

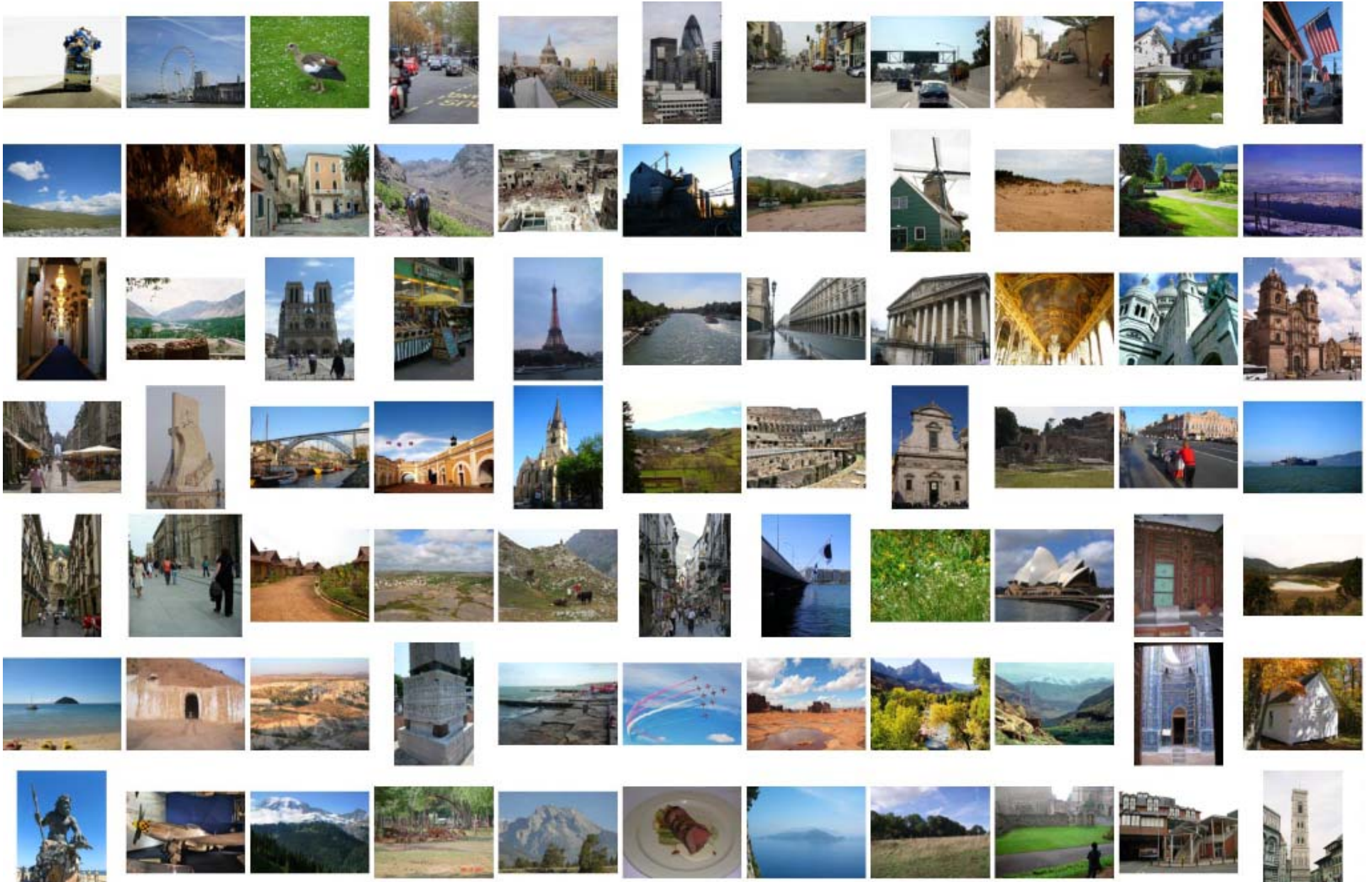
# Photo Tourism and im2gps: 3D Reconstruction and Geolocation of Internet Photo Collections

## *Part II*

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Cornell

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CMU  
MIT (Fall 2009)  
Brown (Spring 2010 - )

# Complexity of the Visual World



# The Internet as a Data Source

- Social Networking Sites (e.g. Facebook, MySpace)
- Image Search Engines (e.g. Google, Bing)
- Photo Sharing Sites (e.g. Flickr, Picasa, Panoramio, photo.net, dpchallenge.com)
- Computer Vision Databases (e.g. CalTech 256, PASCAL VOC, LabelMe, Tiny Images, image-net.org, ESP game, Squigl, Matchin)

# How Big is Flickr?

- As of June 19<sup>th</sup>, 2009
- Total content:
  - 3.6 billion photographs
  - 100+ million geotagged images
- *Public* content:
  - 1.3 billion photographs
  - 74 million geotagged images



# How Annotated is Flickr? (tag search)

- Party – 7,355,998
- Paris – 4,139,927
- Chair – 232,885
- Violin – 55,015
- Trashcan – 9,818

# Trashcan Results



From [PoPPaP](#)



From [howlinhill](#)



From [Jennay Jazz](#)



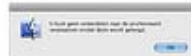
From [Norma Tub](#)



From [ianjacobs](#)



From [ella novak](#)



From [bertboerland](#)



From [m114dy](#)



From [ccharland](#)



From [wallyq](#)



From [Patrik Moen](#)



From [dakota.morri...](#)



From [Jimmy...](#)



From [PavelsDog](#)

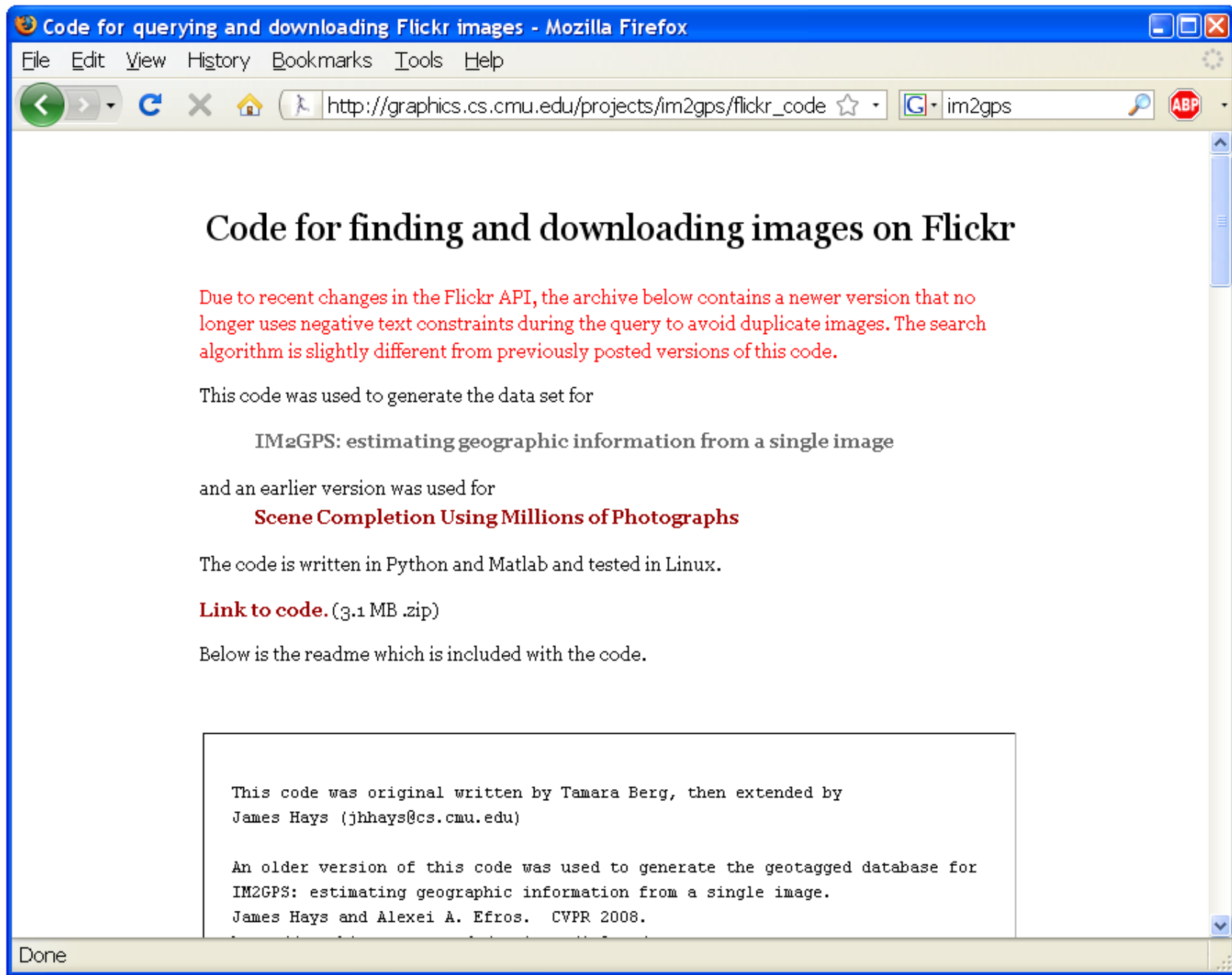


From [ilovecoffee...](#)



From [Daquella...](#)

- <http://www.flickr.com/search/?q=trashcan+NOT+party&m=tags&z=t&page=5>



- [http://graphics.cs.cmu.edu/projects/im2gps/flickr\\_code.html](http://graphics.cs.cmu.edu/projects/im2gps/flickr_code.html)

## flickr.photos.search

Return a list of photos matching some criteria. Only photos visible to the calling user will be returned. To return private or semi-private photos, the caller must be authenticated with 'read' permissions, and have permission to view the photos. Unauthenticated calls will only return public photos.

### Authentication

This method does not require authentication.

### Arguments

#### **api\_key** (Required)

Your API application key. [See here](#) for more details.

#### **user\_id** (Optional)

The NSID of the user whose photo to search. If this parameter isn't passed then everybody's public photos will be searched. A value of "me" will search against the calling user's photos for authenticated calls.

#### **tags** (Optional)

A comma-delimited list of tags. Photos with one or more of the tags listed will be returned.

#### **tag\_mode** (Optional)

Either 'any' for an OR combination of tags, or 'all' for an AND combination. Defaults to 'any' if not specified.

#### **text** (Optional)

A free text search. Photos whose title, description or tags contain the text will be returned.

#### **min\_upload\_date** (Optional)

Minimum upload date. Photos with an upload date greater than or equal to this value will be returned. The date should be in the form of a unix timestamp.

#### **max\_upload\_date** (Optional)

Maximum upload date. Photos with an upload date less than or equal to this value will be returned. The date should be in the form of a unix timestamp.

#### **min\_taken\_date** (Optional)

Minimum taken date. Photos with a taken date greater than or equal to this value will be returned. The date should be in the form of a mysql datetime.

#### **max\_taken\_date** (Optional)

Maximum taken date. Photos with a taken date less than or equal to this value will be returned. The date should be in the form of a mysql datetime.

Undocumented  
tag\_mode: bool  
e.g. tags = cat NOT tiger

#### **license** (Optional)

The license id for photos (for possible values see the flickr.photos.licenses.getInfo method). Multiple licenses may be comma-separated.

#### **sort** (Optional)

The order in which to sort returned photos. Defaults to date-posted-desc (unless you are doing a radial geo query, in which case the default sorting is by ascending distance from the point specified). The possible values are: date-posted-asc, date-posted-desc, date-taken-asc, date-taken-desc, interestingness-desc, interestingness-asc, and relevance.

#### **privacy\_filter** (Optional)

Return photos only matching a certain privacy level. This only applies when making an authenticated call to view photos you own. Valid values are:

- 1 public photos
- 2 private photos visible to friends
- 3 private photos visible to family
- 4 private photos visible to friends & family
- 5 completely private photos

#### **bbox** (Optional)

A comma-delimited list of 4 values defining the Bounding Box of the area that will be searched.

The 4 values represent the bottom-left corner of the box and the top-right corner, minimum\_longitude, minimum\_latitude, maximum\_longitude, maximum\_latitude.

Longitude has a range of -180 to 180, latitude of -90 to 90. Defaults to -180, -90, 180, 90 if not specified.

Unlike standard photo queries, geo (or bounding box) queries will only return 250 results per page.

Geo queries require some sort of limiting agent in order to prevent the database from crying. This is basically like the check against "parameterless searches" for queries without a geo component.

A tag, for instance, is considered a limiting agent as are user defined min\_date\_taken and min\_date\_upload parameters — If no limiting factor is passed we return only photos added in the last 12 hours (though we may extend the limit in the future).

#### **accuracy** (Optional)

Recorded accuracy level of the location information. Current range is 1-16 :

- World level is 1
- Country is ~3
- Region is ~6
- City is ~11
- Street is ~16

Defaults to maximum value if not specified.

#### **safe\_search** (Optional)

Safe search setting:

- 1 for safe.
- 2 for moderate



**media** (Optional)

Filter results by media type. Possible values are `all` (default), `photos` or `videos`

**has\_geo** (Optional)

Any photo that has been geotagged, or if the value is "0" any photo that has *not* been geotagged.

Geo queries require some sort of limiting agent in order to prevent the database from crying. This is basically like the check against "parameterless searches" for queries without a geo component.

A tag, for instance, is considered a limiting agent as are user defined `min_date_taken` and `min_date_upload` parameters &mdash; If no limiting factor is passed we return only photos added in the last 12 hours (though we may extend the limit in the future).

**geo\_context** (Optional)

Geo context is a numeric value representing the photo's geotagging beyond latitude and longitude. For example, you may wish to search for photos that were taken "indoors" or "outdoors".

The current list of context IDs is :

- **0**, not defined.
- **1**, indoors.
- **2**, outdoors.

Geo queries require some sort of limiting agent in order to prevent the database from crying. This is basically like the check against "parameterless searches" for queries without a geo component.

A tag, for instance, is considered a limiting agent as are user defined `min_date_taken` and `min_date_upload` parameters &mdash; If no limiting factor is passed we return only photos added in the last 12 hours (though we may extend the limit in the future).

