### Visipedia Tool Ecosystem for Dataset Curation and Annotation

Serge Belongie







## Outline

- Visipedia Project Overview
- Related Work
- Bird Datasets
- ViBE: Visipedia Back End
- Future Work

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# What Is Visipedia?

- A user-generated encyclopedia of visual knowledge
- An effort to associate articles with large quantities of well-organized, intuitive visual concepts



http://en.wikipedia.org/wiki/Bird

# Motivation

- People will willingly label or organize certain images if:
  - They are interested in a particular subject matter
  - They have the appropriate expertise



Ring-tailed lemur



Thruxton Jackaroo

### A DUBIOUSLY ACCURATE 233 YEAR HISTORY OF CYCLING

posted by Saris - September 4, 2013 - 5pm EDT



While we're not particularly certain about some of the claims in this bicycle family tree (e.g. freeride bikes spawned downhill bikes, which gave birth to 29ers?), we're certain you'll appreciate the artwork. You can head to their website to buy your own copy and laugh at the implication that big wheels evolved into recumbents for only \$22.

[BikeRumor.com]

#### COMMENTS

1000

#### Ben - 09/04/13 - 5:44pm

This is so completely out of order. Why the hell would you pay 22 dollars for a poster that doesn't make any sense?

#### Gillis - 09/04/13 - 6:00pm

I like how the track bike sits in between the randonneur and touring bikes. And a modern looking TT bike some how comes before Boardman's Lotus, which both come after fixie's.

This is junk.

#### Walter - 09/04/13 - 6:09pm

So triathlon bikes gave birth to fixies and early eighties long wheel base recumbents came from modern high racers. These folks are creationists.

#### NotAMachinist - 09/04/13 - 6:20pm

It's sort of cool looking until you really look at it. For instance how does a randonneur differ from a touring or trekking bike? How did cyclocross spawn BMX? The cycling family tree is far more incestuous.

#### Joe - 09/04/13 - 6:22pm

Cyclocross to 20" Dirt, Street, Park to Racing to Freestyle and Flatland????? Not a one of these is right....the whole chart is a nice piece of wallpaper art but that's it.

8888888888

Keith D - 09/04/13 - 6:46pm It's pretty much rubbish.



# Motivation

- Construct comprehensive, intuitive knowledge base of visual objects
- Provide better text-to-image search and image-to-article search









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# **Related Work: Systems**

- {Leaf,Dog,Bird}snap [Belhumeur et al.]
- Oxford Flowers [Nilsback & Zisserman]
- STONEFLY9 [Martínez-Muñoz et al.]
- omoby [IQEngines.com]
- 20 Questions game [20q.net]
- ReCAPTCHA [von Ahn et al.]
- Wikimedia Commons







# **Related Work: Methods**

- Relevance Feedback
- Active Learning
- Expert Systems
- Decision Trees
- Feature Sharing & Taxonomies
- Parts & Attributes
- Crowdsourcing & Hum $an^{x_2 \leq \theta_2}$ Computation



 $x_1 > \theta_1$ 

 $x_1 \leqslant \theta_4$ 

R

# Motivation: Computer Vision Perspective

### • Need for more training data

- Beyond the capacity of any one research group
- Better quality control
- Need for more realistic data
  - Let people define what tasks are important
  - Study tightly-related categories

# Dealing With a Large Number of Related Classes

- Standard classification methods fail because:
  - Few training examples per class available
  - Variation between classes is small
  - Variation within a class is often still high



Brewer's Sparrow



Vesper Sparrow

### (A) Easy for Humans





Chair? Airplane? ...

### (B) Hard for Humans





Finch? Bunting?...

### (C) Easy for Humans





Yellow Belly? Blue Belly? ...

