



Large Scale Visual Recognition Challenge 2010

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Large Scale Recognition

- Millions to billions of images
- hundreds of thousands of labels
- Classifiers should provide information for indexing and retrieval
- Complement current Pascal VOC competitions
- This year's challenge:
 - Image categorization
 - 1000 categories
 - 1.2 million training images
 - 50 thousand new verification images
 - 150 thousand new test images
 - Success would provide techniques for image indexing and search
- Next year
 - Categorization and detection (now with bounding boxes!)
 - Object detection and image categorization but not full image parsing

Source for categories and training data

ImageNet

- 12 million images, 15 thousand categories IMAGENET
- Image found via web searches for WordNet noun synsets
- Hand verified using Mechanical
- All new data for validation and testing this year

WordNet

- Source of the labels
- Semantic hierarchy
- Contains large fraction of English nouns
- Also used to collect other datasets like tiny images (Torralba et al)
- Note that categorization is not the end goal, but should provide information for other tasks, so idiosyncrasies of WordNet may be less critical

http://www.image-net.org



11,231,732 images, 15589 synsets indexed

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About

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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 the ImageNet mailing list.



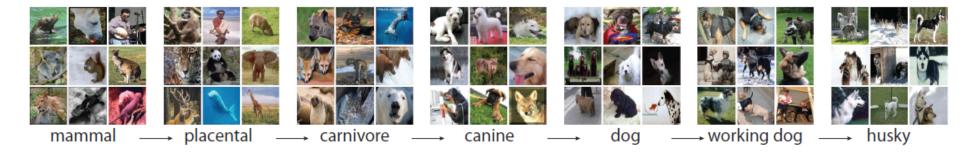


What do these images have in common? Find out!

ImageNet 2010 Spring Release is up! Click here to check out what's new!

IM GENET is a knowledge ontology

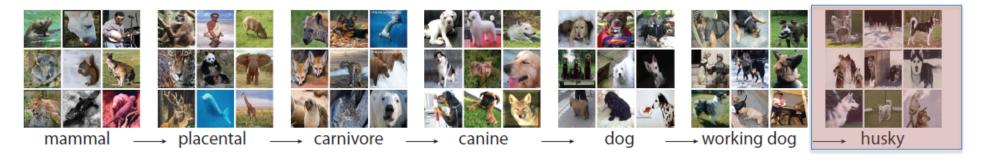
Taxonomy



- S: (n) Eskimo dog, husky (breed of heavy-coated Arctic sled dog)
 - o 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 fissiped 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))

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Step 1: Collect Candidate Images from the Internet

- Query expansion
 - Synonyms: German shepherd, German police dog, German shepherd dog, Alsatian
 - Appending words from ancestors: sheepdog, dog
- Multiple languages
 - Italian, Dutch, Spanish, Chinese
 - e.g. ovejero alemán, pastore tedesco,德国牧羊犬
- More engines YAHOO! (6) Live Search







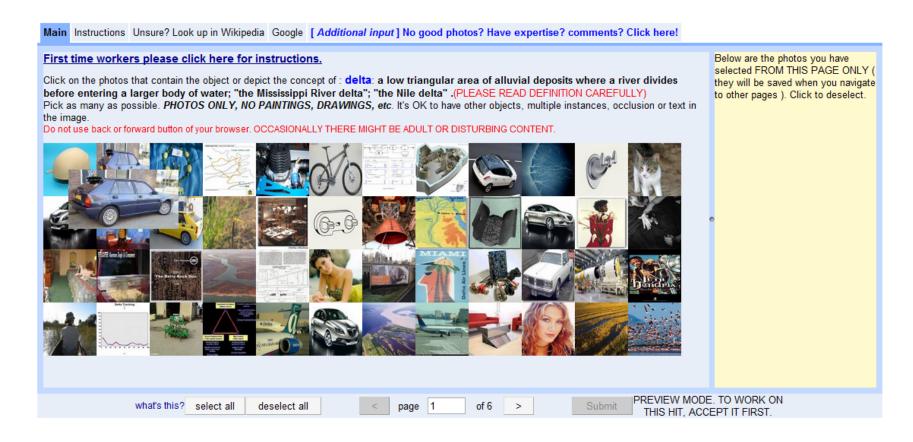


Parallel downloading



IMAGENET Basic User Interface

Click on the good images.