**lat** (Optional)

A valid latitude, in decimal format, for doing radial geo queries.

Geo queries require some sort of limiting agent in order to prevent the database from crying. This is basically like the check against "parameterless searches" for queries without a geo component.

A tag, for instance, is considered a limiting agent as are user defined `min_date_taken` and `min_date_upload` parameters &mdash; If no limiting factor is passed we return only photos added in the last 12 hours (though we may extend the limit in the future).

**lon** (Optional)

A valid longitude, in decimal format, for doing radial geo queries.

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A tag, for instance, is considered a limiting agent as are user defined min\_date\_taken and min\_date\_upload parameters &mdash; If no limiting factor is passed we return only photos added in the last 12 hours (though we may extend the limit in the future).

**radius** (Optional)

A valid radius used for geo queries, greater than zero and less than 20 miles (or 32 kilometers), for use with point-based geo queries. The default value is 5 (km).

**radius\_units** (Optional)

The unit of measure when doing radial geo queries. Valid options are "mi" (miles) and "km" (kilometers). The default is "km".

**is\_commons** (Optional)

Limit the scope of the search to only photos that are part of the [Flickr Commons project](#). Default is false.

**extras** (Optional)

A comma-delimited list of extra information to fetch for each returned record. Currently supported fields are: license, date\_upload, date\_taken, owner\_name, icon\_server, original\_format, last\_update, geo, tags, machine\_tags, o\_dims, views, media.

**per\_page** (Optional)

Number of photos to return per page. If this argument is omitted, it defaults to 100. The maximum allowed value is 500.

**page** (Optional)

The page of results to return. If this argument is omitted, it defaults to 1.

## Example Response

This method returns the standard photo list xml:

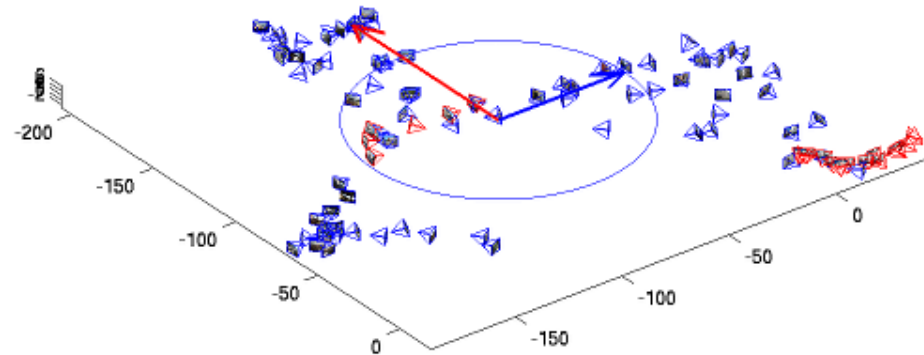
```
<photos page="2" pages="89" perpage="10" total="881">
  <photo id="2636" owner="47058503995@N01"
    secret="a123456" server="2" title="test_04"
    ispublic="1" isfriend="0" isfamily="0" />
  <photo id="2635" owner="47058503995@N01"
    secret="b123456" server="2" title="test_03"
    ispublic="0" isfriend="1" isfamily="1" />
  <photo id="2633" owner="47058503995@N01"
    secret="c123456" server="2" title="test_01"
    ispublic="1" isfriend="0" isfamily="0" />
  <photo id="2610" owner="12037949754@N01"
    secret="d123456" server="2" title="00_tall"
    ispublic="1" isfriend="0" isfamily="0" />
</photos>
```

To map <photo> elements to urls, please read the [url documentation](#).

# Global Image Geolocation

# Background: Image Localization

- *Where Am I?* Contest. ICCV 2005. Organized by Richard Szeliski.
- *Nokia Challenge: Where was this Photo Taken, and How?* Multimedia Grand Challenge 2009.



# Background: Image Localization

- Zhang and Kosecka. *Image Based Localization in Urban Environments*. I3DPVT 2006.
- Philbin, Chum, Isard, Sivic, and Zisserman. *Lost in Quantization: Improving Particular Object Retrieval in Large Scale Image Databases*. CVPR 2008



# Global Image Geolocation

- Hays and Efros. *Im2gps: estimating geographic information from a single image*. CVPR 2008.
- Quack, Leibe, Van Gool. *World-Scale Mining of Objects and Events from Community Photo Collections*. CIVR 2008
- Crandall, Backstrom, Huttenlocher, and Kleinberg. *Mapping the World's Photos*. WWW 2009.

# im2gps

- Big Question: is geolocation just instance-level, landmark recognition or can you reason about location and geography from non-specific scene properties?

# Where is this?



# Where is this?



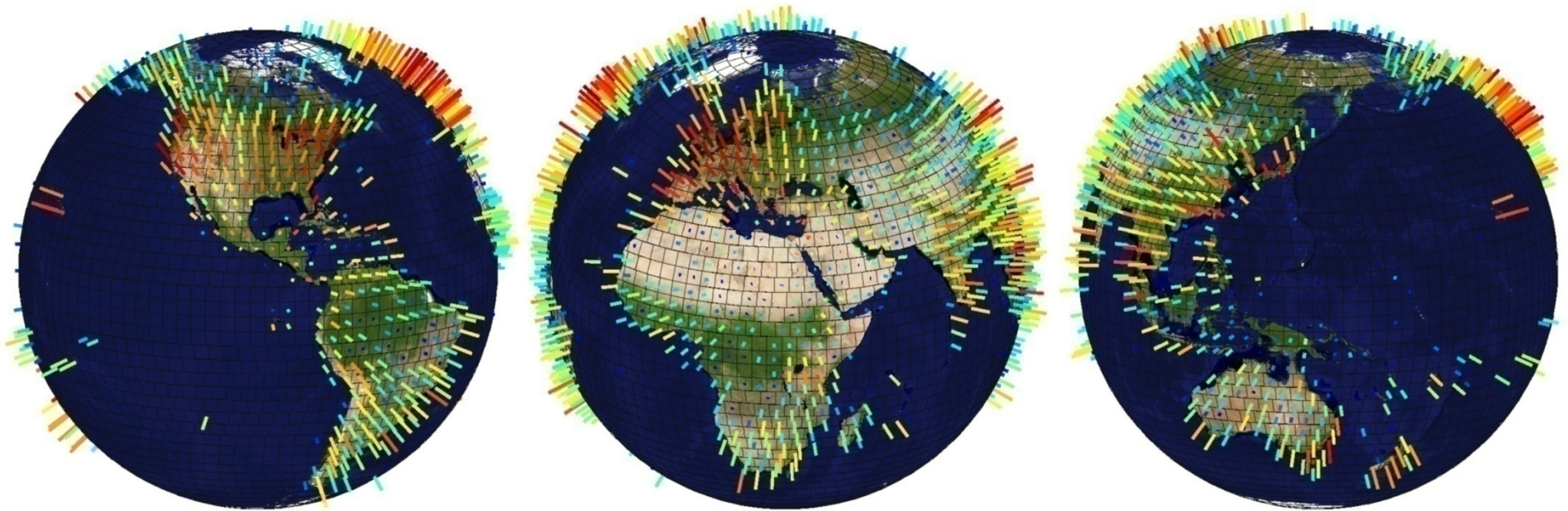


# Where is this?





# im2gps Geographic Photo Density



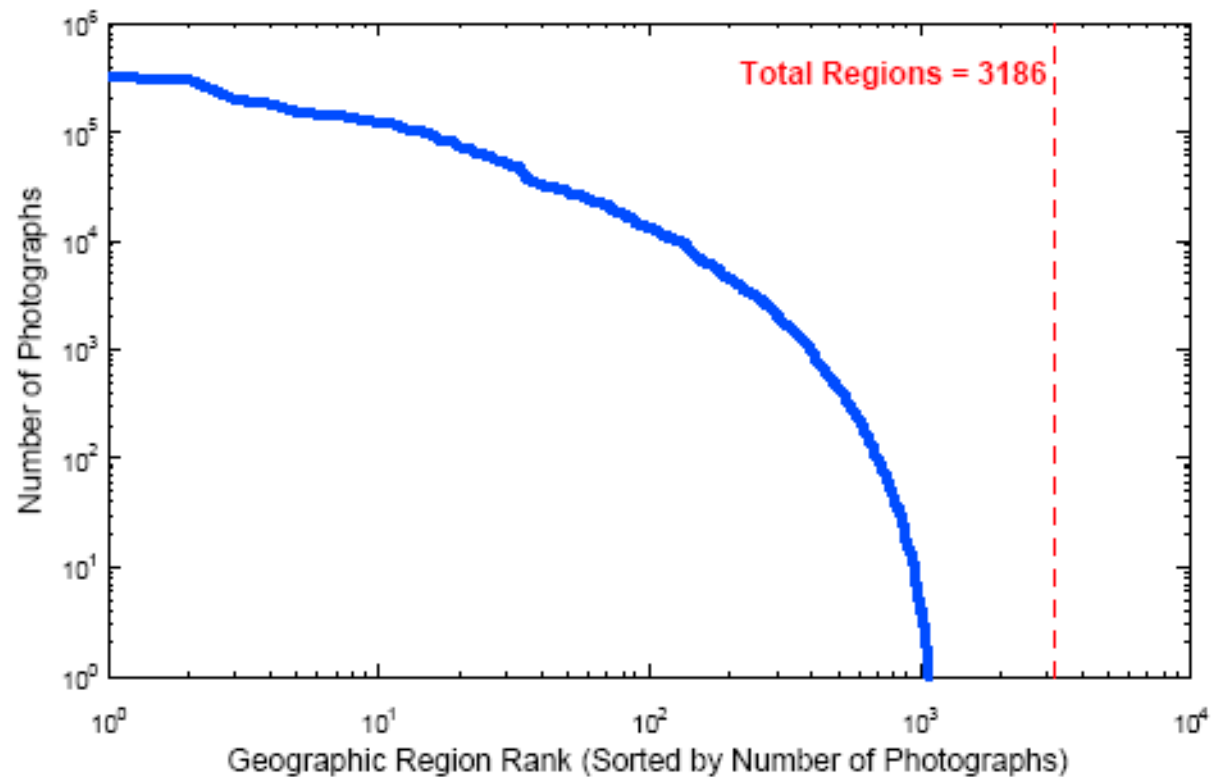
6.4 mil. photos by 110K photographers.

1 TB of visual data.

Photographs had at least one place keyword.

Photos average ~1 content descriptive keyword.

# Photo Region Frequencies



# Im2gps Image Features

- **Gist descriptor** – 5x5 spatial resolution, 4 scales, 8 orientations. [Code](#)
- **Tiny Color Image** – 5x5 and 16x16 spatial resolutions.
- **Color Histogram** – L\*A\*B\* 4x14x14 histograms.
- **Texon Histogram** – 512 entry, filter bank based. [Code](#)
- **Line Features** – Histograms of straight line lengths and angles.
- **Geometric Context** – 8x8 probability of geometric class (e.g. Ground, Sky, Vertical, Porous). [Code](#)
- Histograms are compared with Chi Squared measure, other features with L1 distance.

# Scene Matches



Madrid



england



France



Paris



Croatia



heidelberg



Macau



Malta



Cairo



Italy



Italy



Italy



Latvia



europe



Barcelona



Austria







# Scene Matches



Paris



Paris



Paris



Paris



Paris



Paris



Paris



Madrid



Rome



Paris



Cuba



Paris



Paris



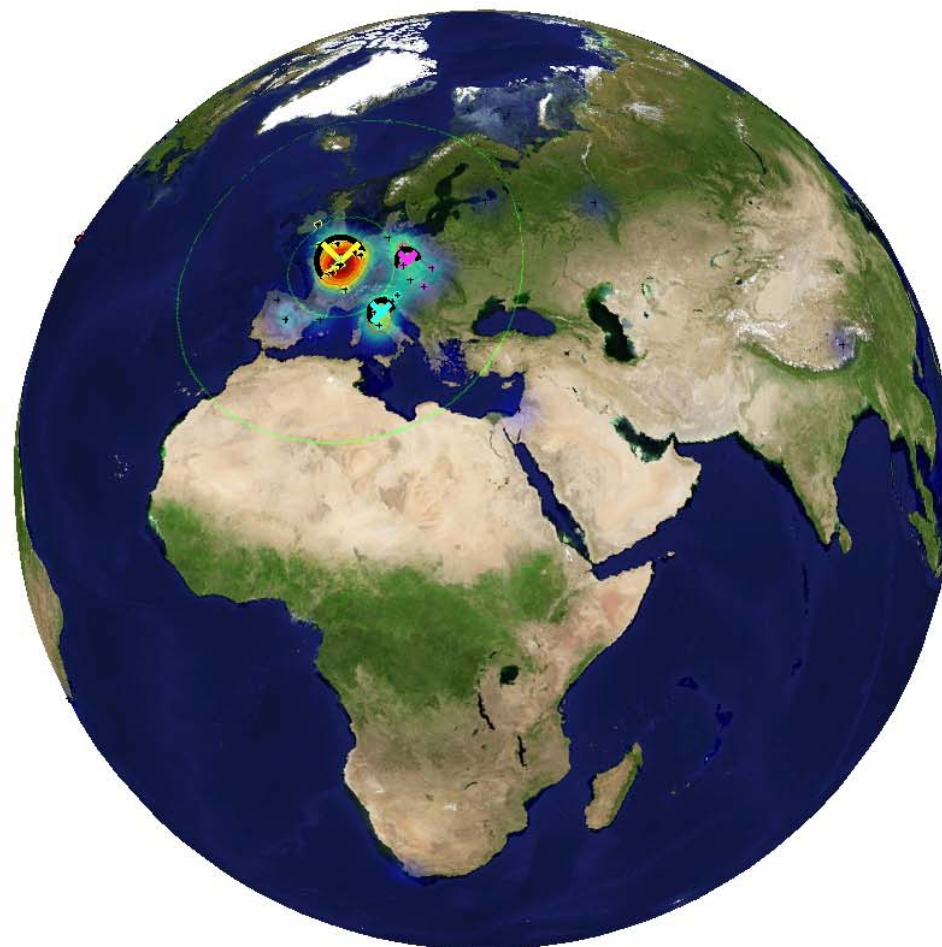
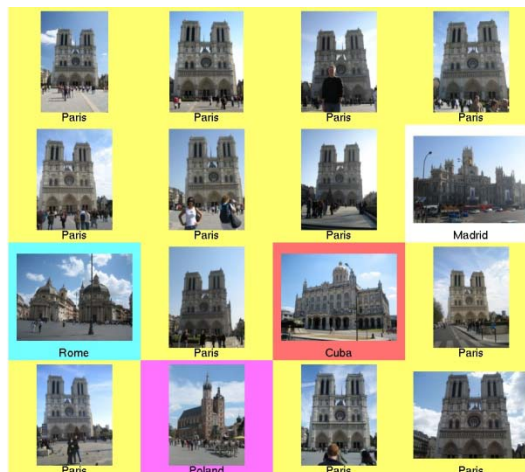
Poland



