

The PASCAL Visual Object Classes Challenge 2010 (VOC2010)

Part 2 – Detection Task

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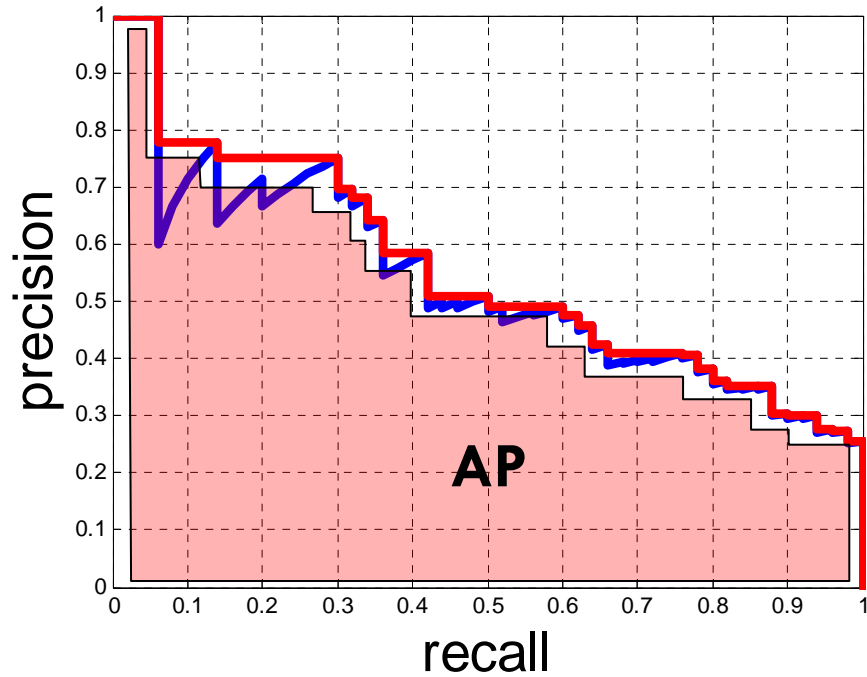
Andrew Zisserman



Detection Challenge

- Predict the bounding boxes of all objects of a given class in an image (if any)
- Competition 3: Train on the supplied data
 - Which methods perform best given specified training data?
- Competition 4: Train on any (non-test) data
 - How well do state-of-the-art methods perform on these problems?

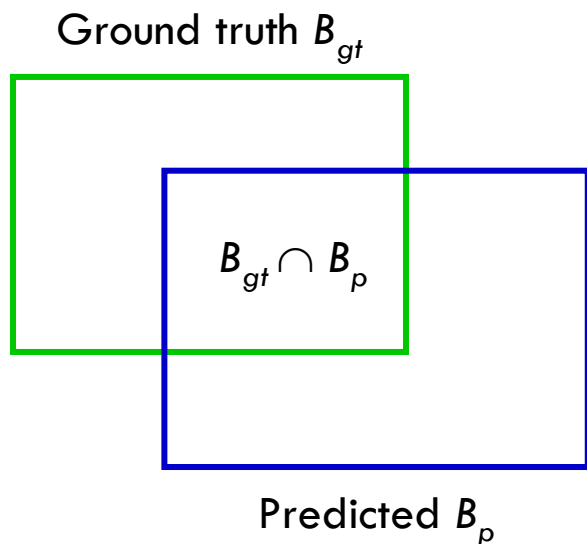
Average Precision: VOC2010



- Interpolate curve to create version for which the precision is monotonically non-increasing
- Measure area under interpolated curve
- Sawtooth shape is ignored
- Area is measured with maximum accuracy
- Good score requires both high precision **and** recall

Evaluating Bounding Boxes

- Area of Overlap (AO) Measure



$$AO(B_{gt}, B_p) = \frac{|B_{gt} \cap B_p|}{|B_{gt} \cup B_p|}$$

- Need to define a threshold t such that $AO(B_{gt}, B_p)$ implies a correct detection: 50%