### **Plants vs Birds**



2d

3d

Doesn't move

Okay to pluck from tree

Mostly single color

Very few parts

Adequately described by boundary

Relatively easy to segment



Not okay to pluck from tree

Many colors

Many parts

Not well described by boundary

Hard to segment

slide credit: Neeraj Kumar

# Visual 20 Questions



- "Computer Vision" module = Vedaldi's *VLFeat*
- VQ Geometric Blur, color/gray SIFT spatial pyramid
- Multiple Kernel Learning
- Per-Class 1-vs-All SVM
- 15 training examples per bird species
- Choose question to maximize expected Information Gain

# Pose Normalized Deep ConvNets



[Van Horn, Branson, Perona, Belongie BMVC 2014]

### Algorithm 1 Visual 20 Questions Game

- $1: \ U^0 \leftarrow \emptyset$
- 2: for t = 1 to 20 do
- 3:  $j(t) = \max_k I(c; u_k | x, U^{t-1})$
- 4: Ask user question  $u_{j(t)}$ , and  $U^t \leftarrow U^{t-1} \cup u_{j(t)}$ .
- 5: end for
- 6: Return class  $c^* = \max_c p(c|x, U^t)$

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## **Birds-200 Dataset**



6033 images over 200 bird species

# Image Harvesting

- Flickr: text search on species name
- MTurk: presence/absence and bounding boxes



\*



# The human annotation process

- Modeling various aspects of annotation:
  - Worker competency accuracy in labeling
  - Worker expertise better at labeling some things than others, based on their strengths
  - Worker bias how one weighs errors
  - Task difficulty ambiguous images are universally hard to label
  - True label the ground truth label value
- We leverage the "Multidimensional Wisdom of Crowds" [Welinder et al. 2010]

# Types of annotator errors

### Task: Find the Indigo Bunting

Indigo Bunting



Blue Grosbeak



















### Image formation process



### **Entire annotation process**



### Image difficulty











[Welinder et al., 2010]



[Welinder et al., 2010]






## Worker "schools of thought"





## Discussion: quality management

- Models can capture multidimensionality of annotation process
- How well does this generalize to continuous annotations?

Different tasks require different reviewing strategies.

Predicting quality accurately can reduce the number of labels needed.



Please provide feedback in the box below if you have comments:

# **Attribute Labeling**

- Attributes from whatbird.com
- 25 visual attributes ~288 binary attributes
   o similar to "dichotomous key" in biology
- MTurk interface
  - o {guessing, probably, definitely}
- 3-5x redundancy factor



\*











# **MTurker Label Certainty**



# MTurker Feedback

- "These hits were fun. Will you be posting more of them anytime soon? Thanks!"
- "These are Beautiful birds and I am enjoying this hit collection"
- "I really enjoy doing your hits, they are fun and interesting. Thanks."
- "Love doing these because I'm a bird watcher."
- "the birds are so cute..hope u can send more kind of birds"
- "I haven't really studied birds, but doing these HITs has made me realize just how beautiful they are. It has also made me aware of the many different types of birds. Thank you"
- "I REALLY LOVE THE COLOR OF THE BIRDS."
- "Thank you for providing this job. The fact that the images are beautiful to look at make it a lot more enjoyable to do!"

Try out on a new dataset for fine-grained recognition, featuring 550 of North America's most common birds. The full dataset will be available in the fall. Join the competition today and download the "taster" dataset!

### http://birds.cornell.edu/nabirds

CCUB NABirds includes:

- More than 700 visual categories, . organized taxonomically
- Photos curated in collaboration with domain experts
- Data organized in a researcher-friendly, widely-used PASCAL VOC format

For more information contact: Ryan Farrell (farrell@eecs.berkeley.edu)

![](_page_49_Picture_7.jpeg)

![](_page_49_Picture_8.jpeg)

![](_page_49_Picture_9.jpeg)

**CCUB**NABirds

700

## **CCUB** Taster25

![](_page_50_Picture_1.jpeg)

"Bitter" Taster

## **CCUB Taster25 Results**

![](_page_51_Figure_1.jpeg)

### **Baseline Performance:**

The winning ILSVRC '11 approach of Florent Perronnin and Jorge Sanchez.

- Dense SIFT and Color Descriptors
- Aggregated using Fisher vectors [Perronnin, et al. ECCV 10]
- Linear SVMs with SGD
- Same parameters used in ILSVRC

Using the winning ILSVRC '11 approach by [F. Perronnin, et al.], training on 25 images/category

## **CCUB Taster25 Results**

![](_page_52_Figure_1.jpeg)

Average Performance: 79.4%

Using the winning ILSVRC '11 approach by [F. Perronnin, et al.], training on 50 images/category Try out on a new dataset for fine-grained recognition, featuring 550 of North America's most common birds. The full dataset will be available in the fall. Join the competition today and download the "taster" dataset!