Paris



Paris





# Scene Matches



Philippines



Houston



Thailand



Houston



Maldives



Philippines



NewZealand



Bermuda



Palau



Mexico2



Brazil



Mendoza



Brazil



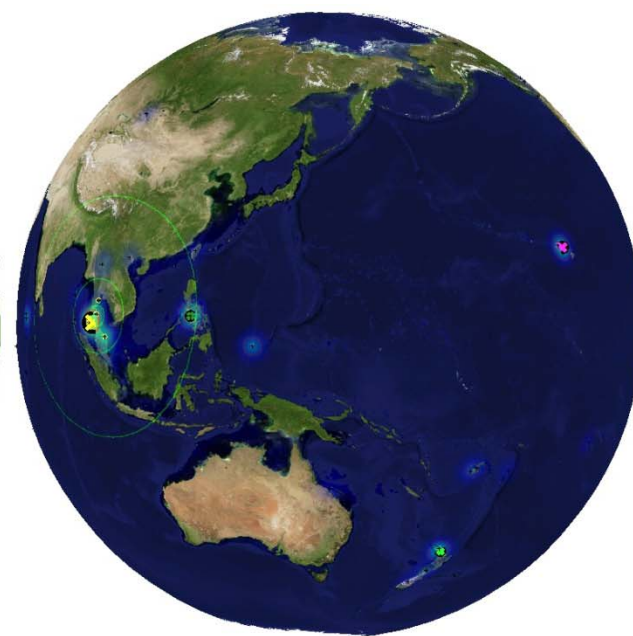
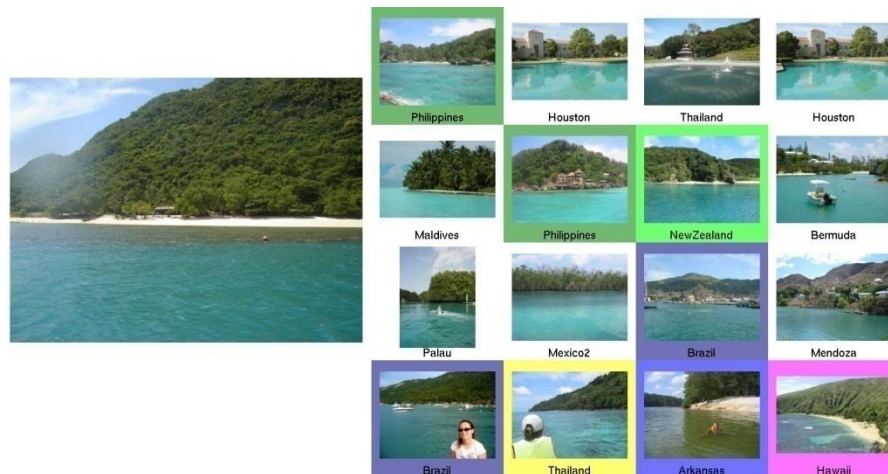
Thailand



Arkansas

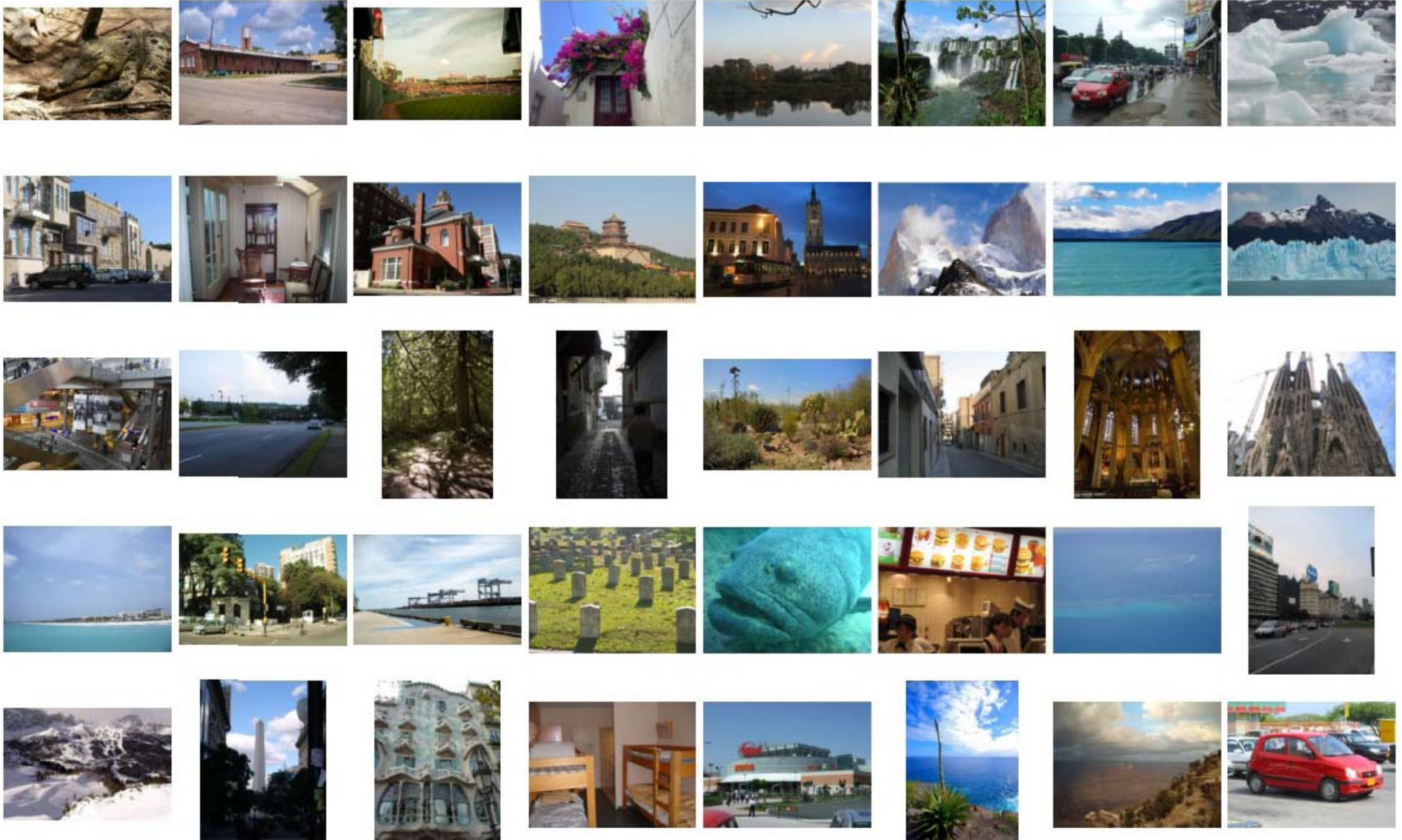


Hawaii

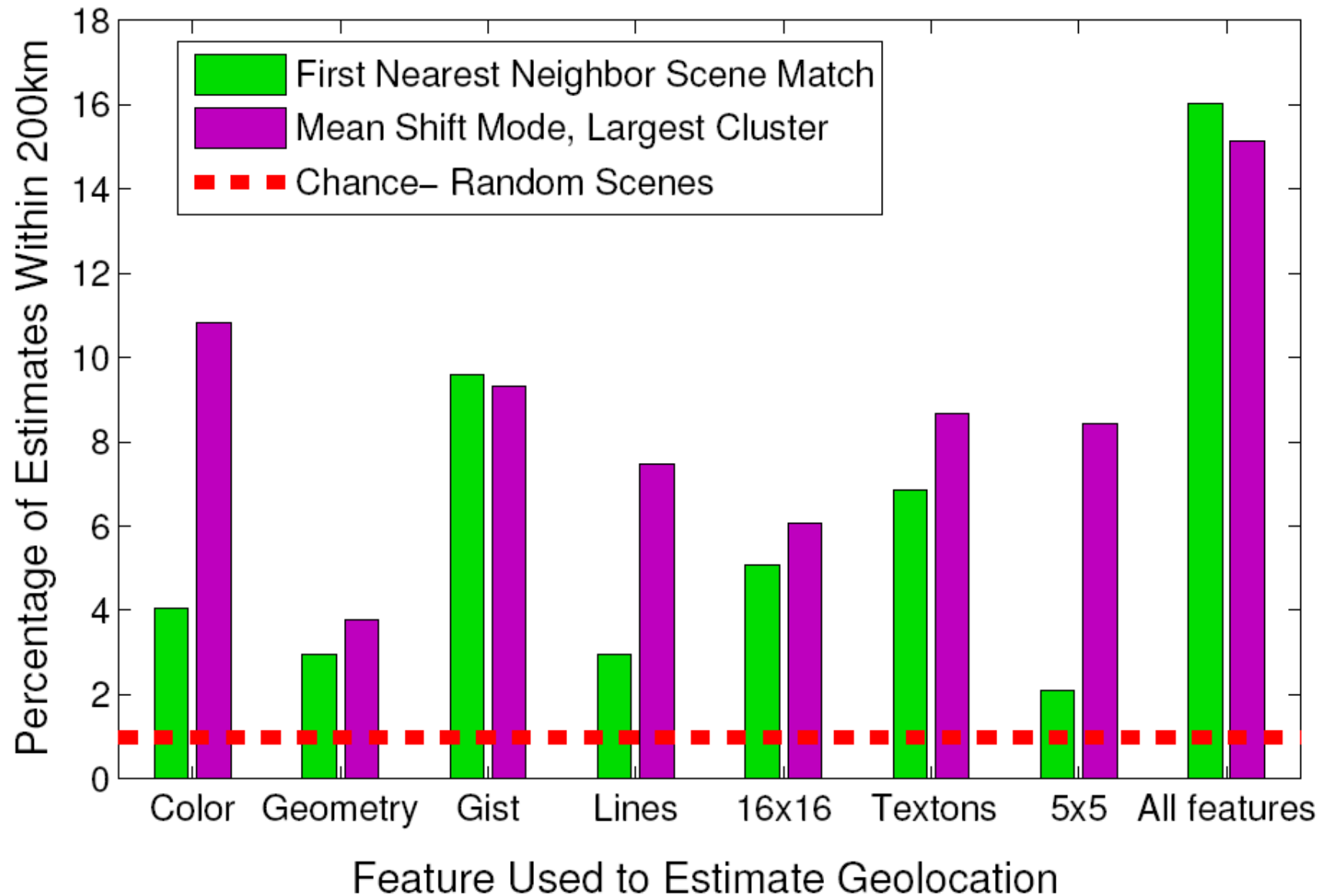




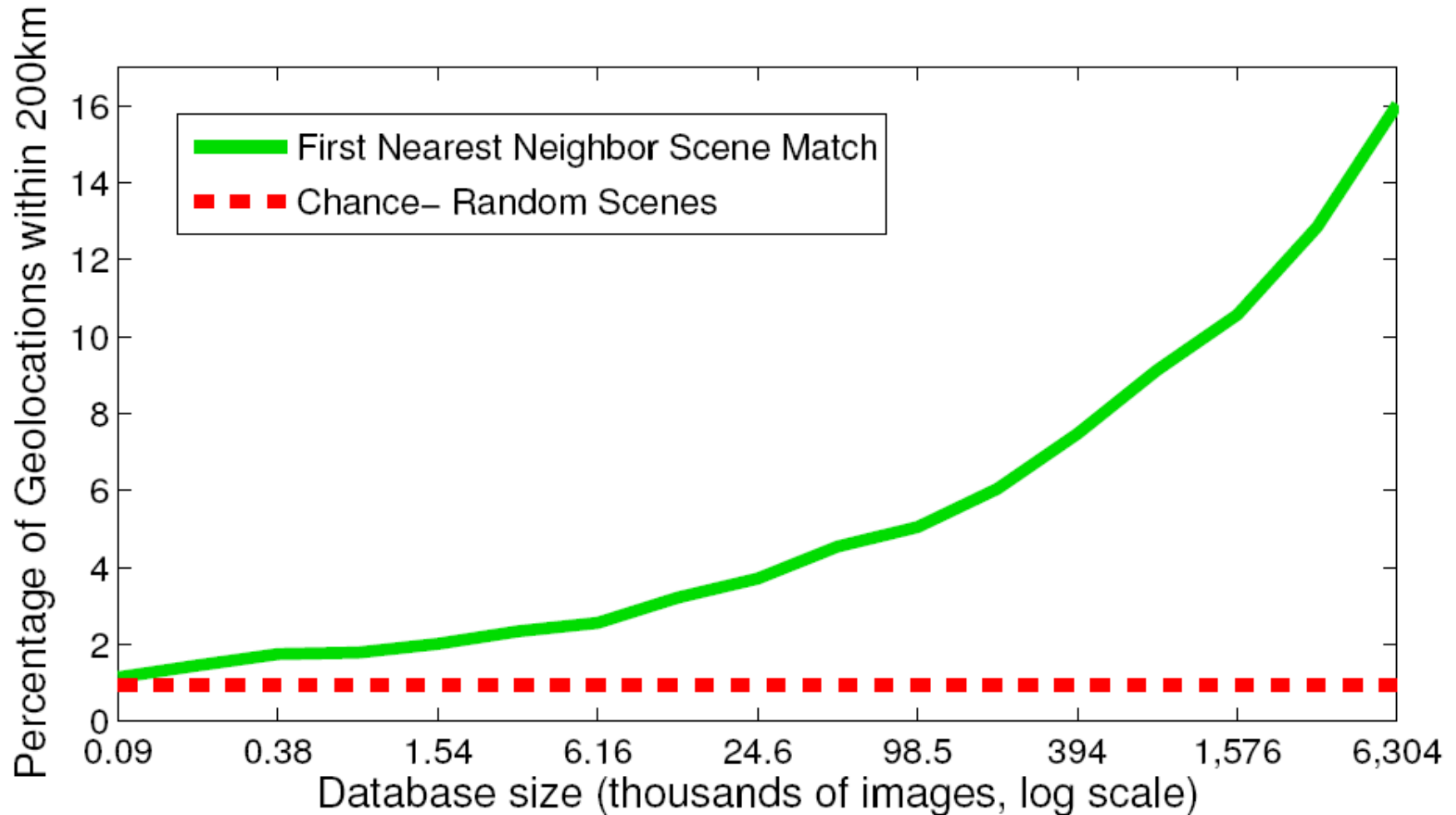
# Quantitative Evaluation Test Set



# Feature Performance



# The Importance of Data





# im2gps Feature Problems

- im2gps features aren't very robust to minor scene differences (e.g. translation, cropping, illumination change).
- im2gps features aren't ideal for instance-level landmark recognition.

