image Servers

CPU

Object cache

Application Servers  
(PHP, ASP, JSP)

Session cache

Opcode cache

I/O

Database Servers  
and Replicas

Query Cache

Replication

Partitioning

NON-TRADITIONAL WEB CLIENTS

Performance and Scalability Problems in the  
Web Application Ecosystem





## Resource Contention

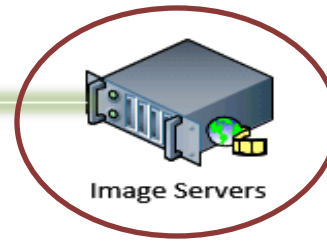
Many opportunities to study systems of systems and their complex interactions, often resulting in Pareto-optimal (multi-objective tradeoff) scenarios.

bing  
Google

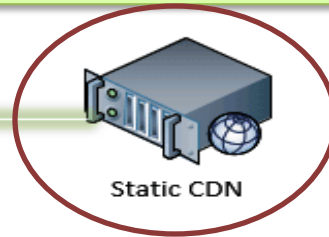


NON-TRADITIONAL WEB CLIENTS

Performance and Scalability Problems in the  
**Web Application Ecosystem**



Have we done everything we can for  
static image servers?



NON-TRADITIONAL WEB CLIENTS

Web Application Ecosystem

# Social Media

## ► Blogs and social media

- Social networks and online communities
- Exploding with multimedia content
- 73% of the world population now use

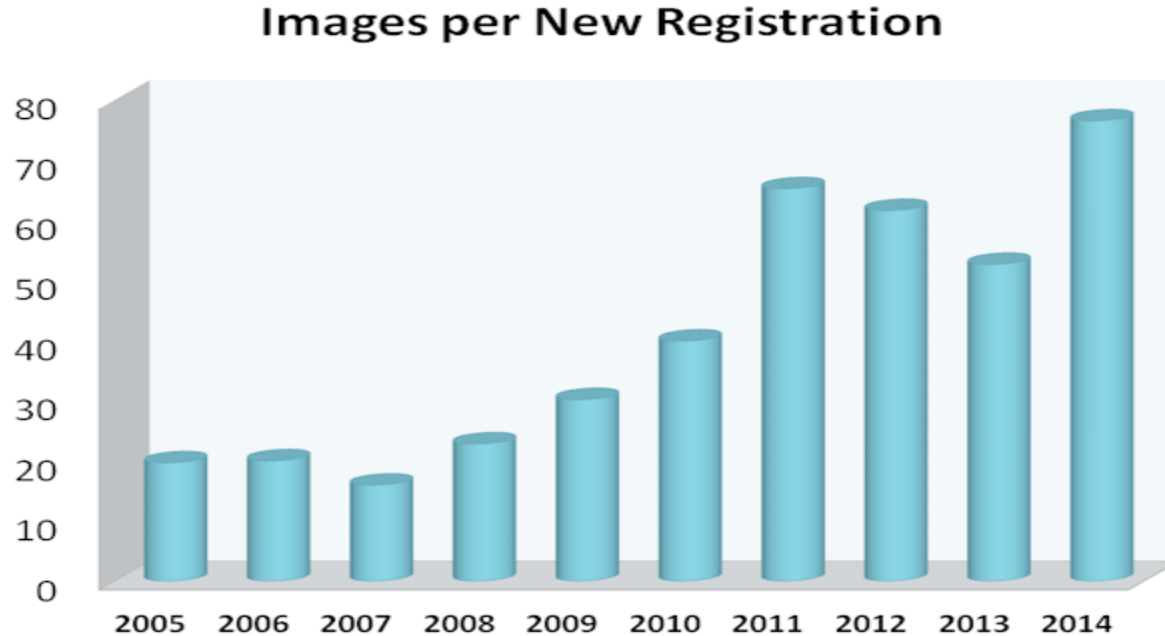


## ► Facebook

- 2<sup>nd</sup> busiest site globally
- 1.28 billion monthly users
- 609 million mobile – daily

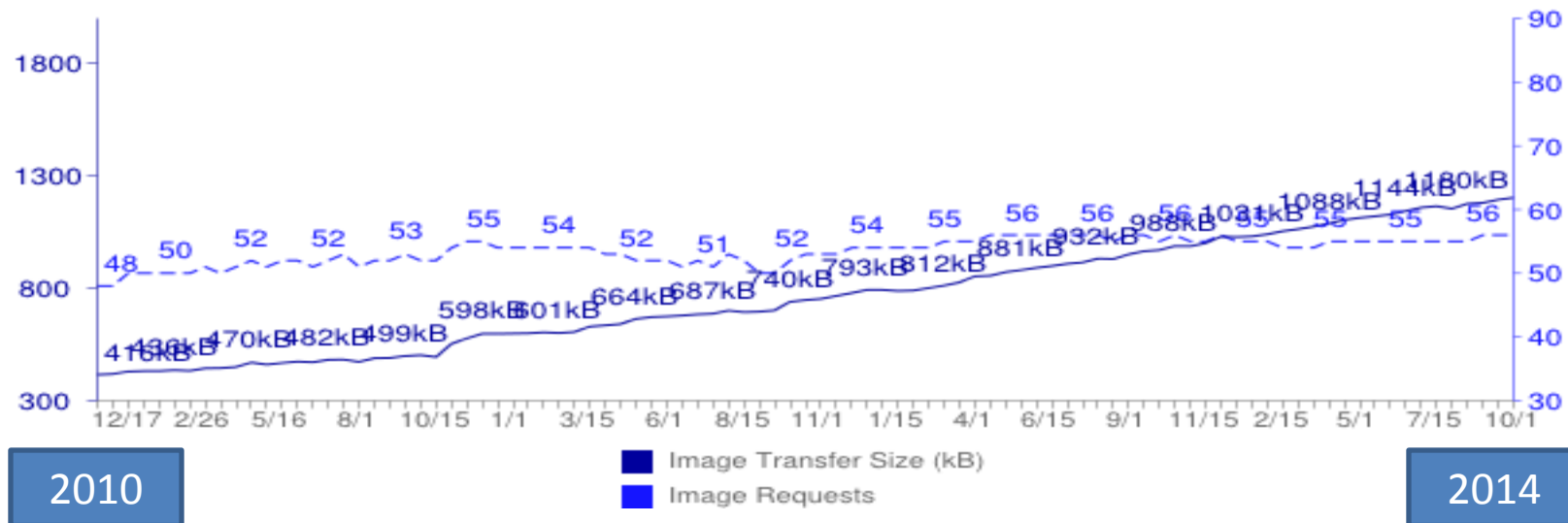


# Social Media Images



*Images per new registration on the author's online community.*

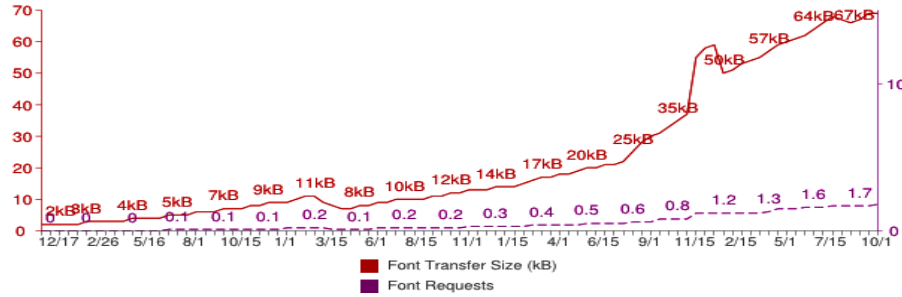
## Image Transfer Size & Image Requests



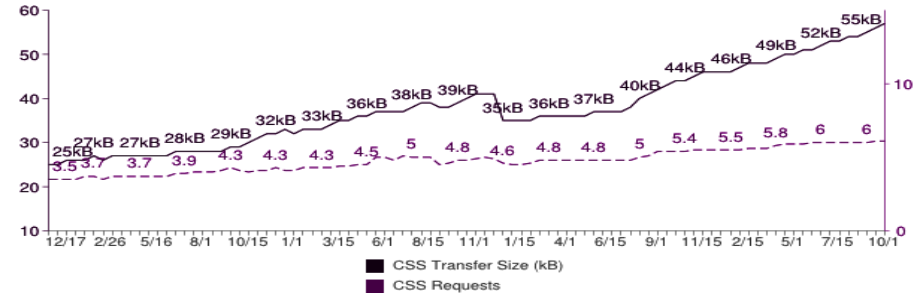
Source: <http://archive.org> crawl statistics

