



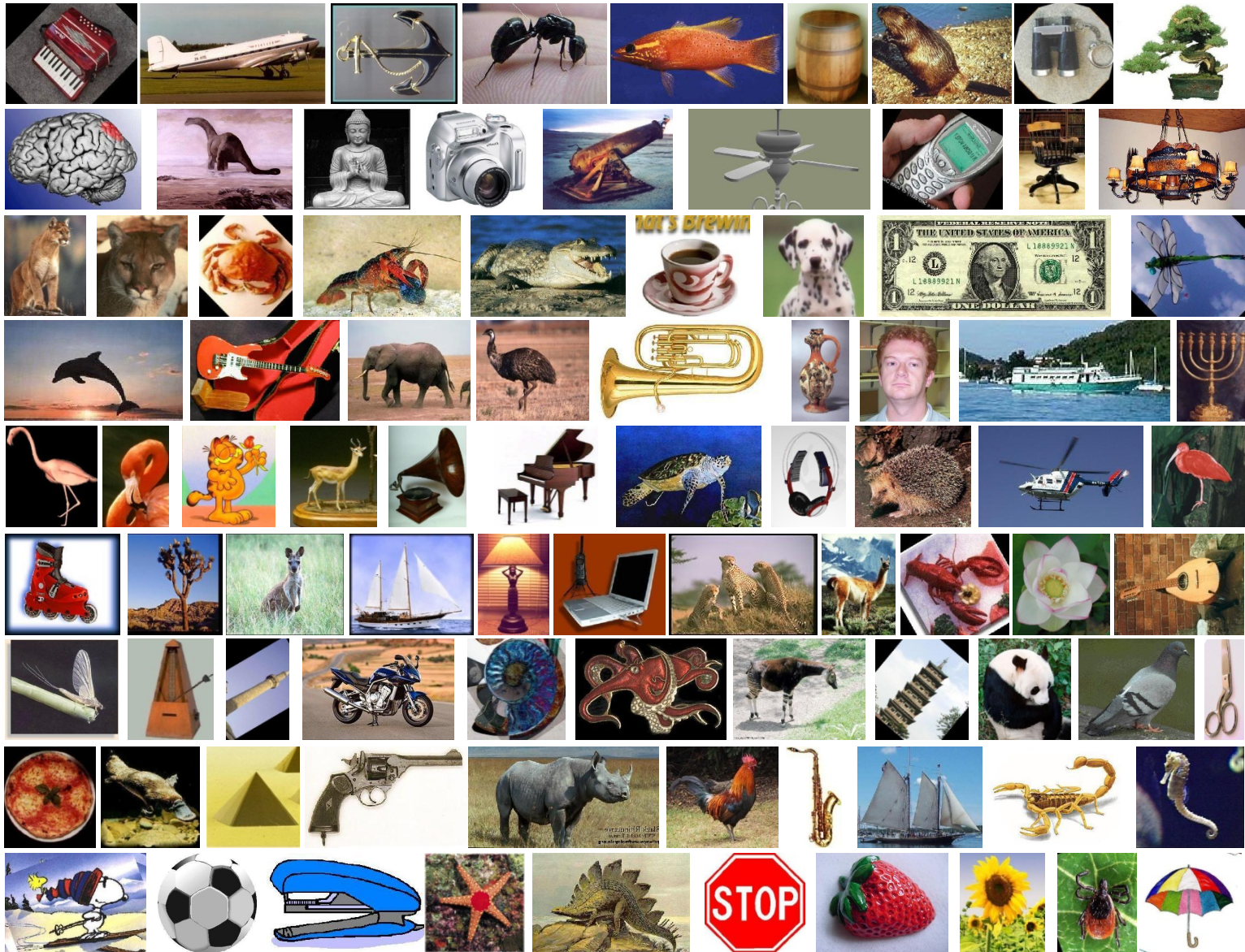
EECS 442 – Computer vision

Databases for object recognition and beyond

Caltech 101

- Pictures of objects belonging to 101 categories.
- About 40 to 800 images per category. Most categories have about 50 images.
- The size of each image is roughly 300 x 200 pixels.