# im2gps Feature Problems



# Scene Geometry Derived Features

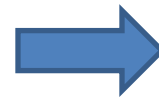
Gist at Geometric  
Region of  
Interest



Sky: Color and Texton Histograms



Porous: Color and Texton Histograms



Vertical: Color and Texton Histograms



Ground: Color and Texton Histograms

# im2gps Feature Problems



Australia



Australia



Paris



WashingtonDC



Sydney



Madrid



Spain



London



Mississippi



Venezuela



Spain



Belgium



NewZealand



Finland



London



Spain

# Large Vocabulary Bag of SIFTs

- Inspired by Philbin, Chum, Isard, Sivic, and Zisserman. *Lost in Quantization: Improving Particular Object Retrieval in Large Scale Image Databases*. CVPR 2008.
- 1,000 and 50,000 entry histograms for SIFTs found at MSER and Hessian Affine interest points, with 5-way soft assignment.
- No spatial verification, no inverted file.

# New Feature Summary

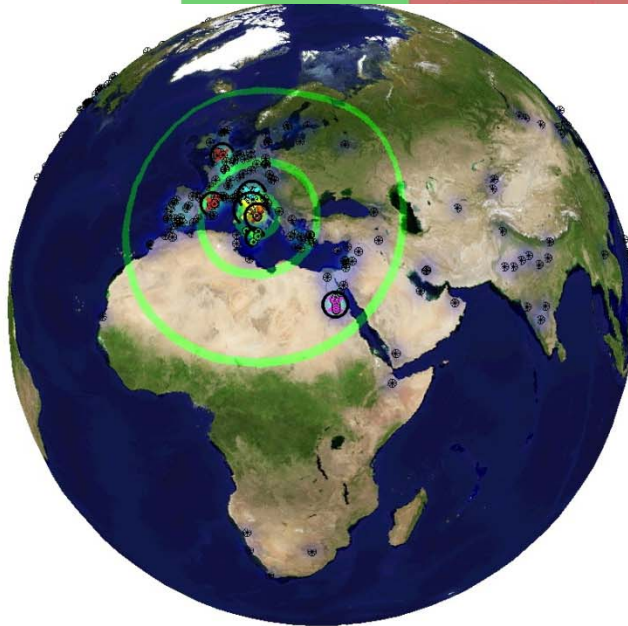
- Gist centered at geometric region of interest.
- Geometry modulated texton and color histograms.
- Bags of SIFT.
- We end up with ~30 elementary features.
- About 8 CPU years to compute.







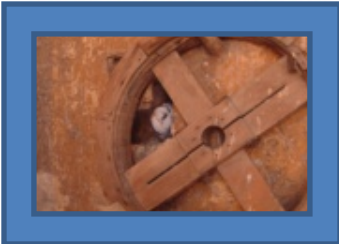
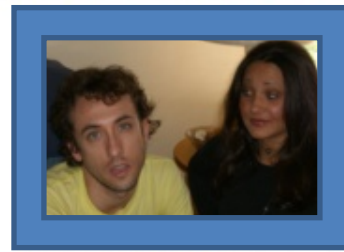
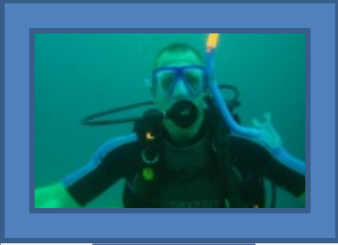
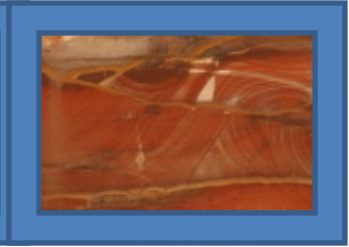
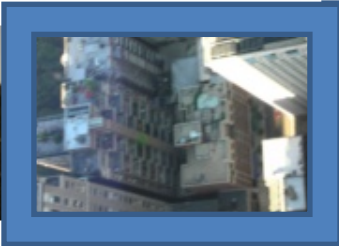
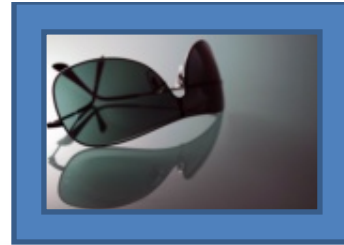




# New Test Set

- 2000 random, geotagged Internet Images.
- Most images are extremely difficult or impossible to geolocate.





# Quantitative Results

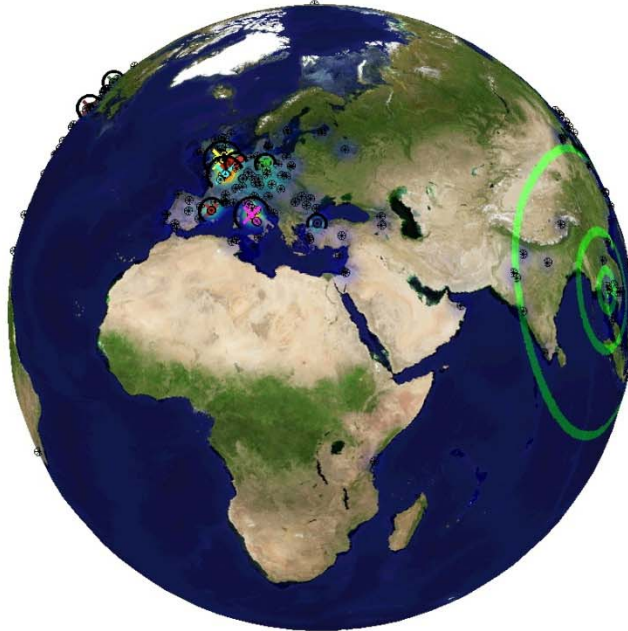
<i>Geolocation Strategy</i>	<i>Accuracy</i>
• Im2gps features, 1nn	6.35%
• Sift features, 1nn	8.05%
• All new features, 1nn	8.85%
• All new features, Largest Cluster	7.00%

# Learning

- Can we use machine learning to improve geolocation accuracy?
- Why would we expect learning to work better than 1 nearest neighbor?



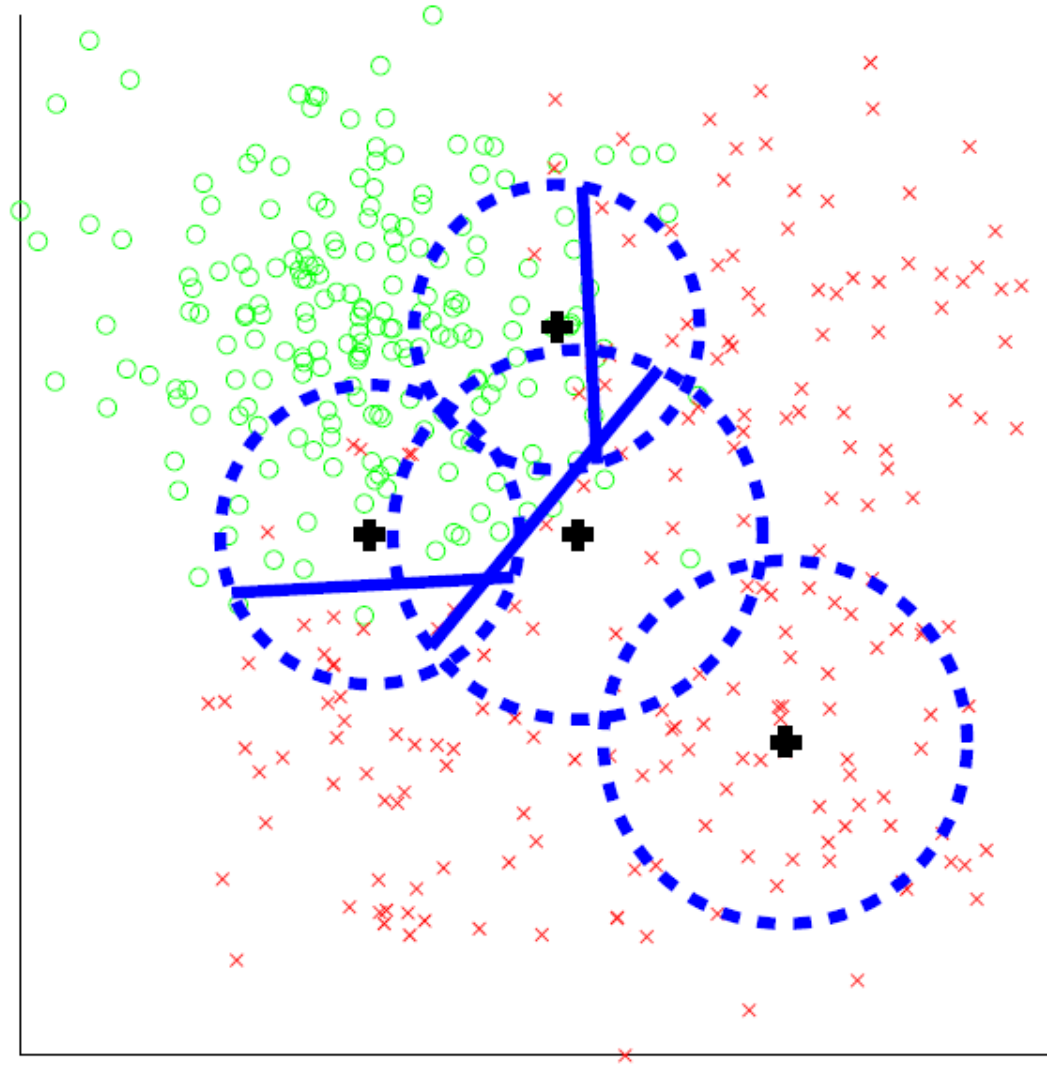




# Learning Issues

- We have 6+ million images, with 100,000 feature dimensions, so naïve application of many learning algorithms is intractable.
- What are our classes? Isn't geolocation more naturally a regression problem?