Image sizes have increased steadily

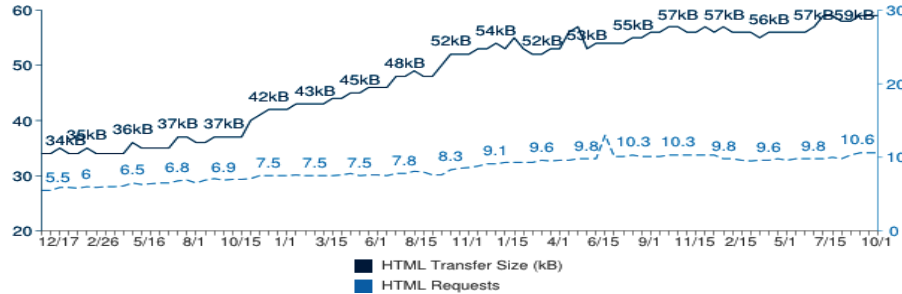
### Font Transfer Size & Font Requests



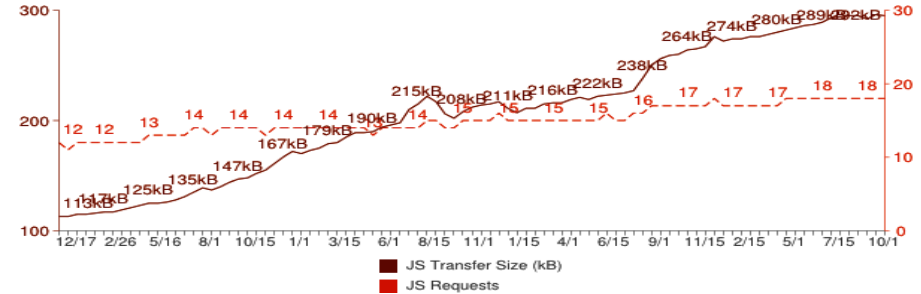
### CSS Transfer Size & CSS Requests



### HTML Transfer Size & HTML Requests



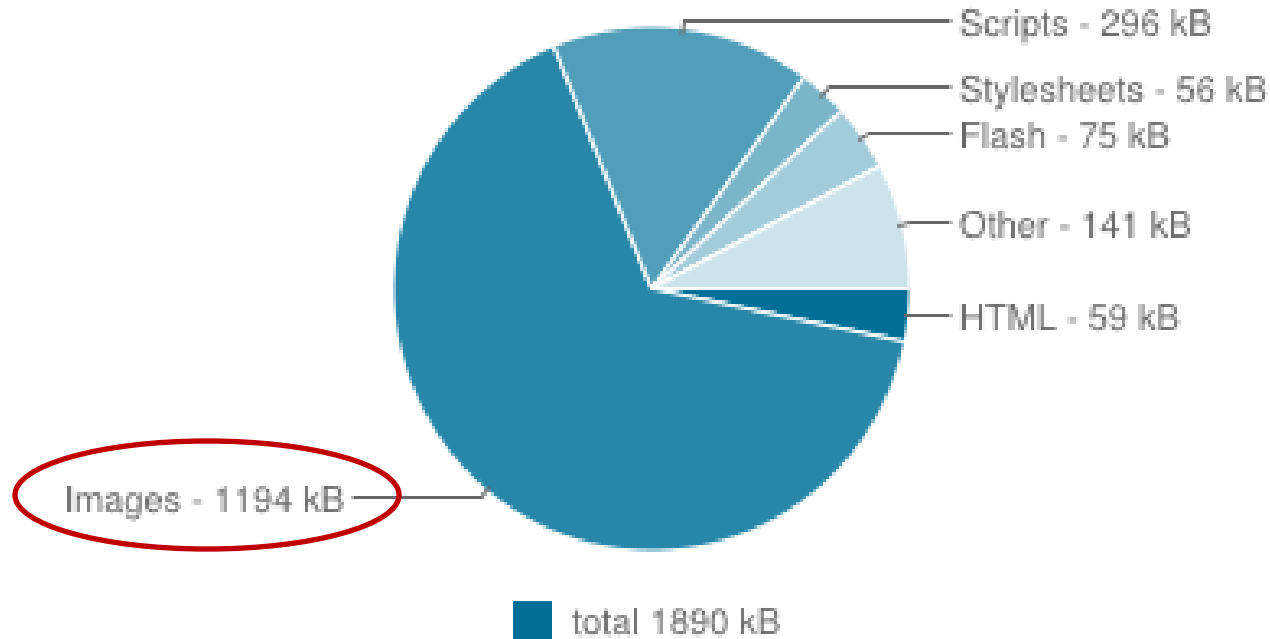
### JS Transfer Size & JS Requests



Source: <http://archive.org> crawl statistics

# In fact, web pages are just getting larger overall

## Average Bytes per Page by Content Type

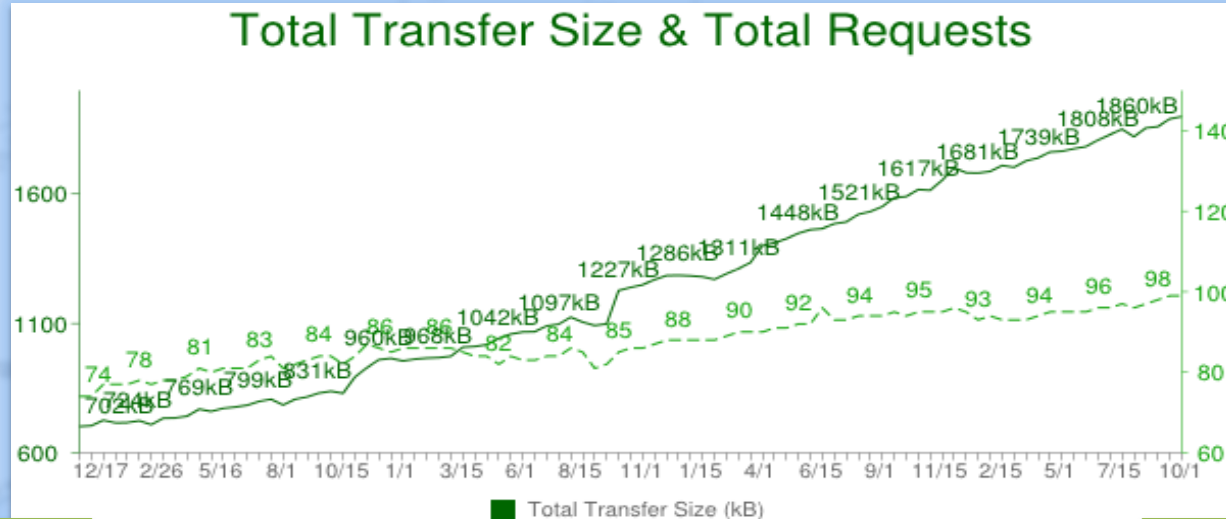


Source: <http://archive.org> crawl statistics

# Images account for 63% of byte transfer

Font Transfer Size & Font Requests

CSS Transfer Size & CSS Requests



2010

2014

Source: <http://archive.org> crawl statistics

In fact, web pages are just getting larger overall



Font Transfer Size & Font Requests

CSS Transfer Size & CSS Requests

Total Transfer Size & Total Requests

Even so, this does not fully account for social media growth, since dynamic sites are not crawled by httparchive, which includes only landing pages in the Alexa top 1M.