### http://birds.cornell.edu/nabirds

# http://birds.cornell.edu/nabirds

For more information contact: Ryan Farrell (farrell@eecs.berkeley.edu)

![](_page_53_Picture_4.jpeg)

![](_page_53_Picture_5.jpeg)

![](_page_53_Picture_6.jpeg)

700

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# **Vibe Demo**

http://visipedia.org http://vibe.visipedia.org

## **Visipedia Backend**

Storage and collaboration infrastructure to support visual search applications.

#### Storage

#### Cloud storage and access for your image datasets and annotations.

#### Collaborate

Divide and conquer your data collection and curation tasks by sharing your data with collaborators.

#### Deploy

Integrate Vibe storage functionality into your app or website for easy image upload and annotation by your users.

#### Organize

Build a hierarchical representation of your domain and use it to organize your images.

#### Annotate

Use our annotation templates to create your custom annotation tasks.

#### Analyze

Hook Vibe into your classification pipeline to analyze how images are being classified.

Vibe Home Buckets Snapshots	Docs				gvanhorn <del>-</del>	
Q A Q ✓NAB 700 Taxonomy i	NAB 700 Taxonomy	100722 total Add Content	Go To  Order By Clear Selection Selection Ope	Size - Si	ant Content	
<ul> <li>&gt; Ducks, Geese, and Swans 1</li> <li>&gt; Grouse, Quail, and Allies 1</li> <li>&gt; Loons 1</li> <li>&gt; Grebes 1</li> <li>&gt; Flamingos 1</li> <li>&gt; Shearwaters, Petrels, Albatross, and Allies 1</li> </ul>						
<ul> <li>Storks :</li> <li>Frigatebirds, Boobies, Cormorants, Darters, a</li> <li>Pelicans, Herons, Ibises, and Allies :</li> <li>Hawks, Kites, Eagles, and Allies :</li> <li>Caracaras and Falcons :</li> </ul>						
<ul> <li>Cranes and Rails :</li> <li>Plovers, Sandpipers, and Allies :</li> <li>Skuas and Alcids :</li> <li>Gulls, Terns, and Allies :</li> <li>Pigeons and Doves :</li> </ul>			~*	3		
<ul> <li>Parrots :</li> <li>Cuckoos :</li> <li>Owls :</li> <li>Nightjars :</li> <li>Swifts and Hummingbirds :</li> <li>Trogons and Quetzals :</li> </ul>		*				) 4 4
<ul> <li>&gt; Kingfishers and Allies :</li> <li>&gt; Woodpeckers :</li> <li>&gt; Perching Birds :</li> </ul>			Click on the The American Click on the Ameri	tt to a bucket to k drag and drop	74	

Vibe Home Buckets Snapshots Docs									gva	nhorn <del>-</del>
Close Tree View Lord S	tistics		Total N	odes: 167	4					
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Fulvous Whistling	Duck 44									
Greater White-fronted	Boose		162							
Snow Goose (White	iorph)		177							
Snow Goose (Blue	iorph)	118								
Ross's	Boose		155							
	Brant	14	49							
Cackling	Boose		166							
Canada	Boose	1	153							
Mute	Swan		2	05						
Trumpete	Swan		171							
Tundra	Swan		190							
Muscov	Duck	91								
Wood Duck (Breeding	male)				261					
Wood Duck (Female/Eclipse	male)		186							
Gadwall (Breeding	male)			226						
Gadwall (Female/Eclipse	male)	109								
American Wigeon (Breeding	male)		190							
American Wigeon (Female/Eclipse	male)	109								
Eurasian Wigeon (Breeding	male)	1	154							
Eurasian Wigeon (Female/Eclipse	male) 24									
American Blac	Duck	121								
Mallard (Breeding	male)			242						
Mallard (Female/Eclipse	male)				278					
Mottle	Duck	139								
Blue-winged Teal	Male)		187							
Blue-winged Teal (Female/ju	venile)	121								
Cinnamon Teal	Male)			211						
Cinnamon Teal (Female/ju	venile)	108								
Northern Shoveler (Breeding	male)			250	D					
Northern Shoveler (Female/Eclipse	male)	11	50							
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#### Start Bird Photo ID