# “Lazy Learning” - SVM-KNN

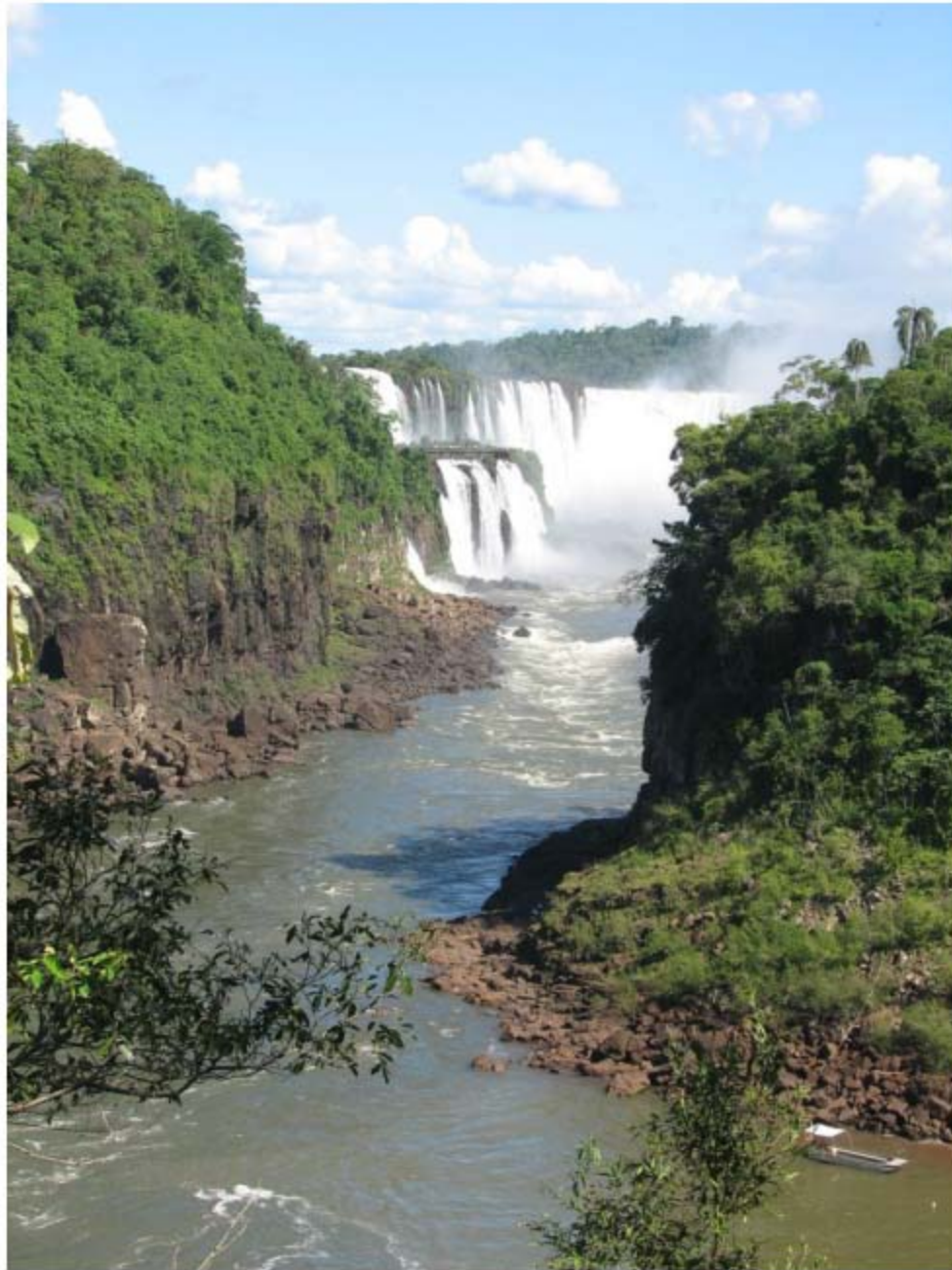


Zhang, Berg, Maire, Malik, CVPR 2006

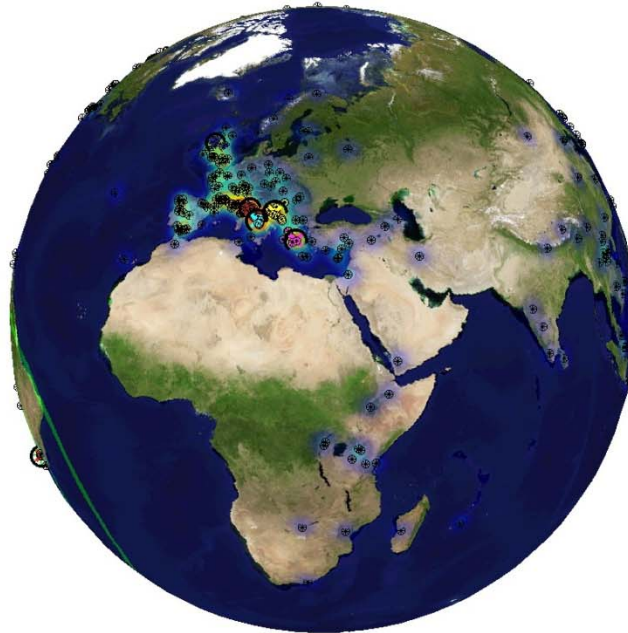
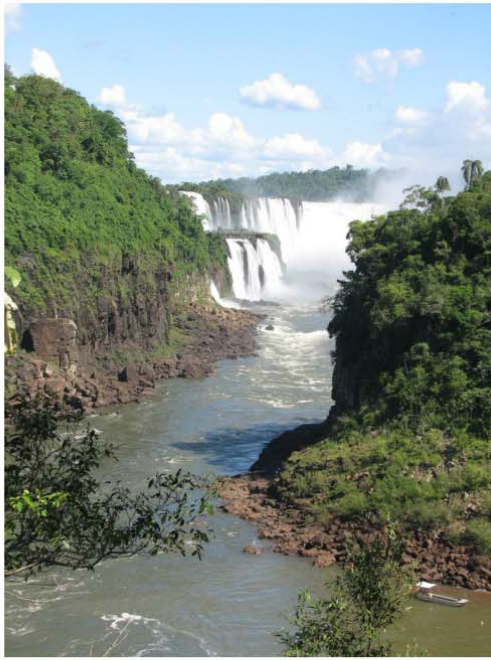
# Lazy Learning

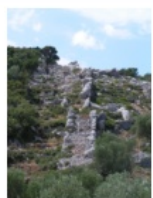
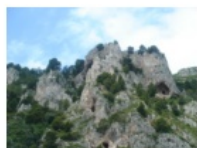
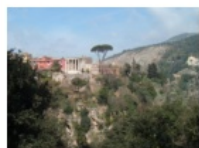
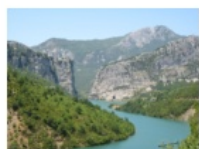
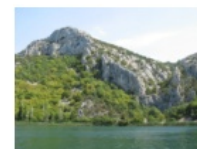
- The mean shift clusters will be our classes (e.g. London class, Seattle class, Kenya class).
- This leaves us an N-way classification problem. For each cluster, we find the non-linear decision boundary which maximally separates those cluster images from all other nearest neighbors.
- We use a kernelized version of our chi squared and L1 distances among the top 400 nearest neighbors



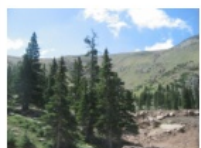




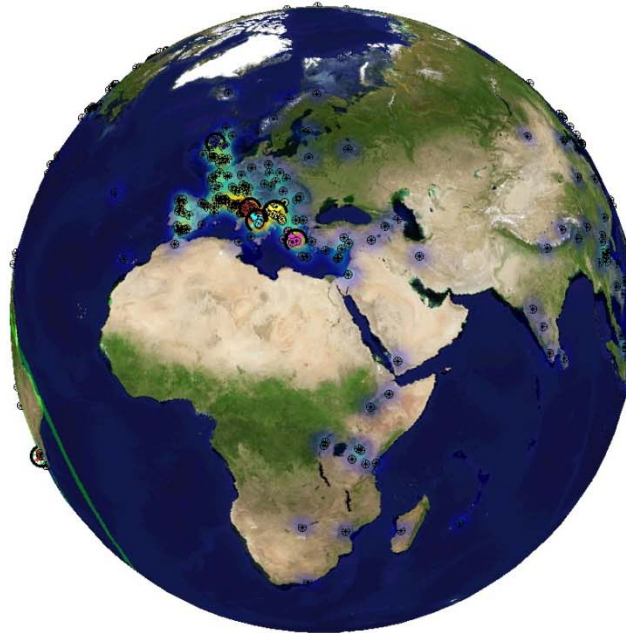
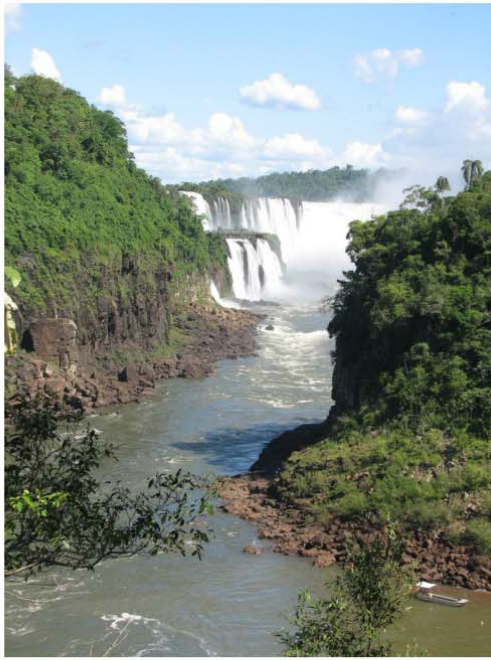










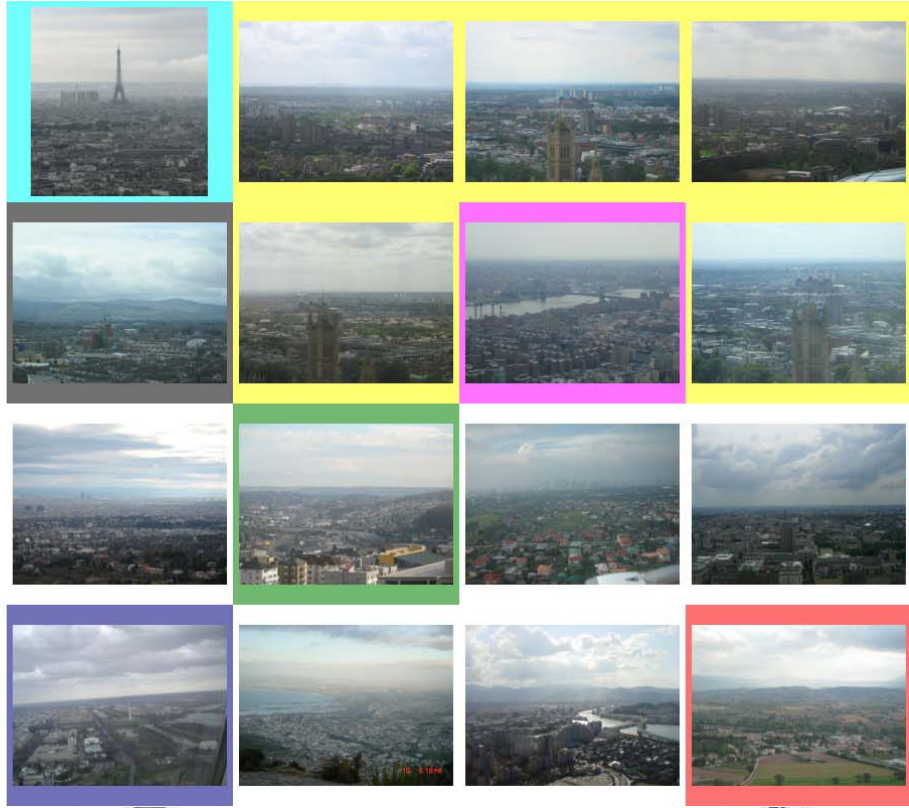


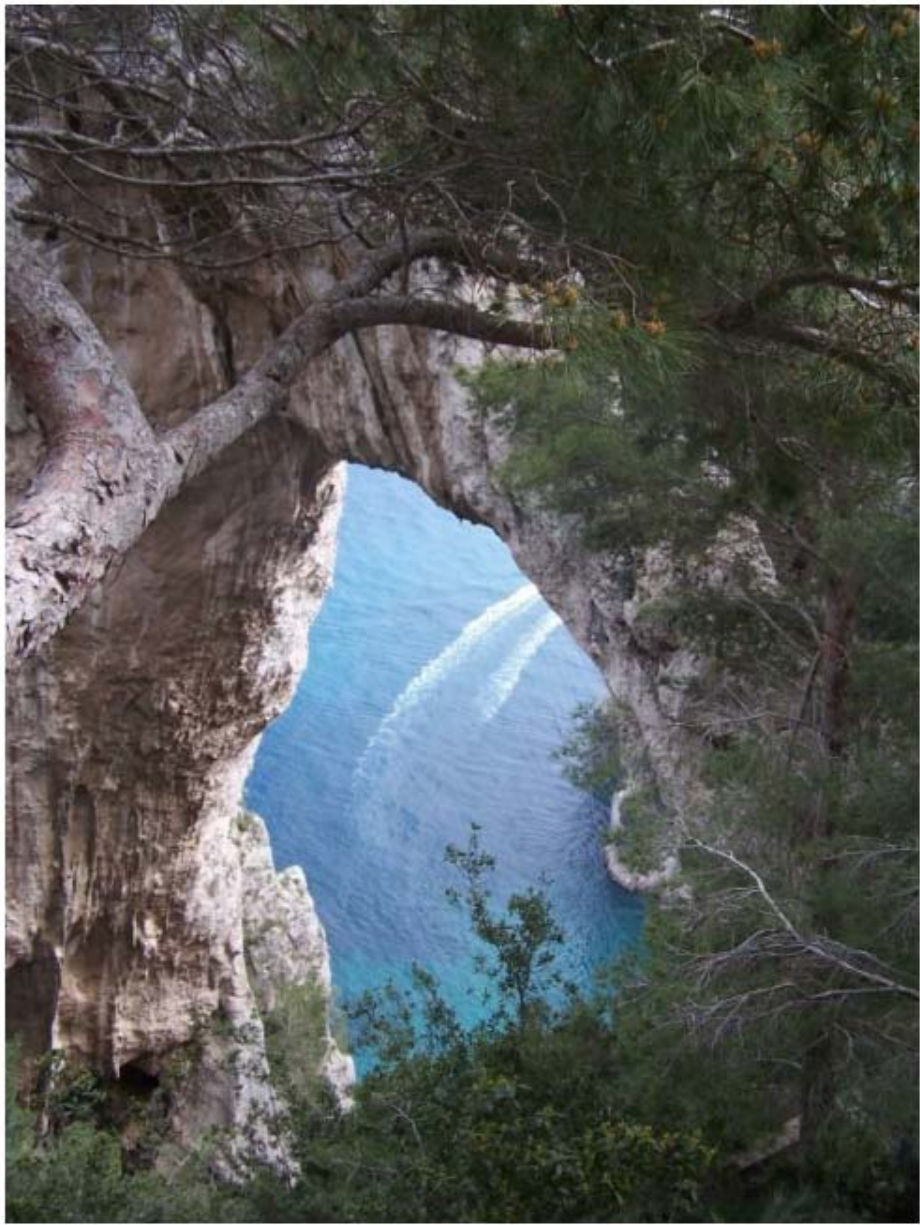
# Lazy Learning

- Avoids computational expense of all-pairs distance computation for SVM training.
- Relies on the nearest neighbor search to retrieve reasonable candidates.