Source: [httparchive.org](http://httparchive.org) crawl statistics

In fact, web pages are just getting larger overall

Img Requests per Page

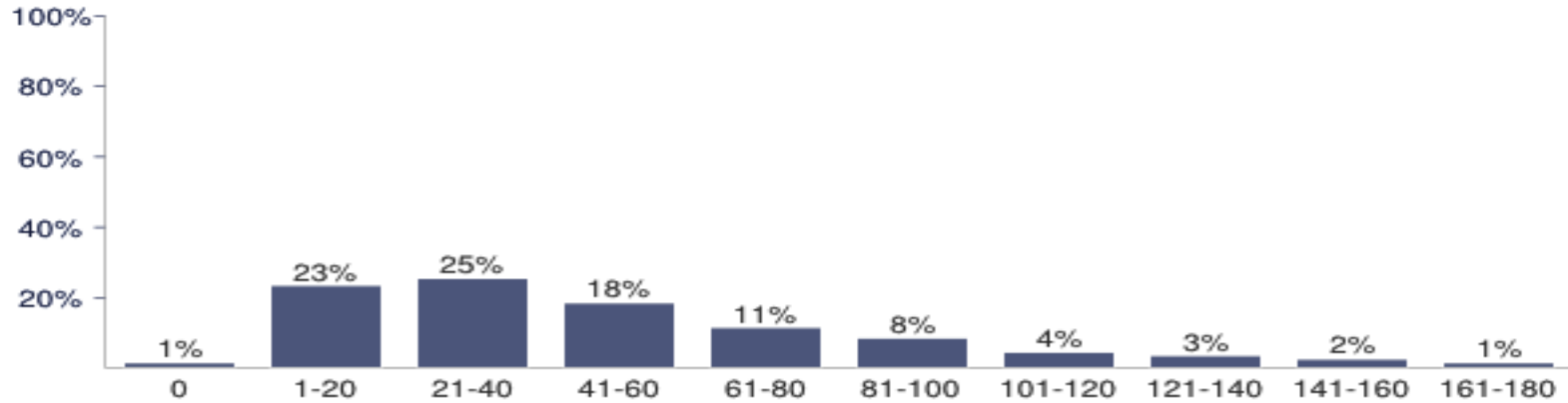
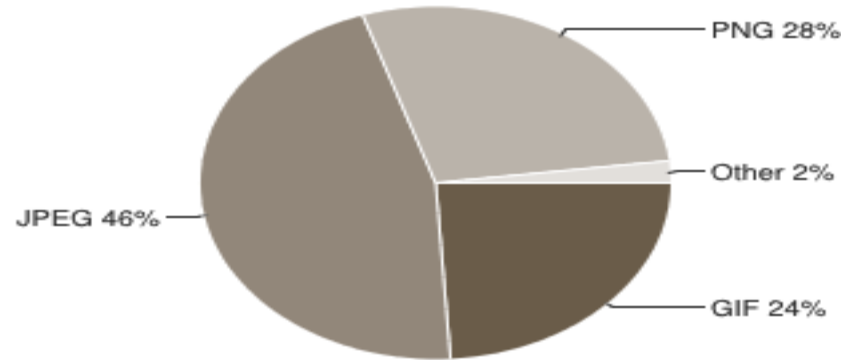


Image Requests by Format



These extensions are typical of promotional and navigational graphics.

Source: <http://archive.org> crawl statistics

JPEG is much more dominant in social media

# File Extensions

Extension	Proportion of Uploads
JPG	96.5 %
CDR	0.94 %
GIF	0.70 %
PNG	0.41 %
ALL OTHERS	2.7 %

Jpeg is the most prevalent uploaded to social networks, due in part because of their suitability for photos, good compression, large palette, and most common availability by mobile devices and consumer cameras.

*Distribution of file extensions in the author's online community.*

# File Extensions

Extension	Proportion of Uploads
JPG	96.5 %

How can we optimize for this?

PNG	0.41 %
ALL OTHERS	2.7 %

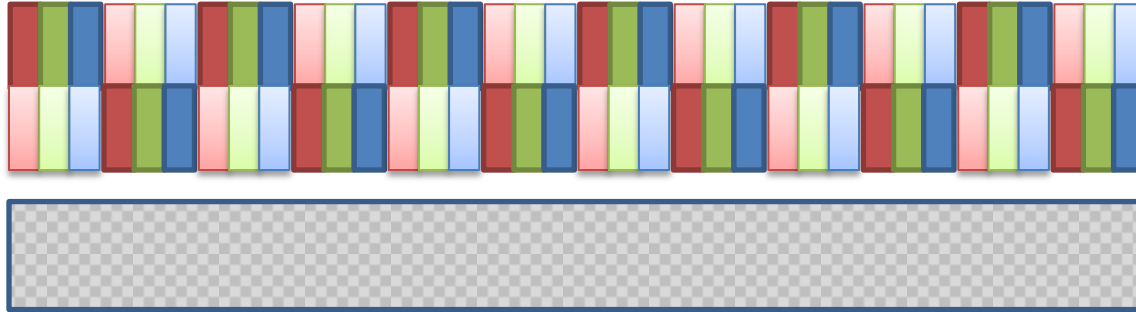
Distribution of file extensions in the author's online community.

# Lossy Compression

(an application of *Psychophysics*...)

## ► RGB Color Space

- (Red, Green, Blue) Pixel Intensity between 0-255



- This representation is costly:  $1024 \times 1024 = 3.1\text{MB}$



RGB Color Space

- ▶ **Luminance**
  - Changes in pixel intensity (brightness)
- ▶ **Chrominance**
  - Color information within the image
- ▶ **Human Visual System (HVS) is much more sensitive to luminance than chrominance.**

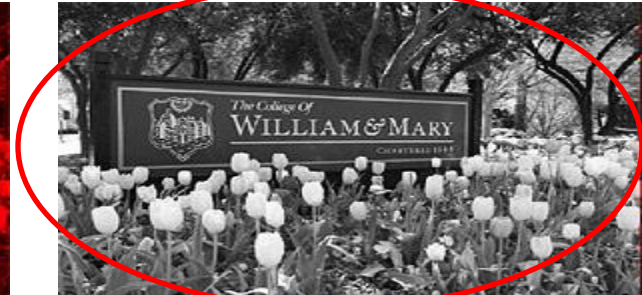
**Lossy Compression (JPEG)**



YCbCr Color Space  
= Luminance (Y)  
+ Chrominance B (Blue-Yellow)  
+ Chrominance R (Red-Green)

RGB

YCbCr



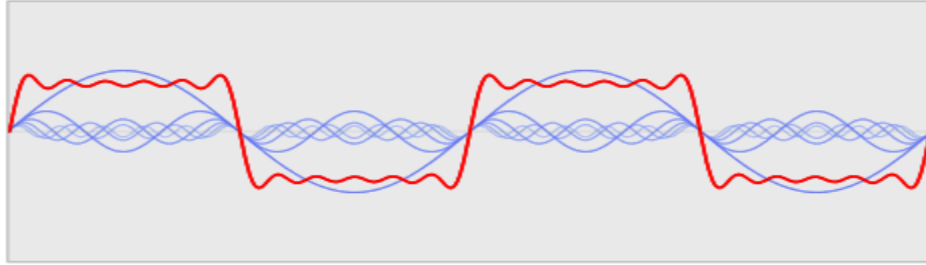
Most detail is in the Y (luminance) component

- ▶ Step 1: Convert from RGB to YCbCr
  - Lossless
- ▶ Step 2: Chrominance Subsampling
  - Replace each 2x2 block of values with an average.
  - Maintain integrity of luminance channel.
  - 75% of chrominance information discarded.
- ▶ Step 3: Discrete Cosine Transform
- ▶ Step 4: Quantization
- ▶ Step 5: Entropy Coding

Lossy Compression (JPEG)