Caltech 101 images



Caltech-101: Drawbacks

- Smallest category size is 31 images: $N_{train} \leq 30$



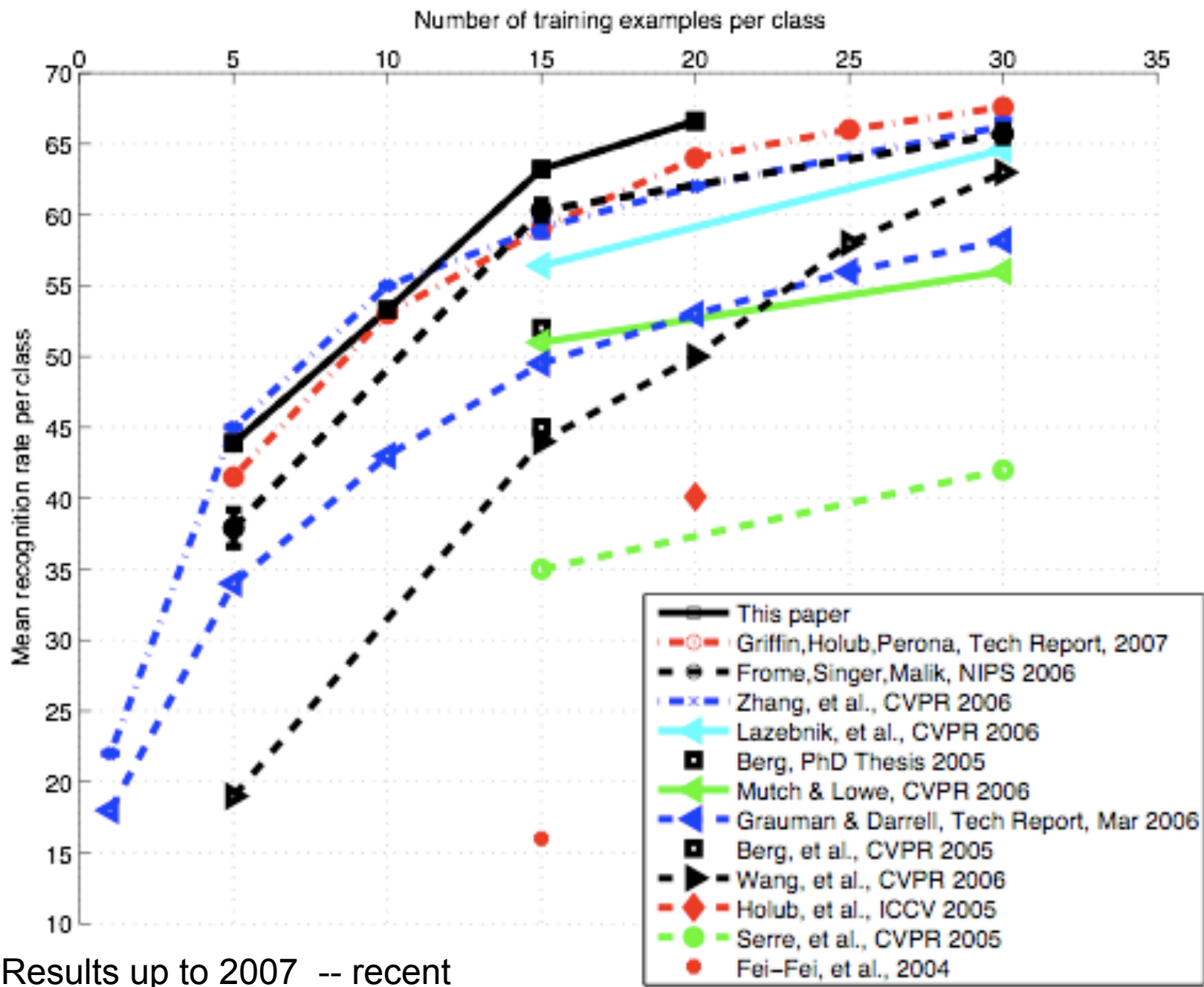
- Too easy?