Definition quiz

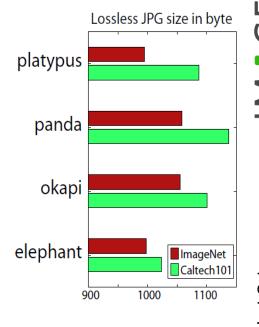
| This | HIT | is al | out ' | 'de | lta' |
|------|-----|-------|-------|-----|------|
| | | | | | |

Definition: a low triangular area of alluvial deposits where a river divides before entering a larger body of water; "the Mississippi River delta"; "the Nile delta"

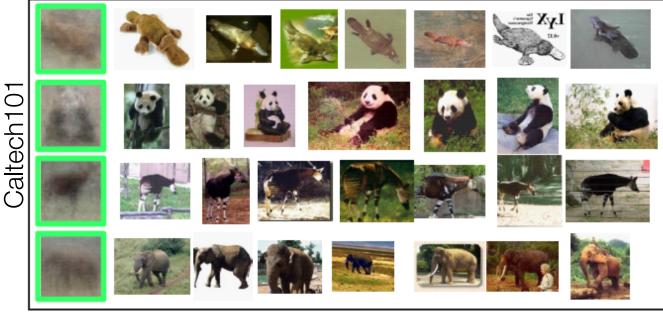
Please read the above definition carefully. 'delta' might mean something different from what you think.

I HAVE READ IT

Diversity







This year's challenge

- Given an image predict categories of objects that may be present in the image
- 1000 "leaf" categories from ImageNet
- Two evaluation criteria based on cost averaged over test images
 - Flat cost pay 0 for correct category, 1 otherwise
 - Hierarchical cost pay 0 for correct category, height of least common ancestor in WordNet for any other category
- Allow a shortlist of up to 5 predictions
 - Use the lowest cost prediction each test image
 - Allows for incomplete labeling of all categories in an image

Participation

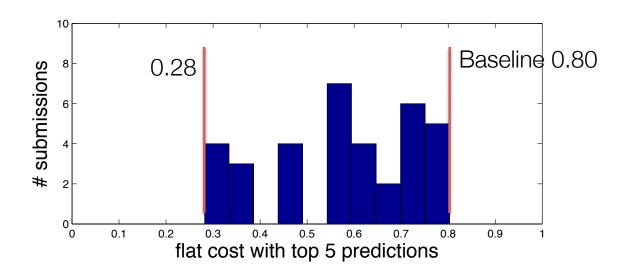
150+ registrations

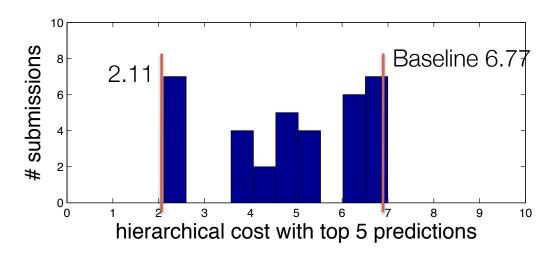
35 submissions from 11 teams

Teams from around the world

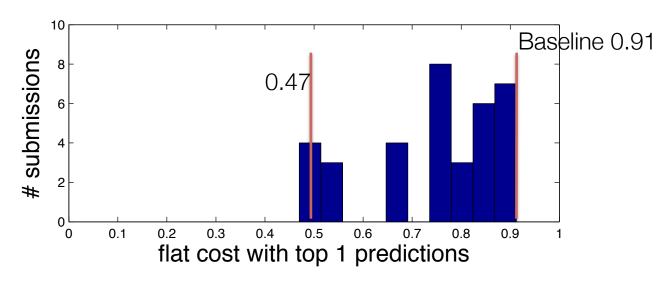
- •NEC-UIUC, USA
- •XRCE, France
- •ISIL, University of Tokyo, Japan
- •UC Irvine, USA
- •MIT, USA
- Nanyang Technological University, Singapore
- •LIG Grenoble, France
- •IBM-ensemble, USA
- •SRI International, USA
- National Institute of Informatics, Tokyo, Japan
- Harbin Institute of Technology, China