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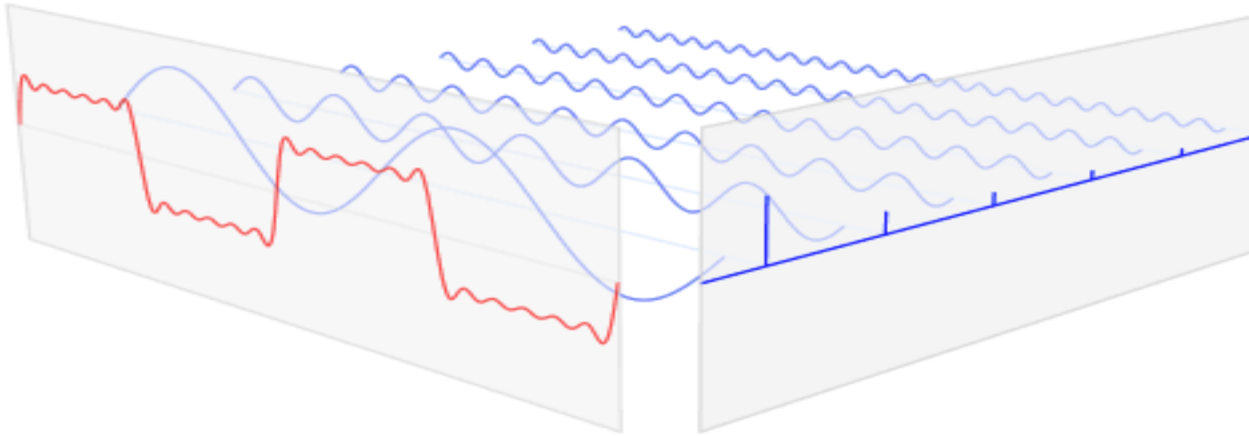
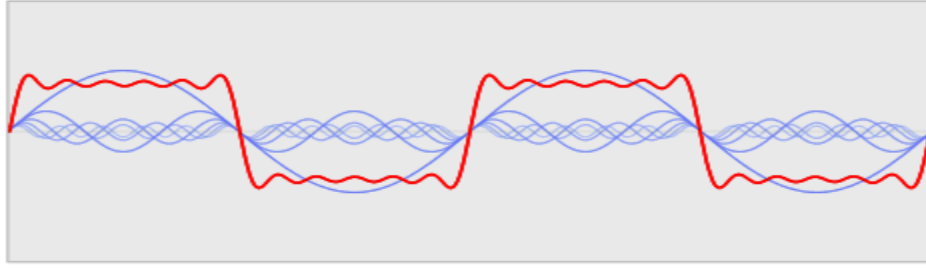
**Lossy Compression (JPEG)**



## ► DCT

- Like FFT, converts a signal from **time** domain to **frequency** domain
- Uses superposition of cosines
- **Red** wave is a superposition of **blue** waves.
  - Each wave has a fixed frequency.
  - Each wave has an amplitude (**coefficient**) that corresponds to the wave's influence.

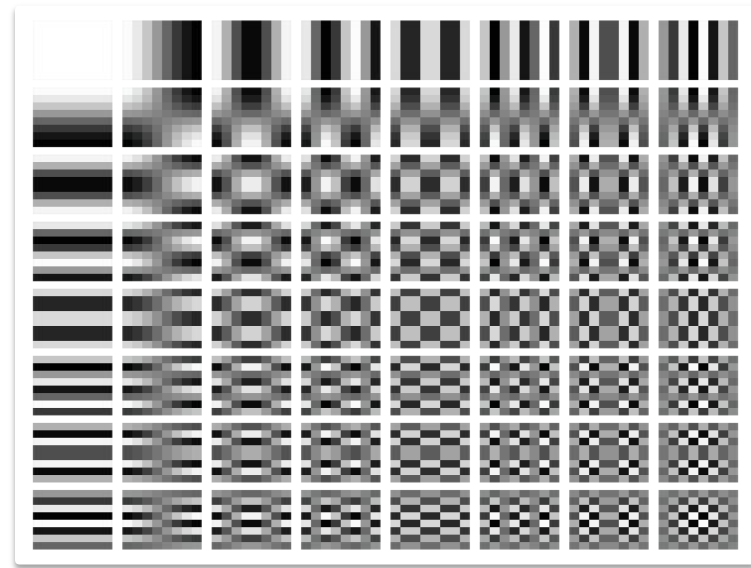
# Step 3: Discrete Cosine Transform



Lower frequency (large) waves are more important to the shape of the signal.

## Step 3: Discrete Cosine Transform

- ▶ Set of 64 basis images
  - JPEG expresses each 8x8 block in an image as a linear combination of these 2D basis functions, each basis image gets assigned a coefficient.



## ▶ Observation

- HVS is more sensitive to weights in the upper left, and less sensitive to lower right (high frequency).

Each channel, YCbCr, is processed using the DCT separately, to produce an 8x8 matrix of coefficients.

$$G = \begin{matrix} & \xrightarrow{u} & & & & & & \\ \begin{matrix} \downarrow v \\ \end{matrix} & \begin{bmatrix} -415.38 & -30.19 & -61.20 & 27.24 & 56.12 & -20.10 & -2.39 & 0.46 \\ 4.47 & -21.86 & -60.76 & 10.25 & 13.15 & -7.09 & -8.54 & 4.88 \\ -46.83 & 7.37 & 77.13 & -24.56 & -28.91 & 9.93 & 5.42 & -5.65 \\ -48.53 & 12.07 & 34.10 & -14.76 & -10.24 & 6.30 & 1.83 & 1.95 \\ 12.12 & -6.55 & -13.20 & -3.95 & -1.87 & 1.75 & -2.79 & 3.14 \\ -7.73 & 2.91 & 2.38 & -5.94 & -2.38 & 0.94 & 4.30 & 1.85 \\ -1.03 & 0.18 & 0.42 & -2.42 & -0.88 & -3.02 & 4.12 & -0.66 \\ -0.17 & 0.14 & -1.07 & -4.19 & -1.17 & -0.10 & 0.50 & 1.68 \end{bmatrix} \end{matrix}$$

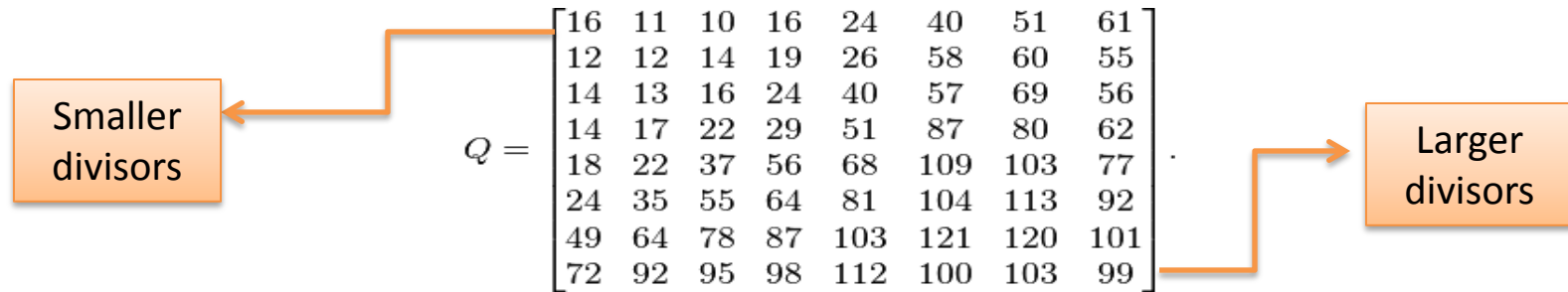
# 2D Discrete Cosine Transform on 8x8 blocks of pixels

- ▶ Step 1: Convert from RGB to YCbCr
  - Lossless
- ▶ Step 2: Chrominance Subsampling
  - Replace each block of 4 pixels with an average
  - Maintains luminance.
  - 75% of chrominance information discarded.
- ▶ Step 3: Discrete Cosine Transform: lossless
- ▶ Step 4: Quantization
- ▶ Step 5: Entropy Coding

**Lossy Compression (JPEG)**

## ► Quantization

- A quantization matrix controls how much information is discarded.



- Dividing our coefficients by  $Q$  causes high-frequency components (lower right) to become mostly zero.

$$B = \begin{bmatrix} -26 & -3 & -6 & 2 & 2 & -1 & 0 & 0 \\ 0 & -2 & -4 & 1 & 1 & 0 & 0 & 0 \\ -3 & 1 & 5 & -1 & -1 & 0 & 0 & 0 \\ -3 & 1 & 2 & -1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

New matrix is easier to compress using entropy (Huffman) coding.



## ► Quantization

- This is where the “lossy” compression occurs.