- left-right aligned

- Rotation artifacts

- Saturated performance





Results up to 2007 -- recent methods obtain almost 100%



Caltech-256



- Smallest category size now 80 images
- About 30K images
- Harder
 - Not left-right aligned
 - No artifacts
 - Performance is halved
 - More categories
- New and larger clutter category



Caltech 256 images

baseball-bat



dog



basketball-hoop



kayac



traffic light



The PASCAL Visual Object Classes (VOC) Dataset and Challenge

Mark Everingham
Luc Van Gool
Chris Williams
John Winn
Andrew Zisserman

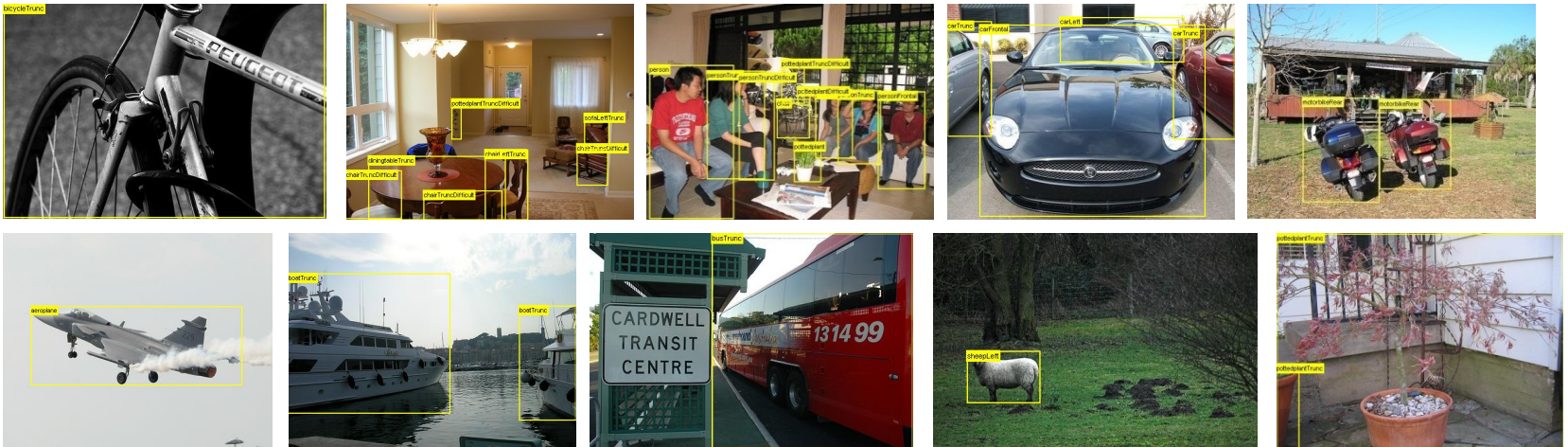


PASCAL

Pattern Analysis, Statistical Modelling and
Computational Learning

Dataset Content

- 20 classes: aeroplane, bicycle, boat, bottle, bus, car, cat, chair, cow, dining table, dog, horse, motorbike, person, potted plant, sheep, train, TV
- Real images downloaded from flickr, not filtered for “quality”



- Complex scenes, scale, pose, lighting, occlusion, ...

Annotation

- Complete annotation of all objects
- Annotated in one session with written guidelines