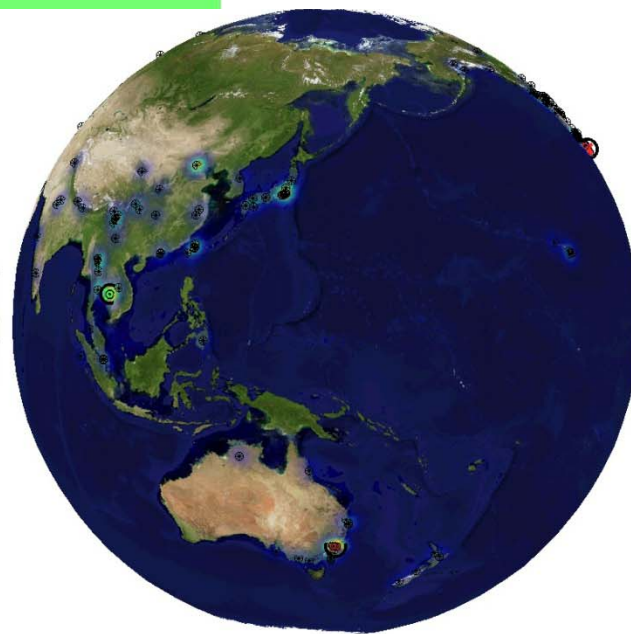
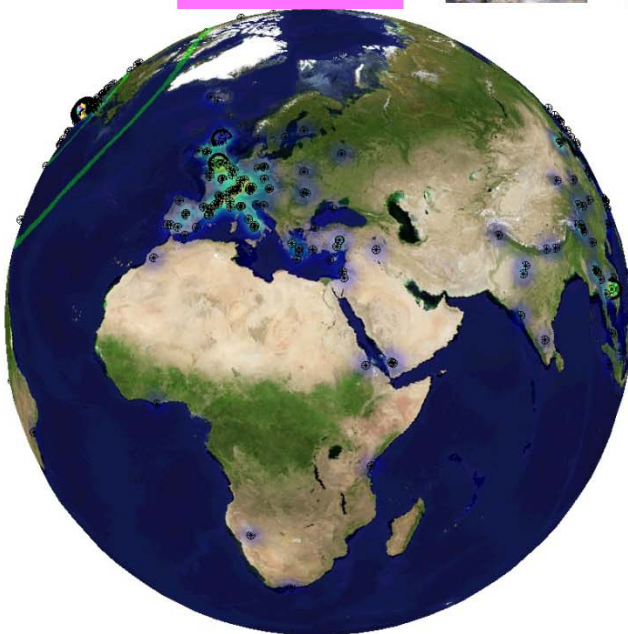




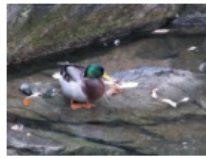
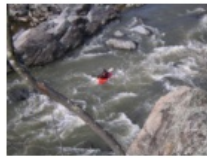












# Quantitative Results

<i>Geolocation Strategy</i>	<i>Accuracy</i>
• Im2gps features, 1nn	6.35%
• Sift features, 1nn	8.05%
• All new features, 1nn	8.85%
• All new features, Largest Cluster	7.00%
• All new features, Lazy SVM	12.50%

# Performance at Different Scales

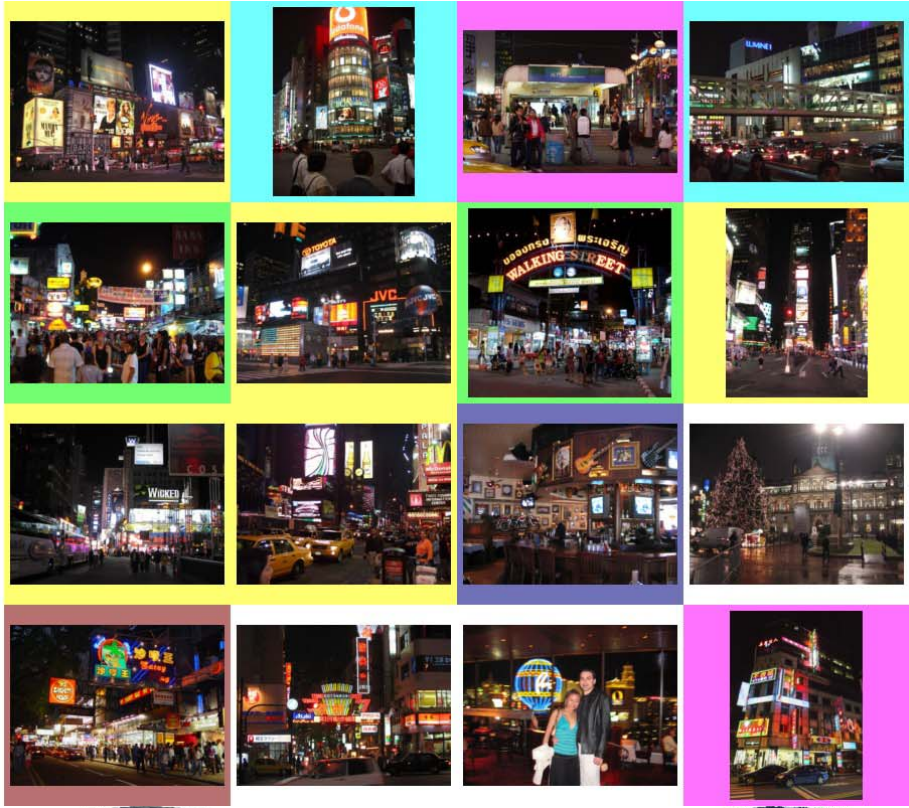
<i>Distance Threshold</i>	<i>Accuracy</i>
• 50km	9.15%
• 200km	12.50%
• 400km	15.95%
• 1200km	27.20%
• 3000km	38.05%



# Remaining issues

- I've defined an overcomplete feature set. How many of these features are really necessary?
- Better parameter fitting on validation set
  - Min cluster size
  - Number of nearest neighbors
  - Mean shift bandwidth
  - What type of kernel should I use
- Geolocation is very rarely correct in sparsely photographed areas.

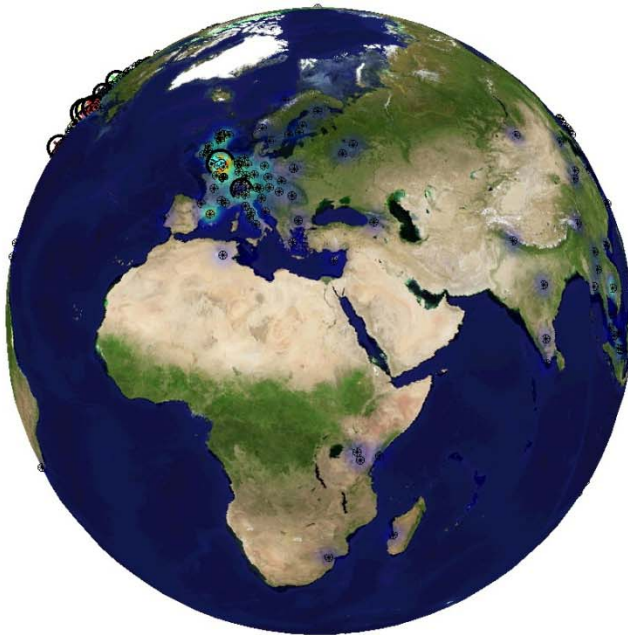






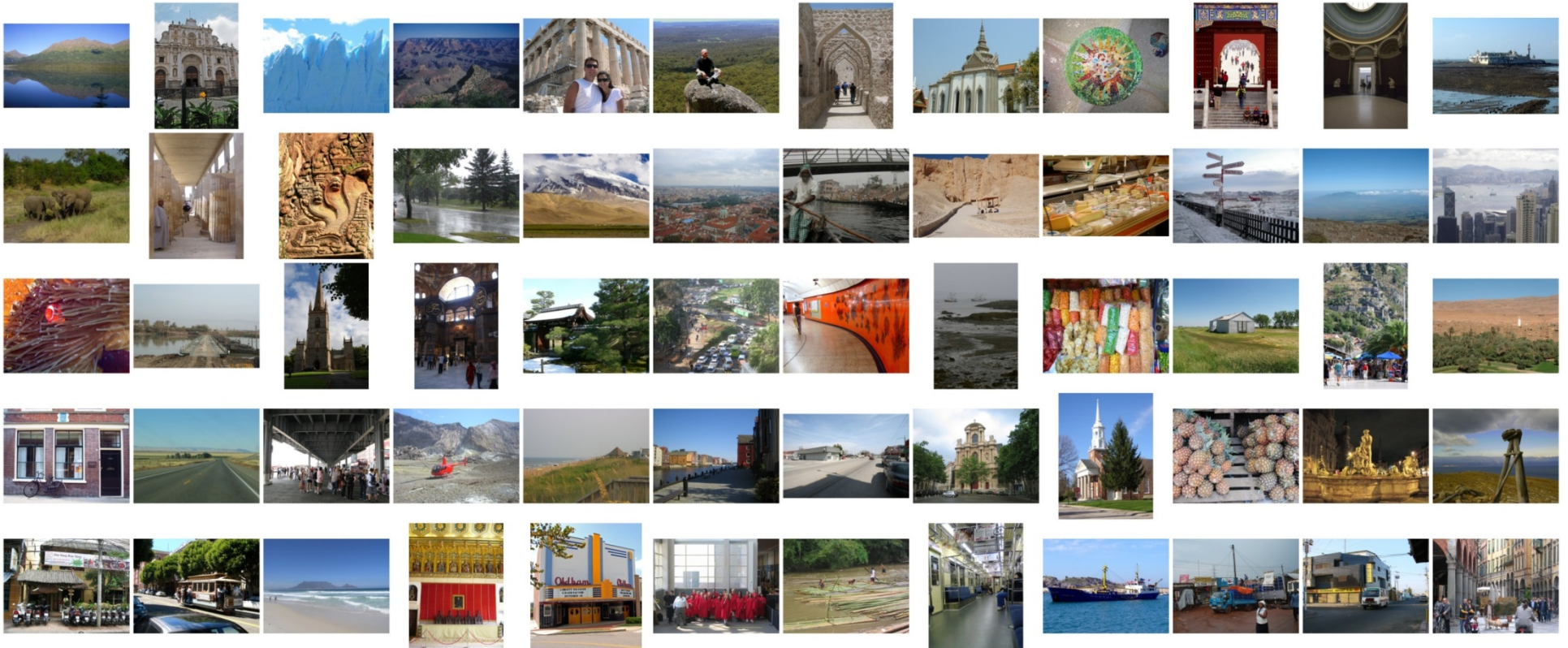








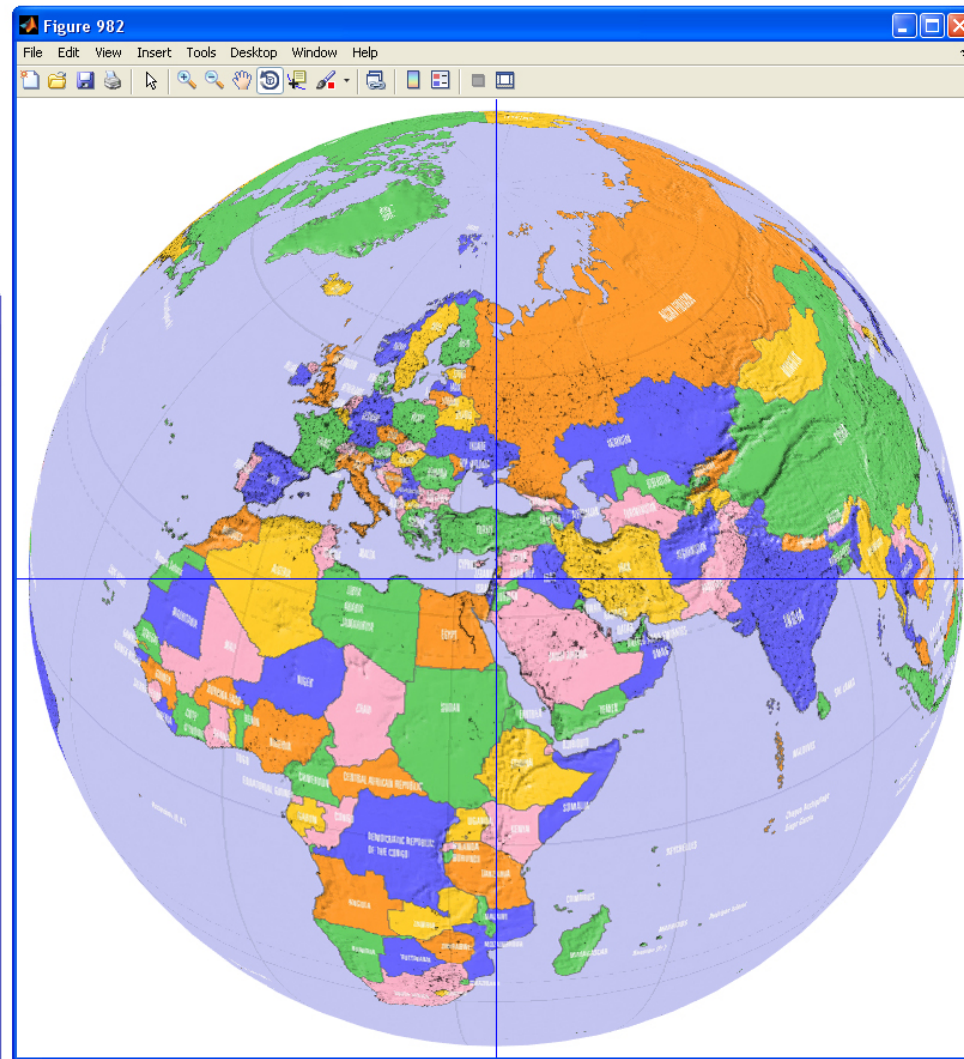
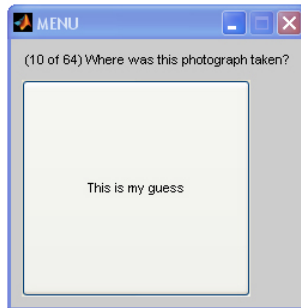
# Human Geolocation of Images



Test set: 64 images pruned from a random  
sample

Hays and Efros. Where in the World? Human and Computer Geolocation of Images. VSS 2009.

20 participants were shown 64 photos and asked to guess the location where each photo was taken.