## ► JPEG Quality

- *Quality factor* scales the quantization matrix. [0..100]%
- The image on the right is at 10% quality, note the sky.
- Manual process.
- Q optimization is combinatorially hard.



JPEG Quality Factor

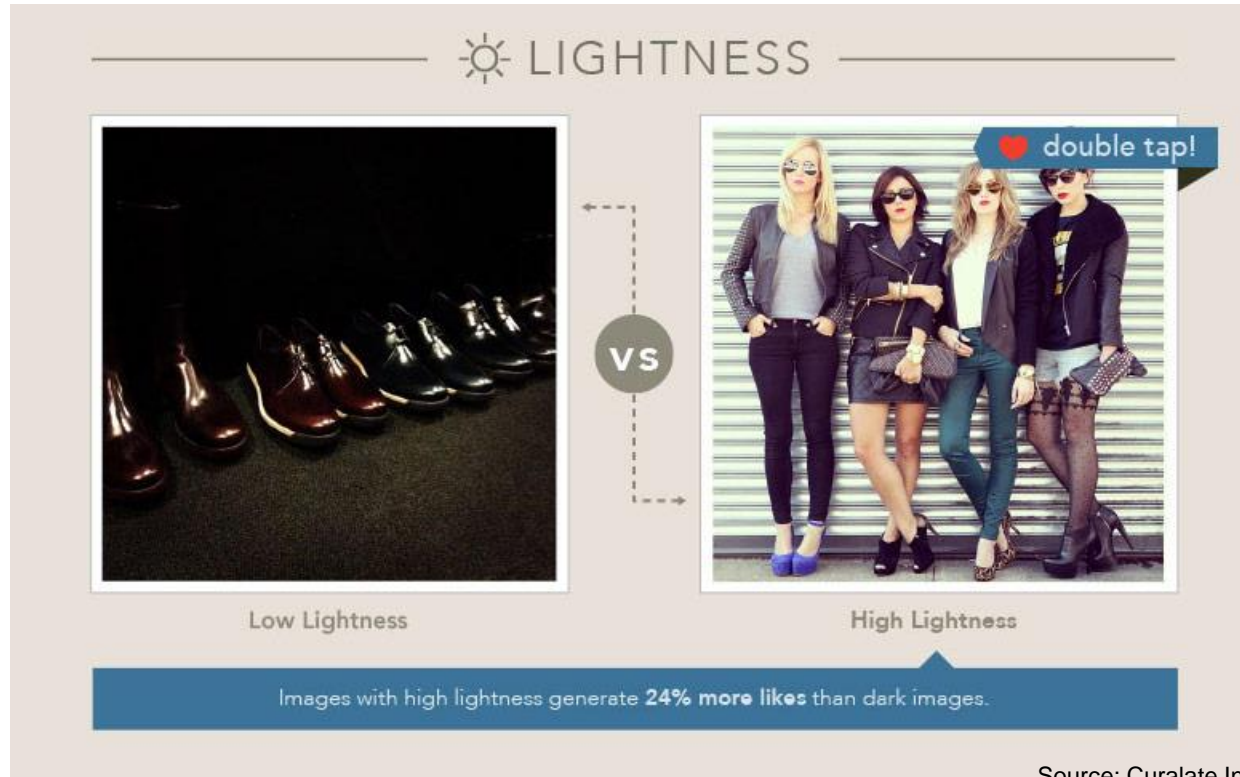
- ▶ Google introduced a new WEBP image format for the web (based on VP8).
  - More smoothing at lower quality settings
  - Produces smaller files than JPEG
  - Uses a similar quality factor (0..100)



**JPEG vs WEBP**

# Psychovisual Enhancements

- ▶ Curalate – Visual Marketing Insights
  - 8,000,000 Instagram photos
  - Data science team correlated features to “likes”



Source: Curalate Inc. social media analysis 2014

# Curalate – Analysis of Instagram

## SATURATION



Moderate Saturation



Low Saturation

Images with low saturation generate **18% more likes** than those with more vibrant colors.

Source: Curalate Inc. social media analysis 2014

# Curalate – Analysis of Instagram



## TEXTURE

❤️ double tap!



Textured Image

VS



Smooth Image

Images with high levels of texture generate **79% more likes** than those without.

Source: Curalate Inc. social media analysis 2014

# Curalate – Analysis of Instagram

- ▶ **JPEGMini** SPIE'11
  - Uses experimentally acquired distributions to reduce quality factor to psychovisual threshold.
  - Works best for large photography archives.
- ▶ **What makes images popular?** WWW'14
  - Analysis of color, gradients, social cues
- ▶ **JPEG quantization table selection by the firefly algorithm (swarm intelligence)** ICMCS'14
- ▶ **Google+ image enhancement** Google
- ▶ **Facebook image compression** Facebook
- ▶ **Curalate – Social Media Image Analysis** Curalate Inc.
- ▶ **Smush.it** Yahoo
  - Removes EXIF metadata, lossless.

. . .

[Related Work]



Empirical Observation



Can we introduce psychovisual  
enhancements to make lossy compression  
more effective?

Research Question

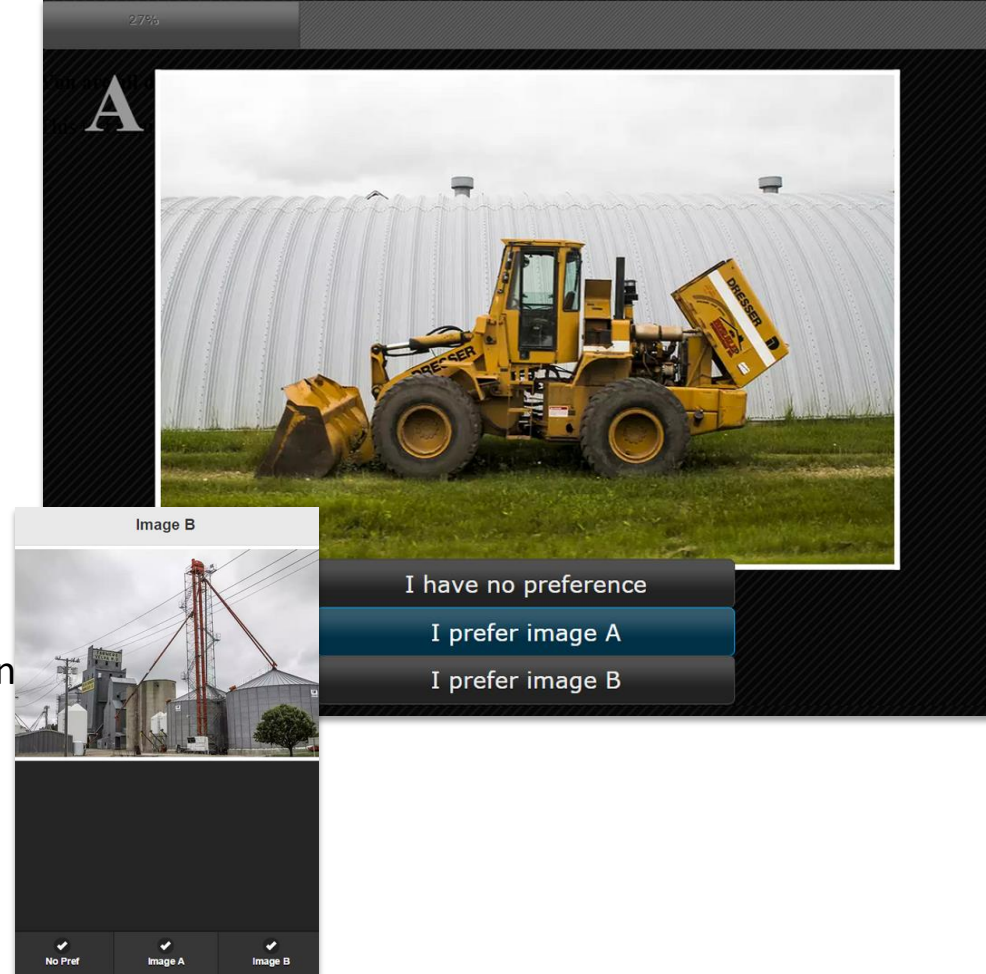


Psychovisual Enhancements

- ▶ 15 images Image Set A
  - Sourced from random selection of photos
  - Indoor and outdoor, scenes and close-ups
  - No human subjects
- ▶ 125,000 images Image Set B
  - Randomly selected from online uploads
- ▶ Explored two psychovisual enhancements:
  - Increase saturation
  - Lighten shadows
- ▶ Explored 6 quality settings
  - 10, 20, 30, 40, 50, and 60
  - 85 is a typical setting used today
  - Many web designers are reluctant to go below 95