Occluded

Object is significantly occluded within BB

Difficult

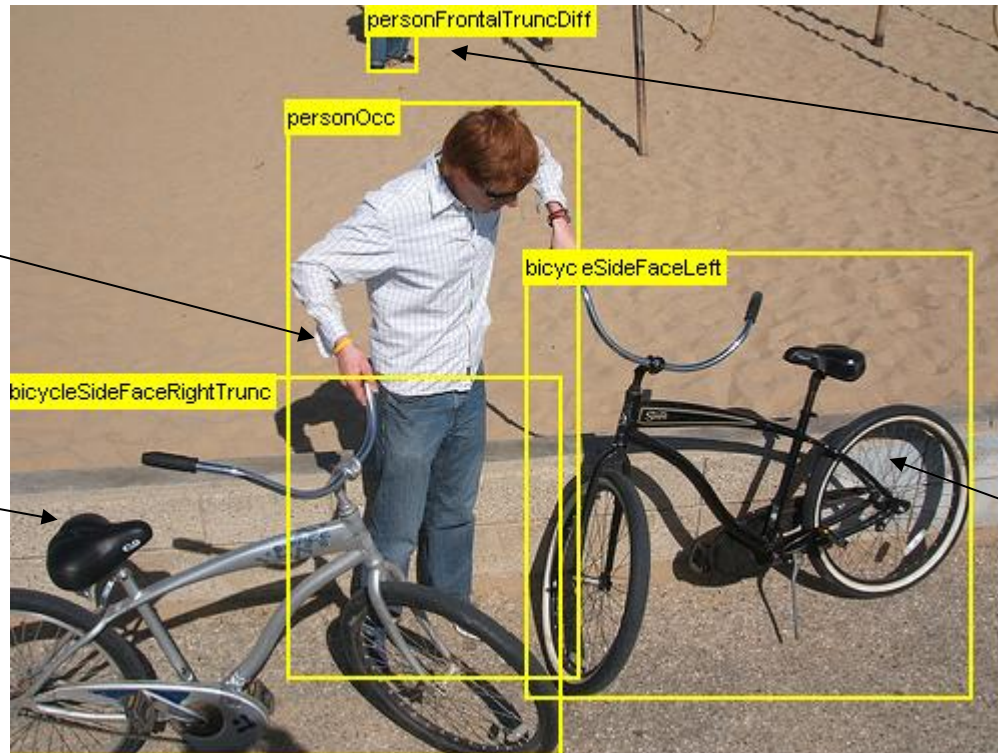
Not scored in evaluation

Truncated

Object extends beyond BB

Pose

Facing left



Examples

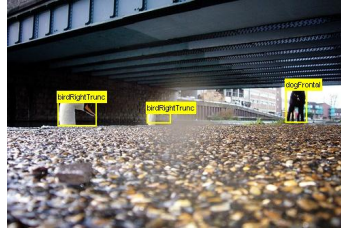
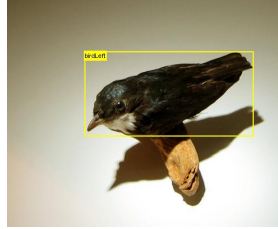
Aeroplane



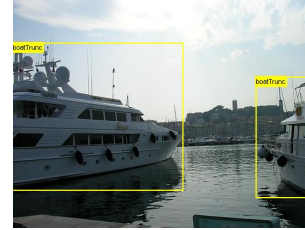
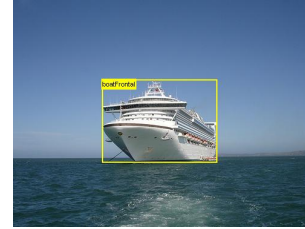
Bicycle



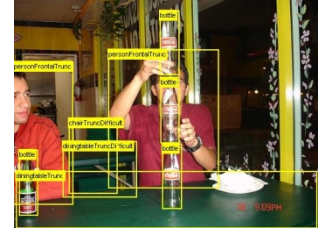
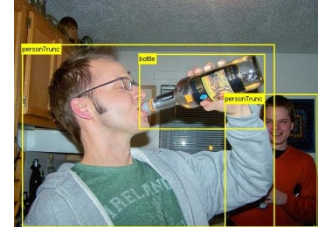
Bird