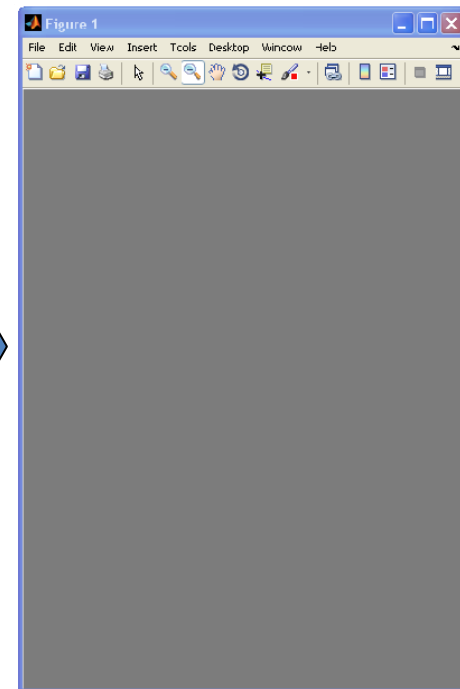
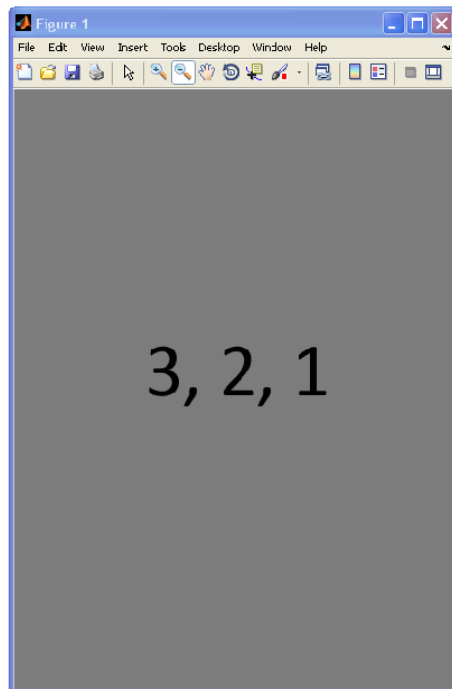


The first set of 32 images allow unlimited viewing. The final 32 images are flashed for only 100 milliseconds.

Countdown

100ms

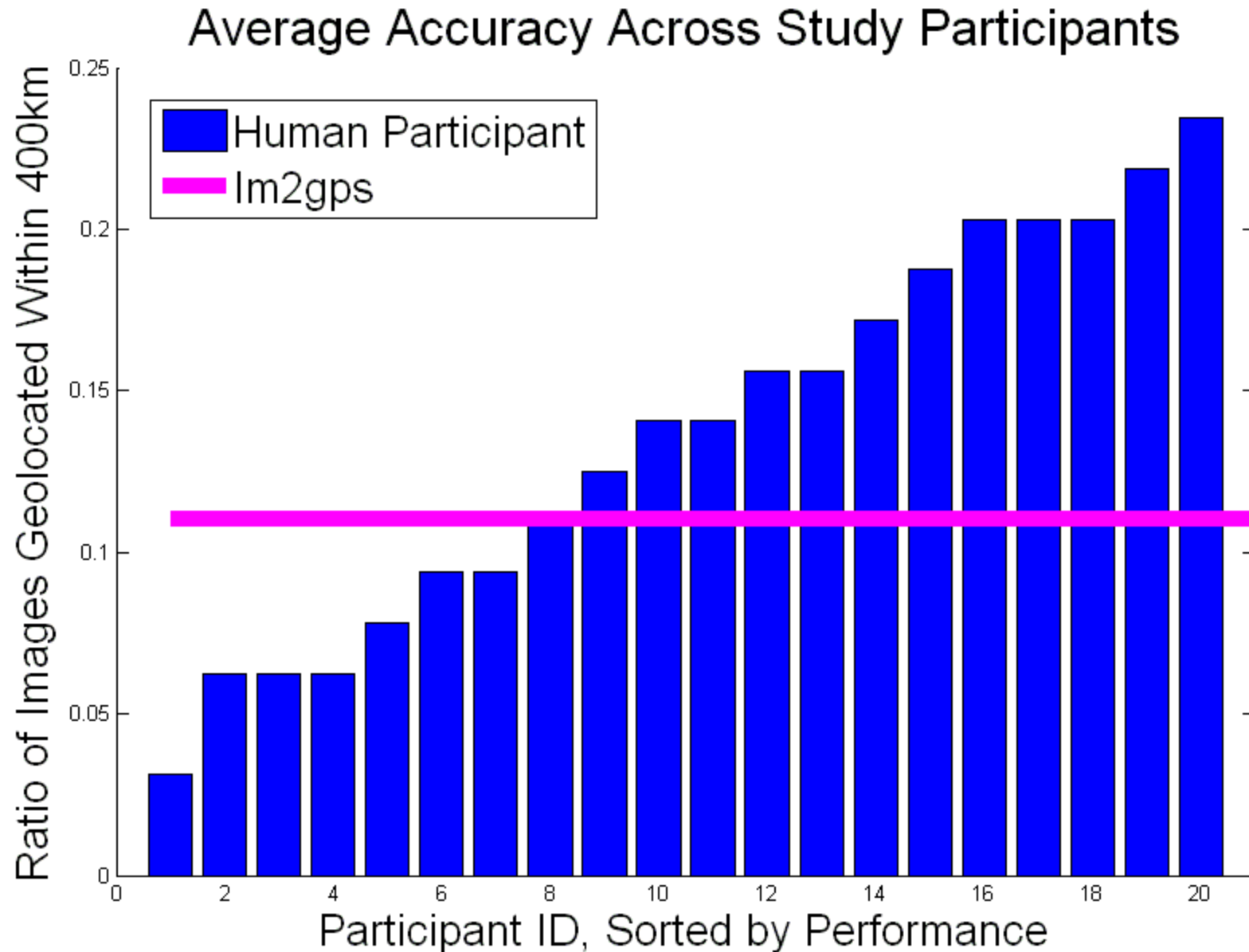
Blank



The image sets alternate viewing conditions. No learning or fatigue effects were observed.

New Features  
and Learning

31%





# Estimating Geographic Properties



Argentina



Andorra



Andorra



Iceland



Idaho



Switzerland



Argentina



Bolivia



Nevada



Hawaii



Hawaii



Egypt



China



Arizona



Peru



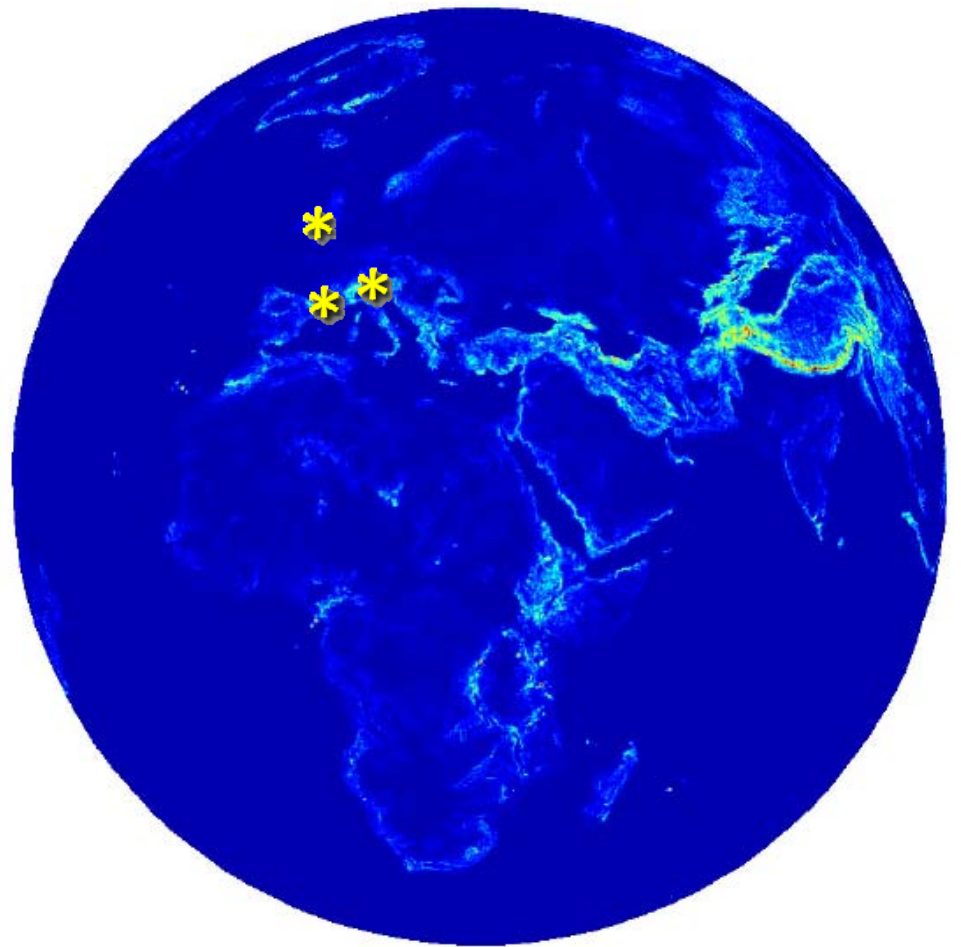
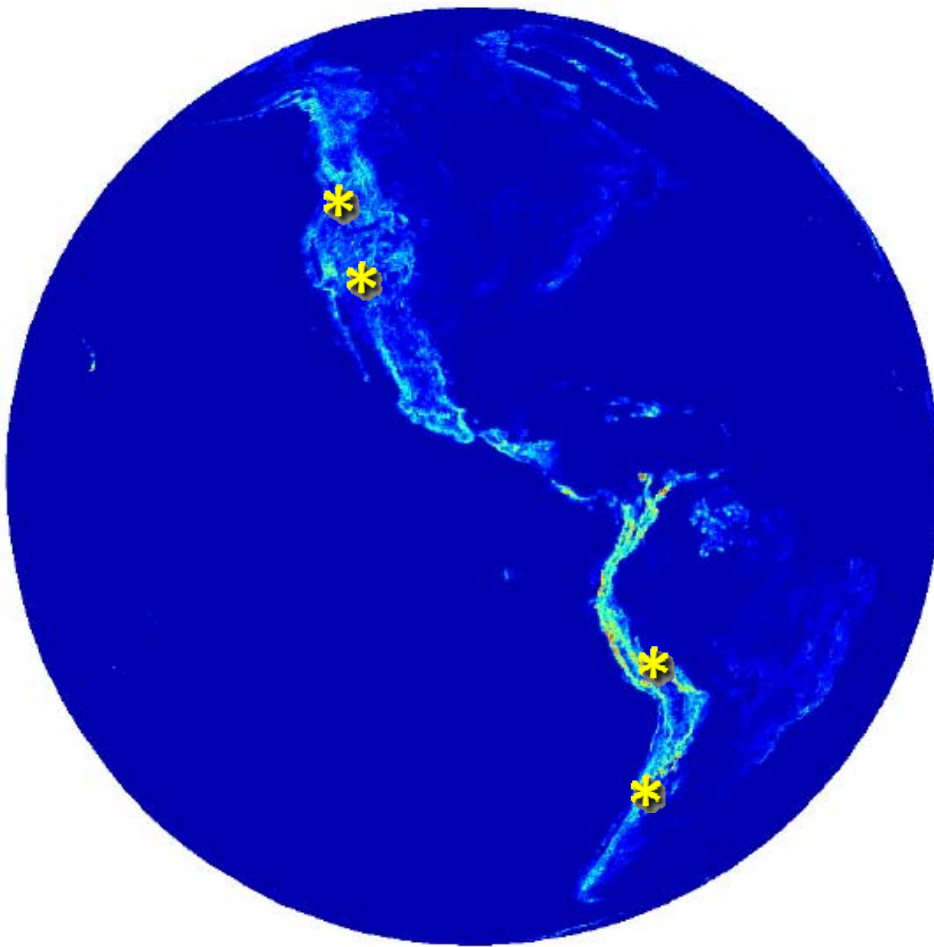
Oregon