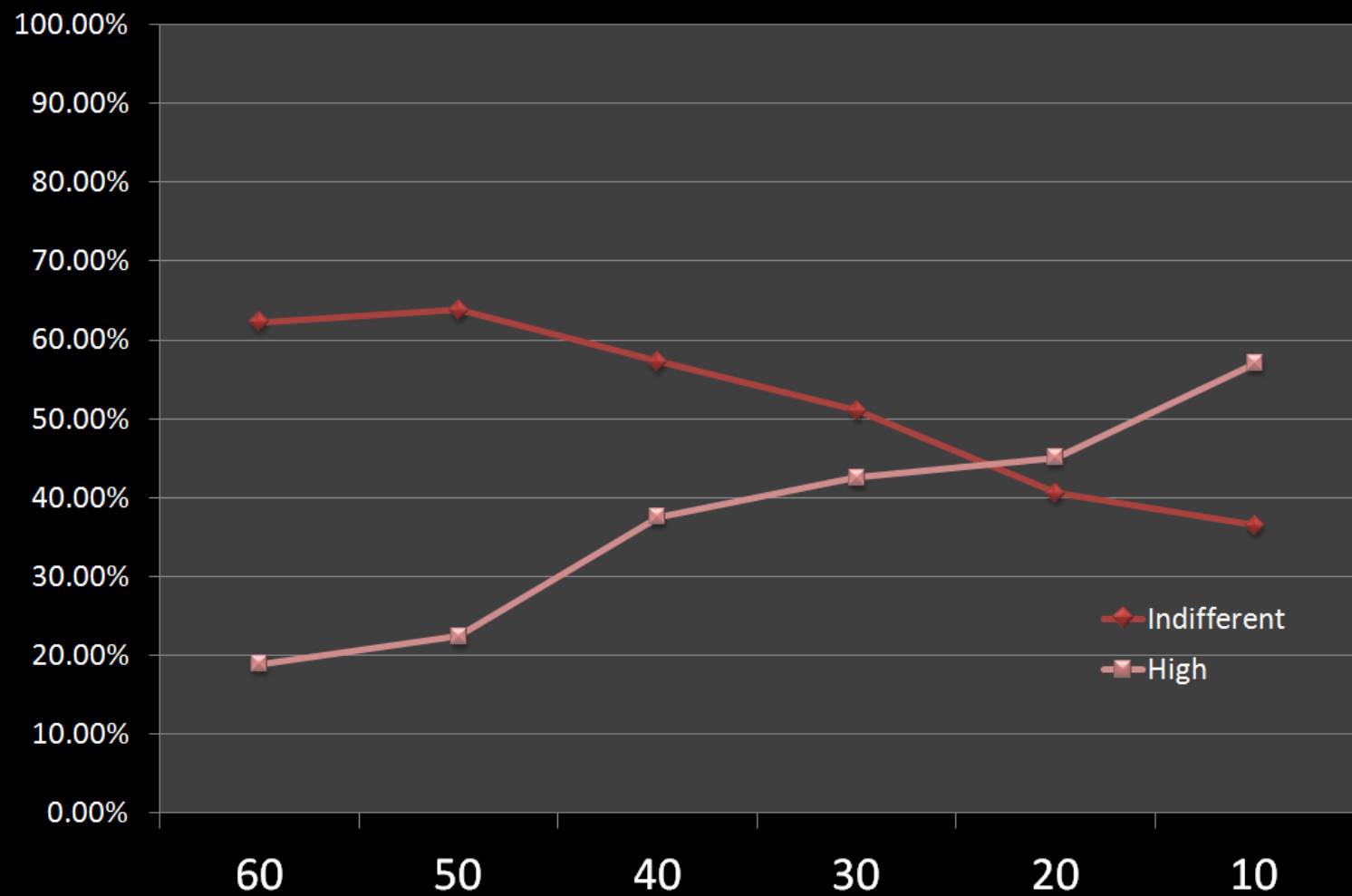
# Experiment

- ▶ Implemented two interfaces
  - Desktop and Mobile
- ▶ Test subjects were not made aware of the nature of the test
- ▶ Order, quality, and sorting of images was randomized
- ▶ Employed DSFCC randomized block design
  - Double Stimulus Forced Choice Comparison