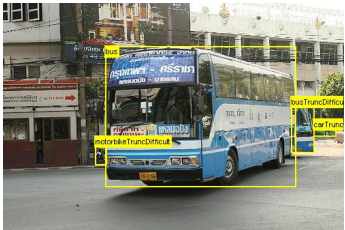
Boat



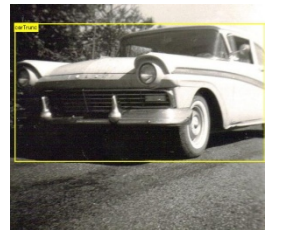
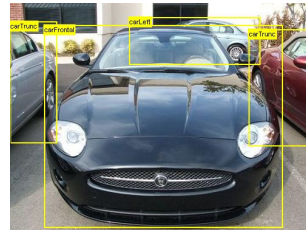
Bottle



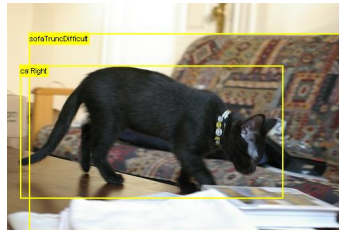
Bus



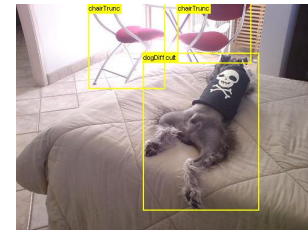
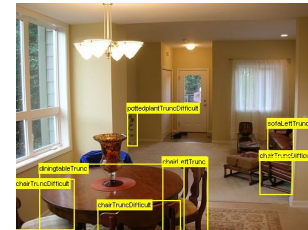
Car



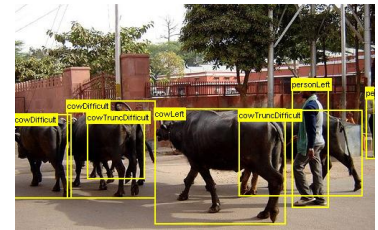
Cat



Chair



Cow



History

	Images	Objects	Classes	Entries	
2005	2,232	2,871	4	12	<i>Collection of existing and some new data.</i>
2006	5,304	9,507	10	25	<i>Completely new dataset from flickr (+MSRC)</i>
2007	9,963	24,640	20	28	<i>Increased classes to 20. Introduced tasters.</i>
2008	8,776	20,739	20		<i>Added “occlusion” flag. Reuse of taster data. Release detailed results to support “meta-analysis”</i>

- New dataset annotated annually
 - Annotation of test set is withheld until after challenge

Other recent datasets

ESP

[Ahn et al, 2006]

LabelMe

[Russell et al, 2005]

TinyImage

Torralba et al. 2007

Lotus Hill

[Yao et al, 2007]

MSRC

[Shotton et al. 2006]

LabelMe Please [contact us](#) if you find any bugs or have any suggestions.

Label as many objects and regions as you can in this image

Show me another image

Sign in (why?)

With your help, there are 91348 labelled objects in the database (more stats)

Instructions (Get more help)

Use your mouse to click around the boundary of some objects in this image. You will then be asked to enter the name of the object (examples: car, window).

Good Bad

Labeling tools

Erase segment Zoom Fit Image

Polygons in this image (XML)

door
door
road
stair
window
window
sidewalk
building region
house
window
window
window

3D object dataset [Savarese & Fei-Fei 07]

bicycle



car



cellphone



iron



mouse



shoe



stapler



toaster

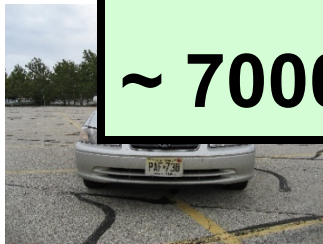


Poses

72

⋮

1



...



- 8 azimuth angles
- 3 zenith
- 3 distances

~ 7000 images!

1

2

...

10

Instances

Largest dataset for object categories up to date



J. Deng, H. Su. K. Li , L. Fei-Fei ,

- ~20K categories;
- 14 million images;
- ~700im/categ;
- free to public at **www.image-net.org**

<http://www.image-net.org>

IMAGENET

9,956,478 images, 14841 synsets indexed

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ImageNet is an image database organized according to the **WordNet** hierarchy (currently only the nouns), in which each node of the hierarchy is depicted by hundreds and thousands of images. Currently we have an average of over five hundred images per node. We hope ImageNet will become a useful resource for researchers, educators, students and all of you who share our passion for pictures.