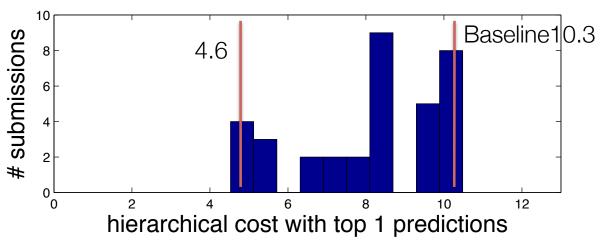
Results: 5 predictions





Results: 1 prediction





Flat versus hierarchical

Performance ordering was almost the same for flat and hierarchical evaluations.

Winners

Winner: NEC-UIUC

Yuanqing Lin, Fengjun Lv, Shenghuo Zhu, Ming Yang, Timothee Cour, Kai Yu (NEC). LiangLiang Cao, Zhen Li, Min-Hsuan Tsai, Xi Zhou, Thomas Huang (UIUC). Tong Zhang (Rutgers).

Honorable mention: XRCE

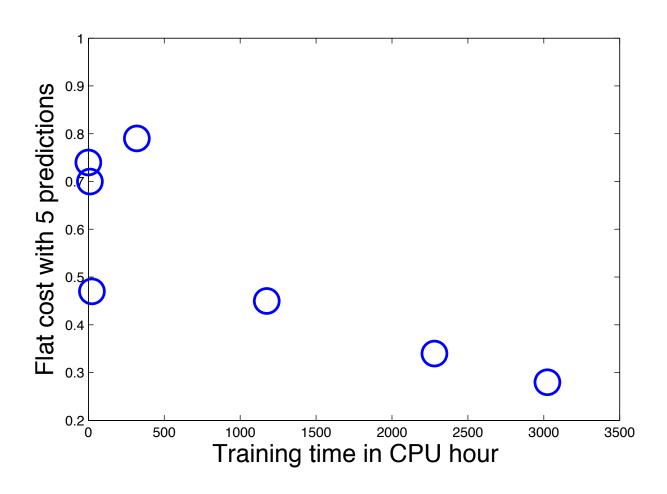
Jorge Sanchez, Florent Perronnin, Thomas Mensink (XRCE)

Ranking by learning methods

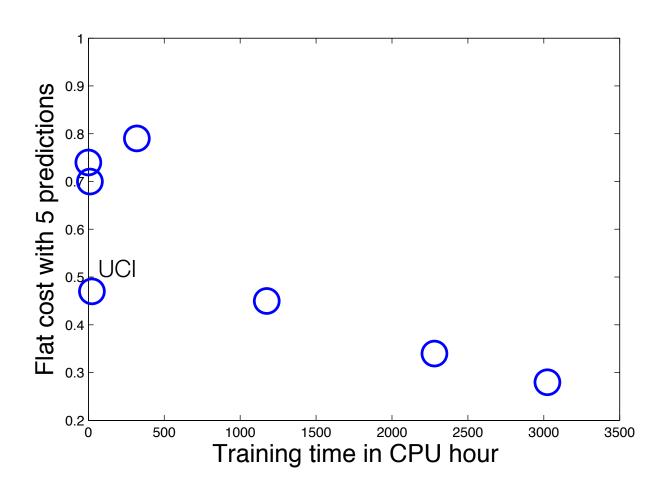
- Descriptor Coding + SVM, 0.28 --- NEC-UIUC
- Fisher kernel + SVM, 0.34 --- XRCE
- LI2C, 0.58 --- NTU_WZX
- KNN, 0.61 --- LIG
- Canonical Correlation Analysis, 0.79 -- NII