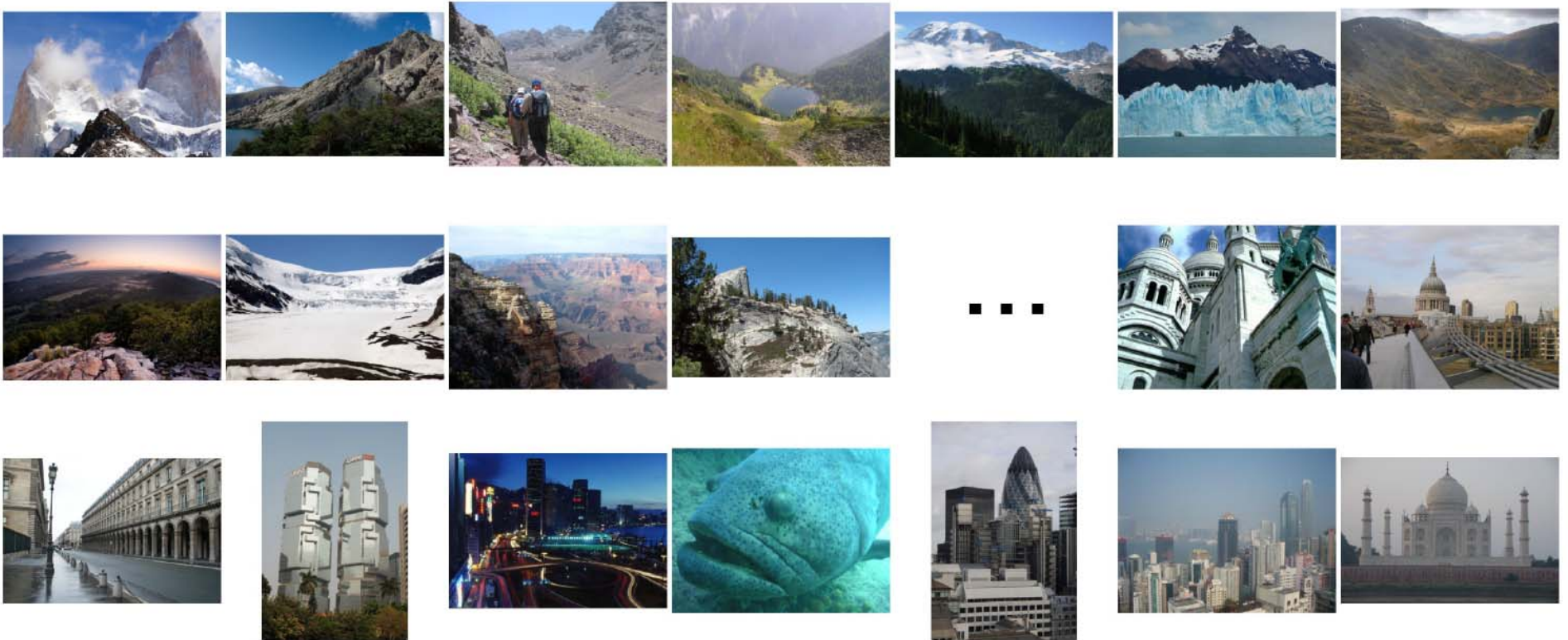


112 m / km





# Elevation gradient magnitude ranking



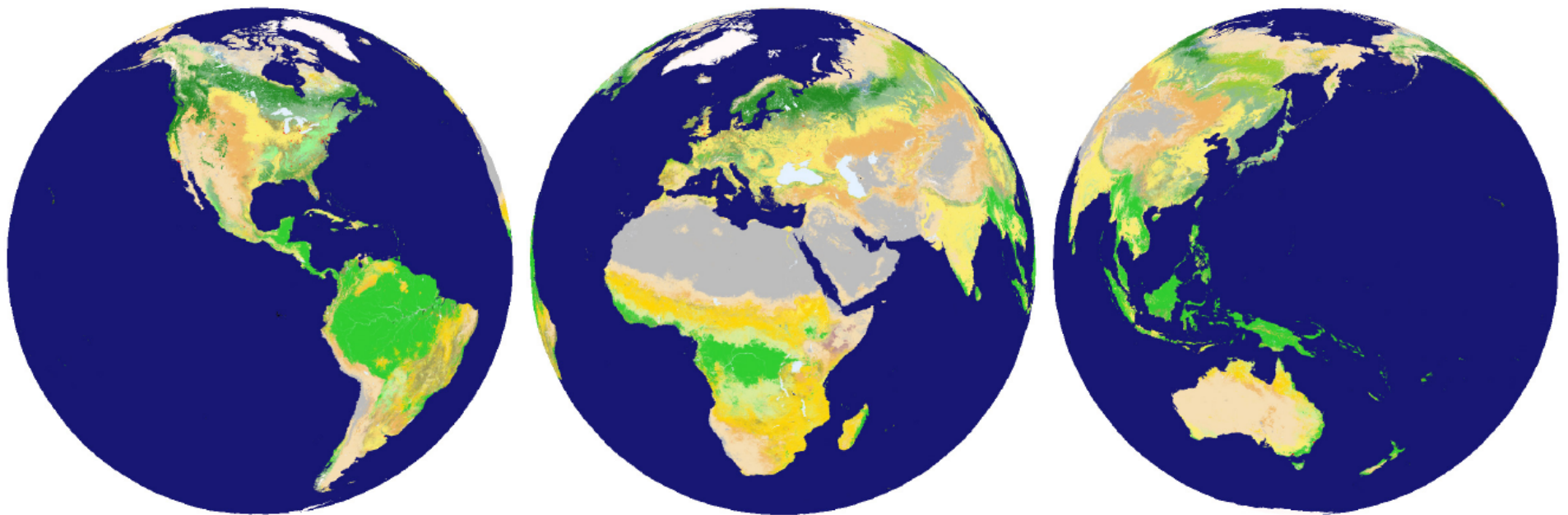


Figure 4. Global land cover classification map.



# Forest



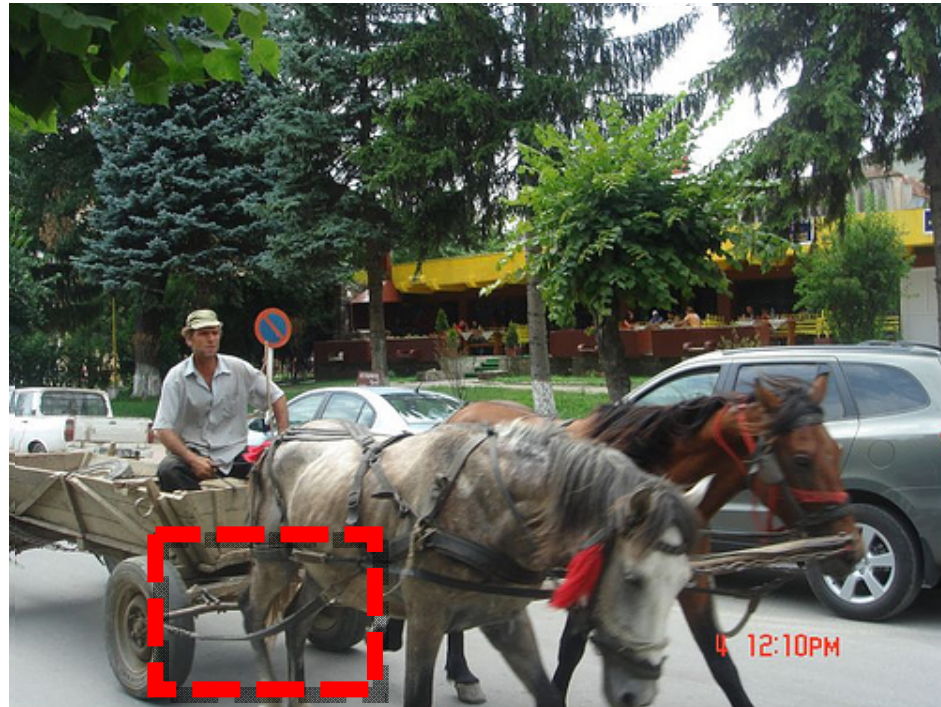


# Geographic and Keyword Context for Object Detection

- Divvala, Hoiem, Hays, Efros, Herbert. *An Empirical Study of Context in Object Detection*. CVPR 2009. **Poster Session 3. Tuesday, June 23**

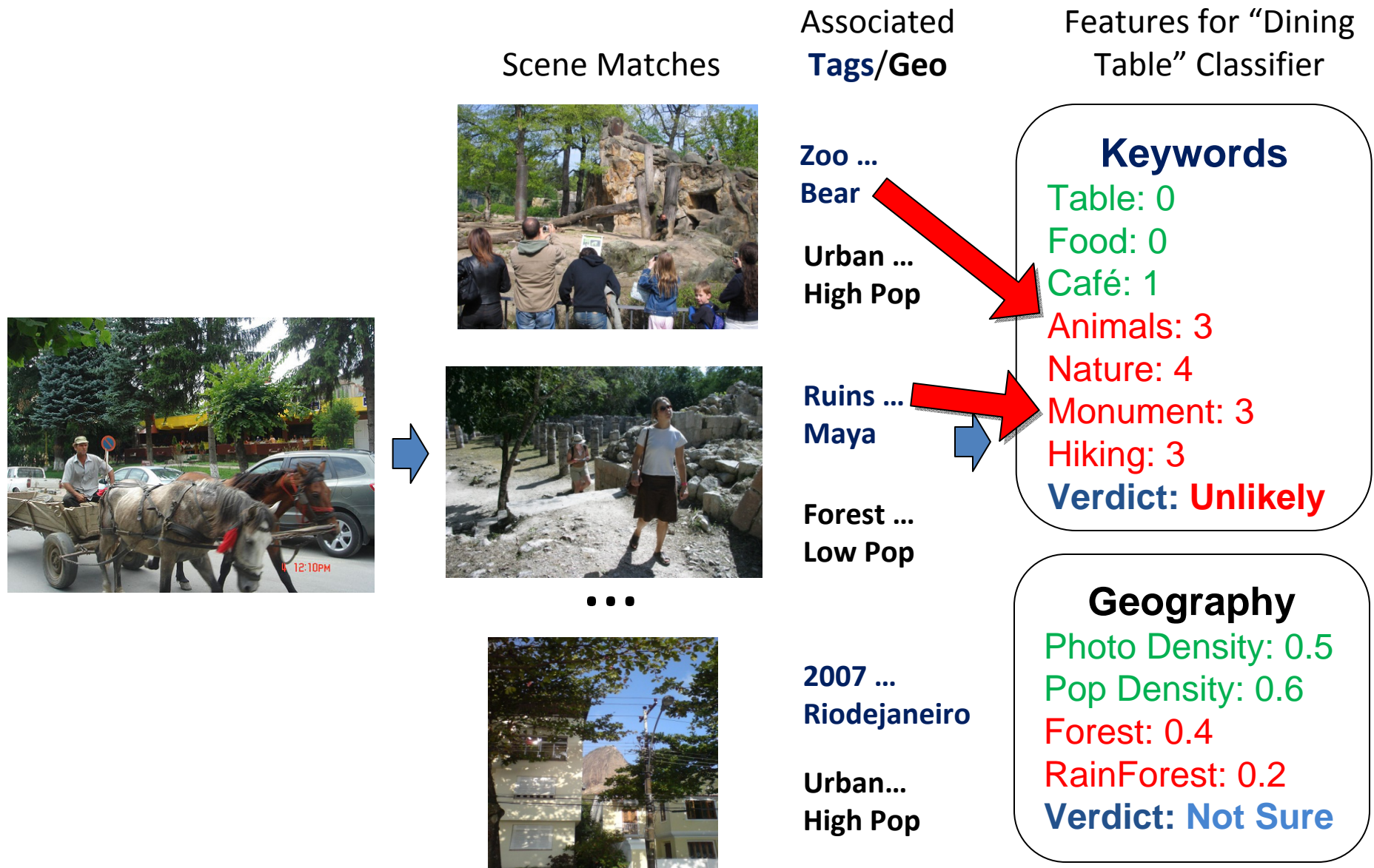


# Goal: Better Detection using Scene Context



Dining Table

# Context from Scene Match Properties



# Geographic context for object detection

- With the entire system (which I haven't discussed) we raise average precision from .182 to .194
- VOC 2008: Best performance in 6 of 20 categories, second best in 6 categories.



# Geographic context for object detection

- Geography properties alone get an average precision of  $\sim .15$  on Pascal VOC 2008 Object Classification Challenge.
- Keyword histograms alone get an average precision of  $\sim .26$ .
- Divvala, Hoiem, Hays, Efros, Herbert. *An Empirical Study of Context in Object Detection*. CVPR 2009. **Poster Session 3. Tuesday, June 23**

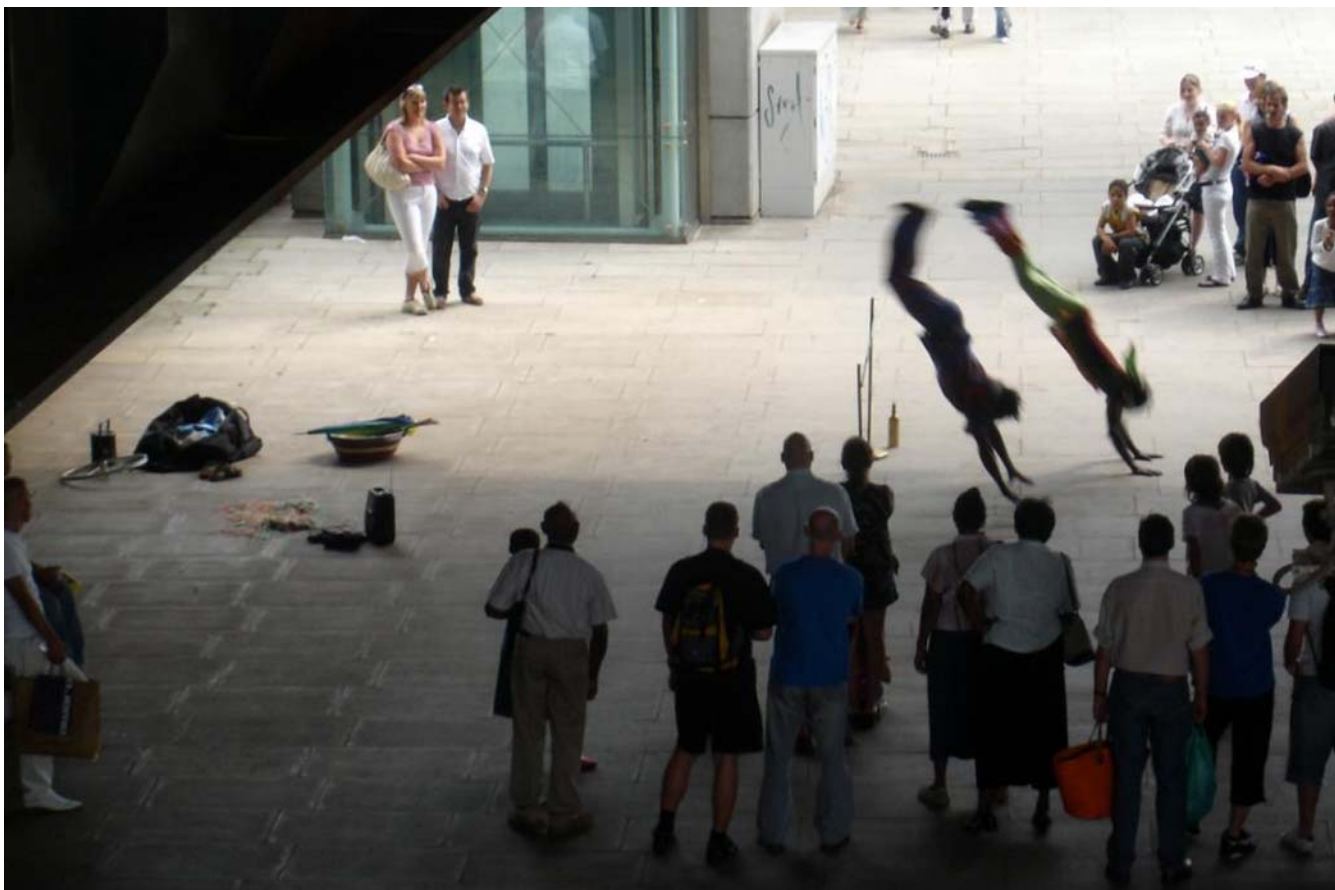
# Image Sequence Geolocation

Kalogerakis, Vesselova, Hays, Efros, Hertzmann.

*Image Sequence Geolocation with Human Travel Priors.*

ICCV 2009

# Where is This?



# Where is This?





# Where are These?



15:14,  
June 18<sup>th</sup>, 2006



16:31,  
June 18<sup>th</sup>, 2006

# Where are These?



15:14,  
June 18<sup>th</sup>, 2006



16:31,  
June 18<sup>th</sup>, 2006



17:24,  
June 19<sup>th</sup>, 2006

# Image Sequence Geolocation with Human Travel Priors



From a single-image accuracy of 10% we reach 58% by letting geolocation estimates reinforce each other *temporally*.