[Click Here](#) to learn more about ImageNet, [Click Here](#) to join ImageNet mailing list.

SEARCH

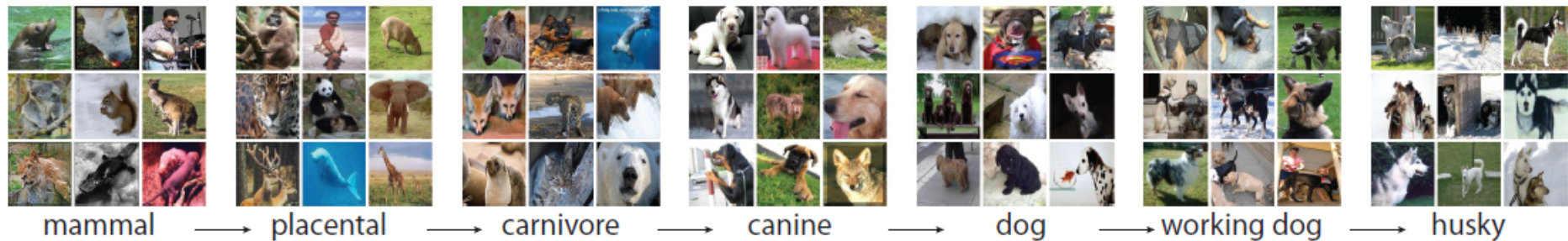


What do these images have in common? [Find out!](#)

Update Notice: ImageNet 2010 Spring Version will be released in April, 2010

IMAGENET is a knowledge ontology

- Taxonomy



- [S: \(n\) Eskimo dog, husky](#) (breed of heavy-coated Arctic sled dog)
 - [direct hypernym](#) / [inherited hypernym](#) / [sister term](#)
 - [S: \(n\) working dog](#) (any of several breeds of usually large powerful dogs bred to work as draft animals and guard and guide dogs)
 - [S: \(n\) dog, domestic dog, Canis familiaris](#) (a member of the genus Canis (probably descended from the common wolf) that has been domesticated by man since prehistoric times; occurs in many breeds) *"the dog barked all night"*
 - [S: \(n\) canine, canid](#) (any of various fissioned mammals with nonretractile claws and typically long muzzles)
 - [S: \(n\) carnivore](#) (a terrestrial or aquatic flesh-eating mammal) *"terrestrial carnivores have four or five clawed digits on each limb"*
 - [S: \(n\) placental, placental mammal, eutherian, eutherian mammal](#) (mammals having a placenta; all mammals except monotremes and marsupials)
 - [S: \(n\) mammal, mammalian](#) (any warm-blooded vertebrate having the skin more or less covered with hair; young are born alive except for the small subclass of monotremes and nourished with milk)
 - [S: \(n\) vertebrate, craniate](#) (animals having a bony or cartilaginous skeleton with a segmented spinal column and a large brain enclosed in a skull or cranium)
 - [S: \(n\) chordate](#) (any animal of the phylum Chordata having a notochord or spinal column)
 - [S: \(n\) animal, animate being, beast, brute, creature, fauna](#) (a living organism characterized by voluntary movement)
 - [S: \(n\) organism, being](#) (a living thing that has (or can develop) the ability to act or function independently)
 - [S: \(n\) living thing, animate thing](#) (a living (or once living) entity)
 - [S: \(n\) whole, unit](#) (an assemblage of parts that is regarded as a single entity) *"how big is that part compared to the whole?"; "the team is a unit"*
 - [S: \(n\) object, physical object](#) (a tangible and visible entity; an entity that can cast a shadow) *"it was full of rackets, balls and other objects"*
 - [S: \(n\) physical entity](#) (an entity that has physical existence)
 - [S: \(n\) entity](#) (that which is perceived or known or inferred to have its own distinct existence (living or nonliving))

More Datasets....