Merlin is equipped with computer vision to identify birds in images. Thanks to help from the community who shared and annotated images, Merlin has been trained to identify 125 species. At this stage, Merlin well-suited to handle highquality images taken with DSLR cameras. Results may not be as good for images taken on iPhones and other mobile devices, however, we encourage you to see how Merlin does with lower-quality images! All images uploaded to this test interface will be used help improve the system in the future. We appreciate your feedback as Merlin development continues.

#### Select your photo.

![](_page_60_Figure_2.jpeg)

### Did you take this photo?

![](_page_61_Picture_2.jpeg)

### Crop the bird by clicking and dragging a box.

![](_page_62_Picture_2.jpeg)

![](_page_63_Figure_0.jpeg)

### Click on the bill tip.

![](_page_63_Picture_2.jpeg)

![](_page_63_Picture_3.jpeg)

### Click on the eye. If both eyes are visible, click on the side of the head that is more visible.

![](_page_64_Picture_2.jpeg)

![](_page_64_Picture_3.jpeg)

![](_page_65_Figure_1.jpeg)

#### Where did you see the bird?

### When did you see the bird?

![](_page_66_Picture_2.jpeg)

Creating list of possible birds...

![](_page_67_Picture_2.jpeg)

![](_page_68_Picture_1.jpeg)

Dryden, NY, USA

Aug 6

Black-capped Chickadee

![](_page_68_Picture_5.jpeg)

#### Can't find your bird?

Merlin considers the 400 most familiar species in North America. The bird you're seeing might not be in our database yet. Browse species.

**ID Another Bird** 

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# Future Work

- Beyond Birds
- Attribute Induction
- Relevance Feedback

![](_page_70_Figure_4.jpeg)

![](_page_70_Figure_5.jpeg)

![](_page_71_Picture_0.jpeg)


Click on the birds to the right that are clearly dissimilar species from the bird above.





 $\mathcal{T}^{k} = \{(i, j, l) | x_{i} \text{ is more similar to } x_{j} \text{ than } x_{l} \}$  $\|\mathbf{z}_{i} - \mathbf{z}_{j}\|_{2} < \|\mathbf{z}_{i} - \mathbf{z}_{l}\|_{2} \iff s(i, j) > s(i, l)$ 

## **Perceptual Embedding**





Query Image

Q1: Most Similar By Shape?

# Thank You

- Caltech: Steve Branson, Grant Van Horn, Pietro Perona
- UCSD: Catherine Wah
- Cornell: Jessie Barry, Miyoko Chu
- BYU: Ryan Farrell
- Google Focused Research Award



# visipedia.org

### **Extra Slides**

#### **Computational Pathology**



**Figure 1.4**. In our crowdsourcing framework, workers are asked which pair images in a triplet look most similar. Given a large collection of such relative measurements, we can learn a similarity function on image patches that captures human perception. Left: triplet depicting three regions of interest from a slide image. Middle: zoomed-in regions. Right: the chosen pair (clicked by the worker) is indicated in green.



# **Populating Visipedia**

 Populate Wikipedia articles with more visual data using large quantities of unlabeled data on the web



# **Attribute-Based Classification**

- Train classifiers on attributes instead of objects
- Attributes are shared by different object classes
- Attributes provide the ingredients necessary to recognize each object class

otter	
black:	yes
white:	по
brown:	yes
stripes:	no
water:	yes
eats fish:	yes

polar bearblack:nowhite:yesbrown:nostripes:nowater:yeseats fish:yes

#### <u>zebra</u> black:

black: yes white: yes brown: no stripes: yes water: no eats fish: no



Lampert et al. 2009 Farhadi et al. 2009

# **Attribute-Based Classification**

- Number of attributes is less than number of classes
- Attribute classification tasks might be easier
- Makes it easier to incorporate human knowledge



www.whatbird.com

#### Annotator bias



[Welinder et al., 2010]