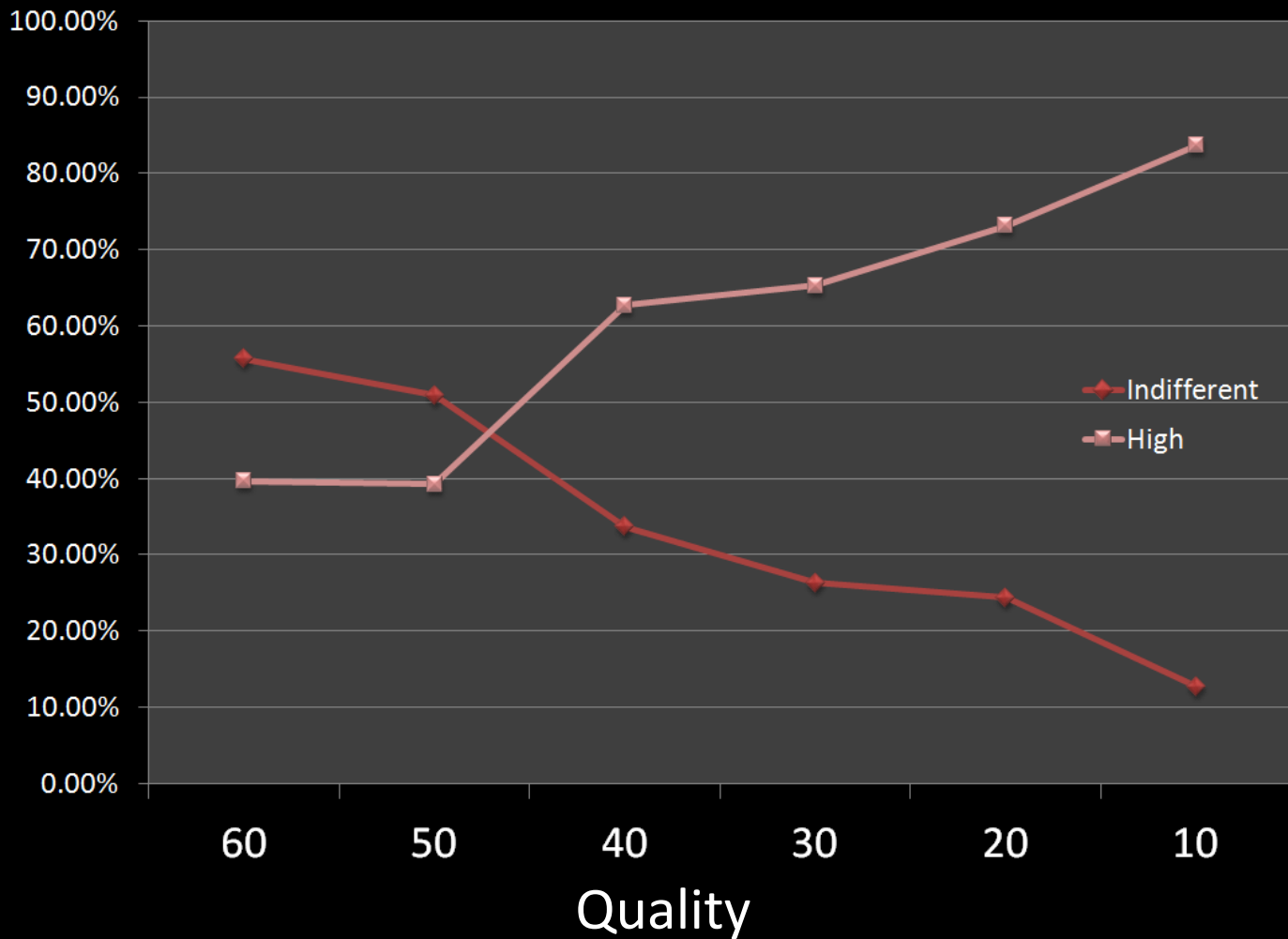


# Experimental Design

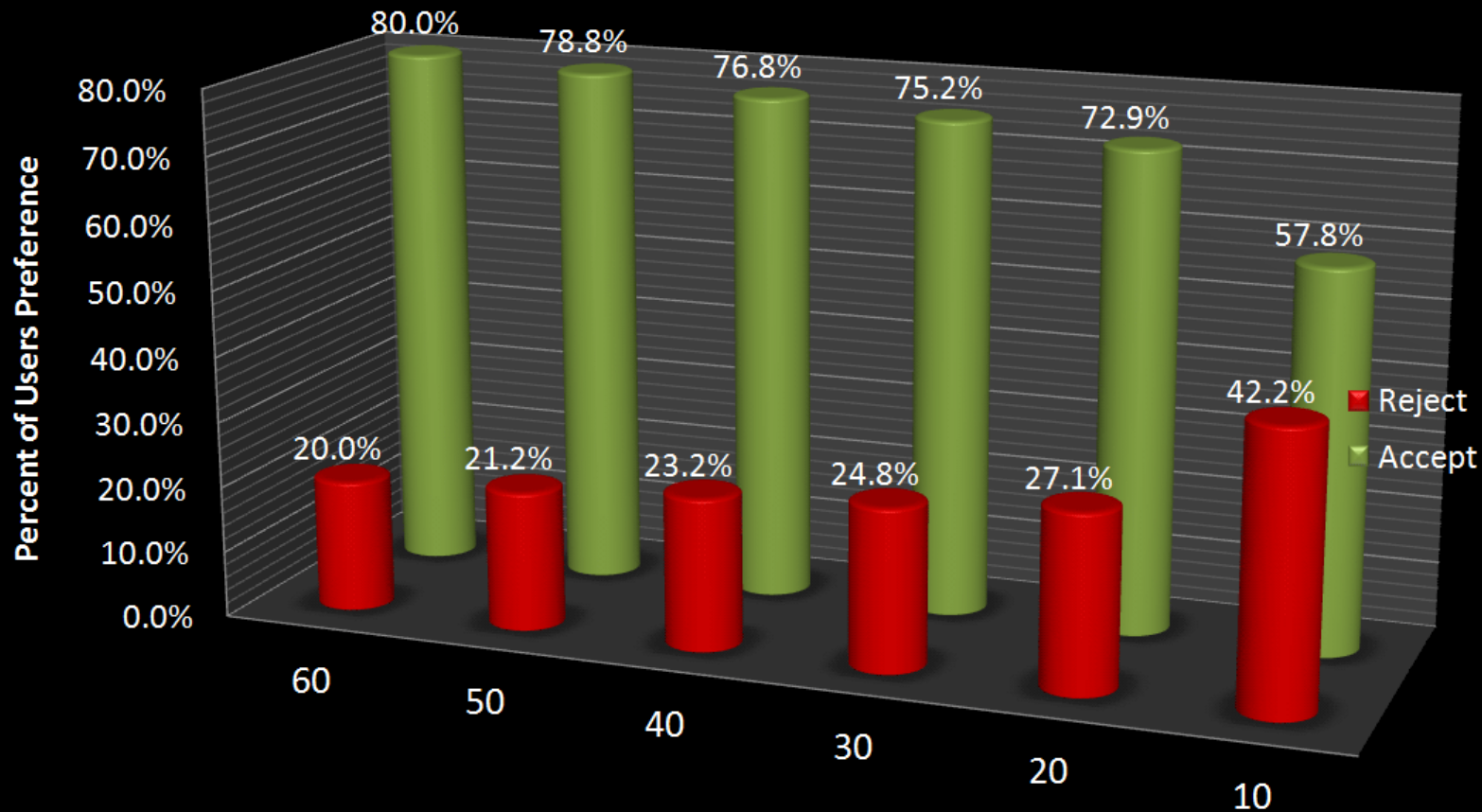
# Baseline Preference and Indifference to High Fidelity



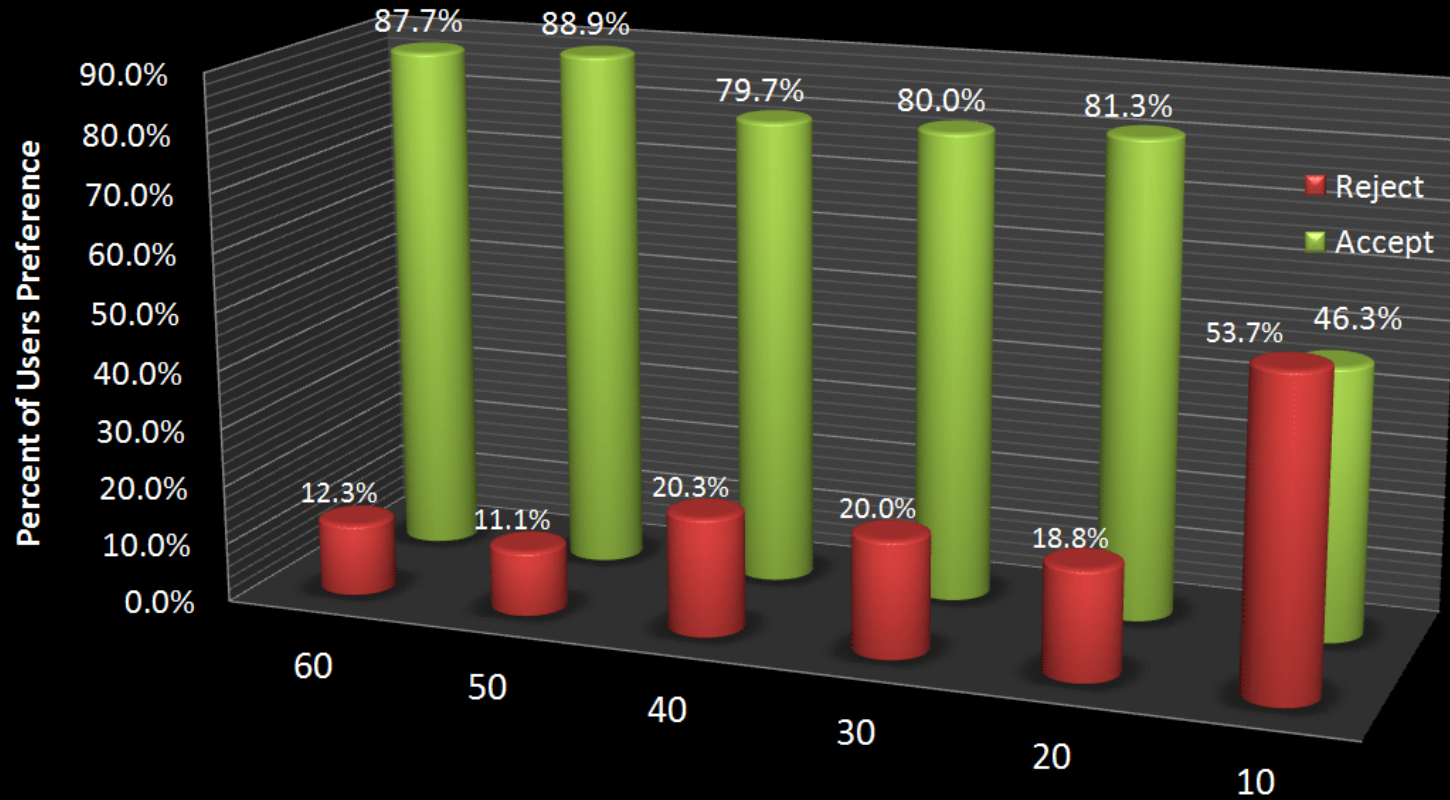
**Baseline Preference and Indifference to High Fidelity - Desktop**