UIUC Cars (2004)

S. Agarwal, A. Awan, D. Roth



CMU/VASC Faces (1998)

H. Rowley, S. Baluja, T. Kanade



FERET Faces (1998)

P. Phillips, H. Wechsler, J. Huang, P. Raus



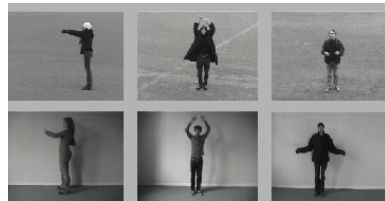
COIL Objects (1996)

S. Nene, S. Nayar, H. Murase



MNIST digits (1998-10)

Y LeCun & C. Cortes



KTH human action (2004)

I. Leptev & B. Caputo



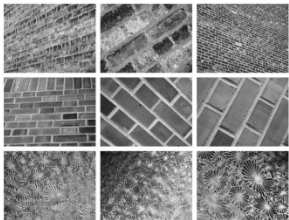
Sign Language (2008)

P. Buehler, M. Everingham, A. Zisserman



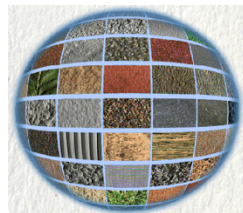
Segmentation (2001)

D. Martin, C. Fowlkes, D. Tal, J. Malik.



3D Textures (2005)

S. Lazebnik, C. Schmid, J. Ponce



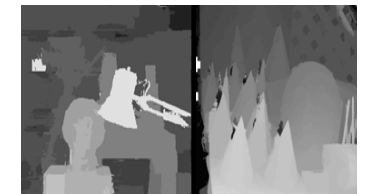
CuRRET Textures (1999)

K. Dana B. Van Ginneken S. Nayar J. Koenderink



CAVIAR Tracking (2005)

R. Fisher, J. Santos-Victor J. Crowley



Middlebury Stereo (2002)

D. Scharstein R. Szeliski

Links to datasets

The next tables summarize some of the available datasets for training and testing object detection and recognition algorithms. These lists are far from exhaustive.

Databases for object localization

| | | | |
|-----------------------|--|--------------------|------------------------|
| CMU/MIT frontal faces | vasc.ri.cmu.edu/idb/html/face/frontal_images
cbcl.mit.edu/software-datasets/FaceData2.html | Patches | Frontal faces |
| Graz-02 Database | www.emt.tugraz.at/~pinz/data/GRAZ_02/ | Segmentation masks | Bikes, cars, people |
| UIUC Image Database | l2r.cs.uiuc.edu/~cogcomp/Data/Car/ | Bounding boxes | Cars |
| TU Darmstadt Database | www.vision.ethz.ch/leibe/data/ | Segmentation masks | Motorbikes, cars, cows |
| LabelMe dataset | people.csail.mit.edu/brussell/research/LabelMe/intro.html | Polygonal boundary | >500 Categories |

Databases for object recognition

| | | | |
|-------------|--|--------------------|----------------|
| Caltech 101 | www.vision.caltech.edu/Image_Datasets/Caltech101/Caltech101.html | Segmentation masks | 101 categories |
| Caltech 256 | http://www.vision.caltech.edu/Image_Datasets/Caltech256/ | Bounding Box | 256 Categories |
| COIL-100 | www1.cs.columbia.edu/CAVE/research/softlib/coil-100.html | Patches | 100 instances |
| NORB | www.cs.nyu.edu/~ylclab/data/norb-v1.0/ | Bounding box | 50 toys |

On-line annotation tools

| | | | |
|----------|--|---------------------------|------------------------|
| ESP game | www.espgame.org | Global image descriptions | Web images |
| LabelMe | people.csail.mit.edu/brussell/research/LabelMe/intro.html | Polygonal boundary | High resolution images |

Collections

| | | | |
|--------|---|---------------------|---------|
| PASCAL | http://www.pascal-network.org/challenges/VOC/ | Segmentation, boxes | various |
|--------|---|---------------------|---------|