^{*}number indicates flat cost with 5 predictions

Computation



Computation



Hardest categories

- Logwood, 0.92
- China tree, 0.92
- Red beech, 0.88
- Cap opener, 0.88
- Paintbrush, 0.87
- Needle, 0.87
- Weasel, 0.87
- Syringe, 0.87
- Alder tree, 0.87
- Sorrel tree, 0.86















Easiest categories

- Odometer, 0.12
- Website, 0.14
- Lunar crater, 0.18
- Geyser, 0.20
- Monarch butterfly, 0.21
- Manhole cover, 0.22
- Rapeseed, 0.22
- Cliff dwelling, 0.24
- Sunflower, 0.25
- Shoji, 0.26

















^{*}number indicates average flat cost with 5 predictions from all submissions

Hardest subtrees

| Subtree | # descendants | Average flat cost* |
|---------------------------|---------------|--------------------|
| Implement (small tools) | 50 | 0.68 |
| Reptile | 29 | 0.65 |
| Fruit | 55 | 0.64 |
| Dog | 31 | 0.62 |
| Tree | 68 | 0.62 |

^{*}number indicates average flat cost with 5 predictions from all submissions

Easiest subtrees

| Subtree | # descendants | Average flat cost* |
|---------------------------------|---------------|--------------------|
| Geological formation | 12 | 0.35 |
| Vehicle | 60 | 0.46 |
| Flower | 87 | 0.49 |
| Structure | 53 | 0.50 |
| Equipment (e.g. home appliance) | 45 | 0.52 |

^{*}number indicates average flat cost with 5 predictions from all submissions

Most difficult sunflowers















Easiest sunflowers













Most difficult?













Most difficult paintbrushes!





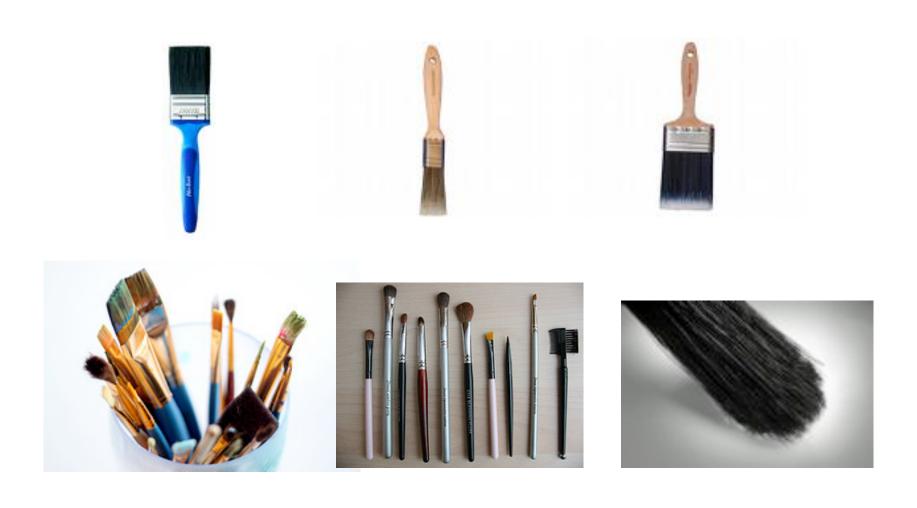






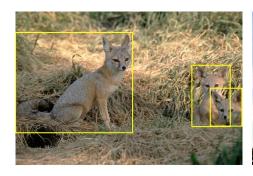


Easiest paintbrushes

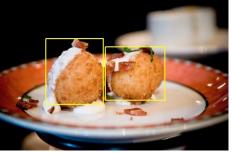


2011 Large Scale Visual Recognition Challenge!

- Data:
 - Bounding boxes
 - Not yet full parsing of images
- Task:
 - Image categorization
 - Object detection/localization









Full 2010 results and team info online

http://www.image-net.org/challenges/LSVRC/2010/results