# Scaling with Enhanced Reduced Fidelity (SERF)

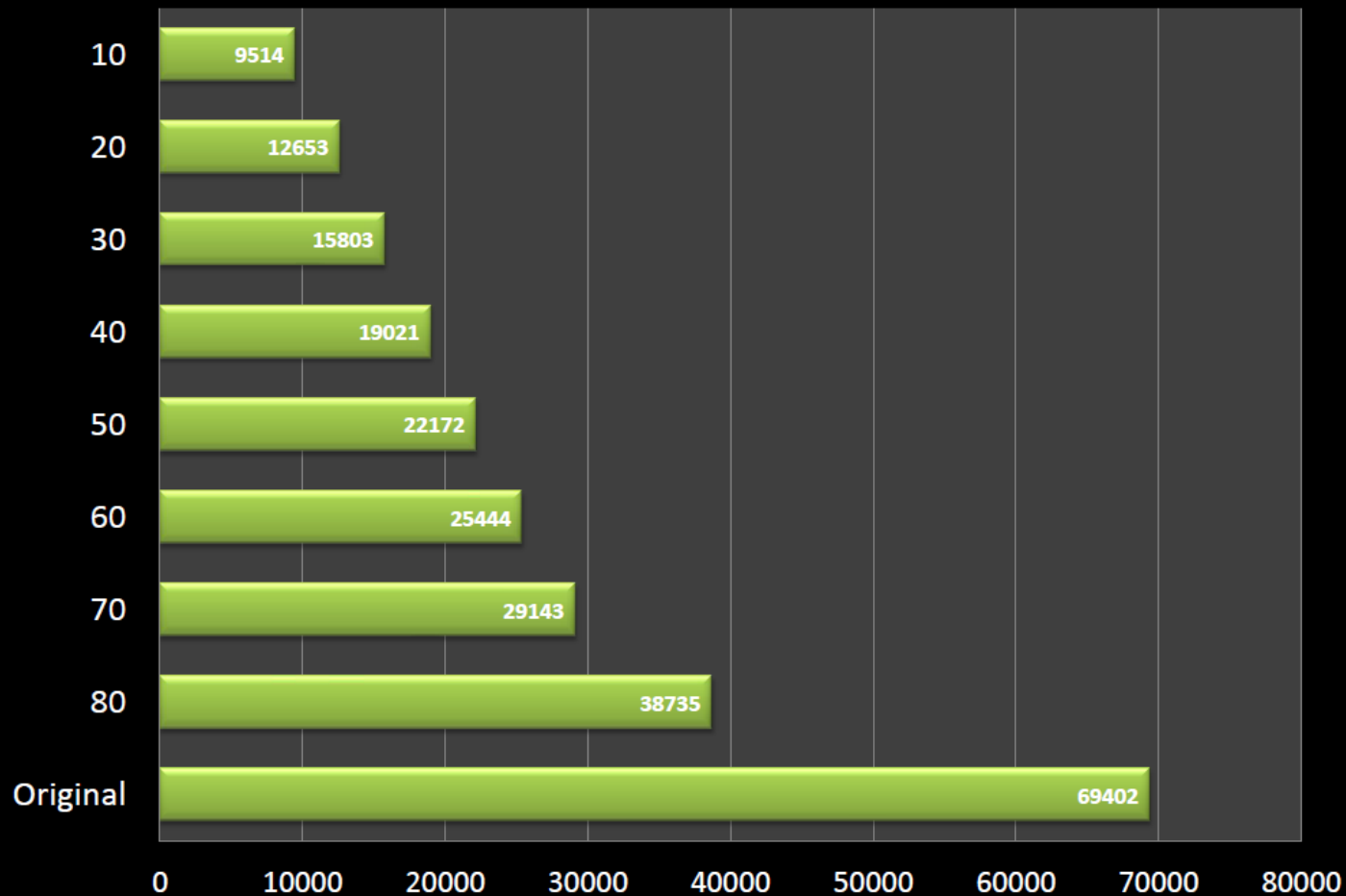


# Scaling with Enhanced Reduced Fidelity (SERF)





**Lossy Compression Effects on Server: 125,000 images**



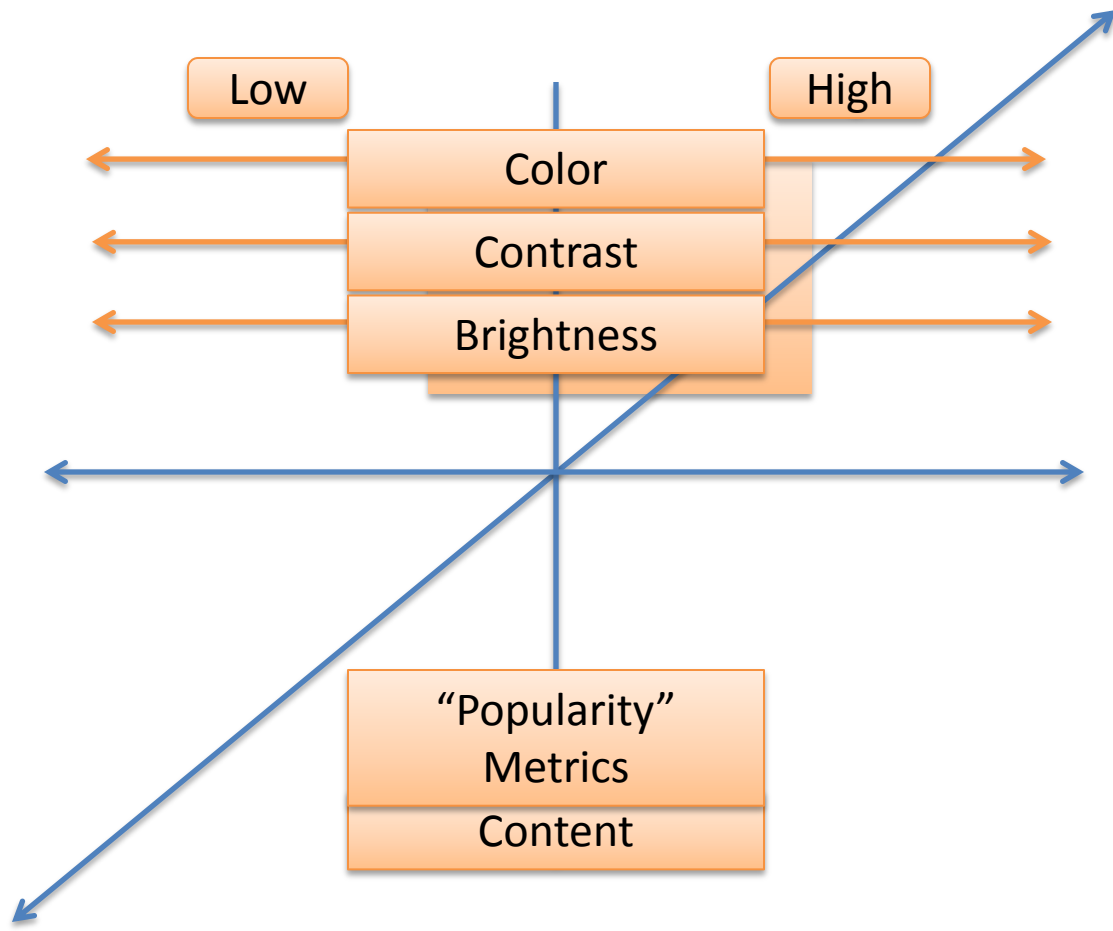
- ▶ Image fidelity can be reduced when psychovisual enhancements are applied.
  - Only 7% point difference between  $q=60$  and  $q=20$  for mobile devices and desktop
- ▶ SERF can enable data centers to scale static image servers by compressing images below the psychovisual threshold.

## Summary

## ► Many exciting directions...

- What psychological effects are at play? Colors?
- Do we apply enhancements indiscriminately, or are there diminishing returns?
- Does the subject matter?
- How does this fare with human subjects?
- Are there stronger enhancements, or are these the best?

Future Work



Many  
Dimensions...

Psychovisual Enhancements to Enable Lossy Compression

► Thank you.