Methods

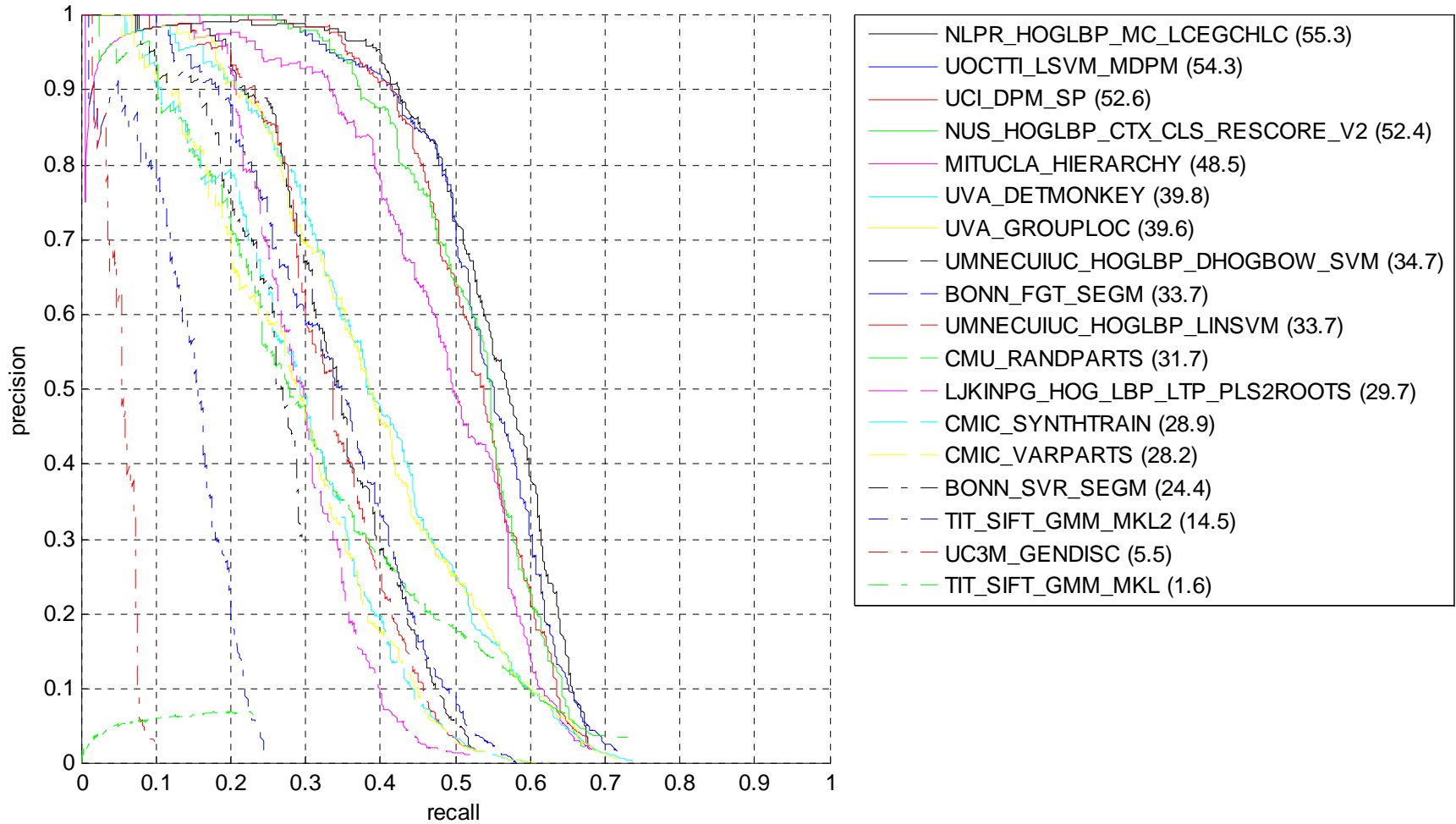
- 22 Methods, 15 Groups
 - VOC2009: 18 Methods, 12 Groups
- Methods
 - Sliding window, SVM, multiple features e.g. HOG, LBP
 - Parts-based models, hierarchical models
 - Segmentation-based object hypotheses (“jumping window”)
 - Combination with whole-image classification
- Use of external training data
 - Additional data (e.g. annotated with Mechanical Turk)
 - More detailed annotation e.g. parts

AP by Class/Method

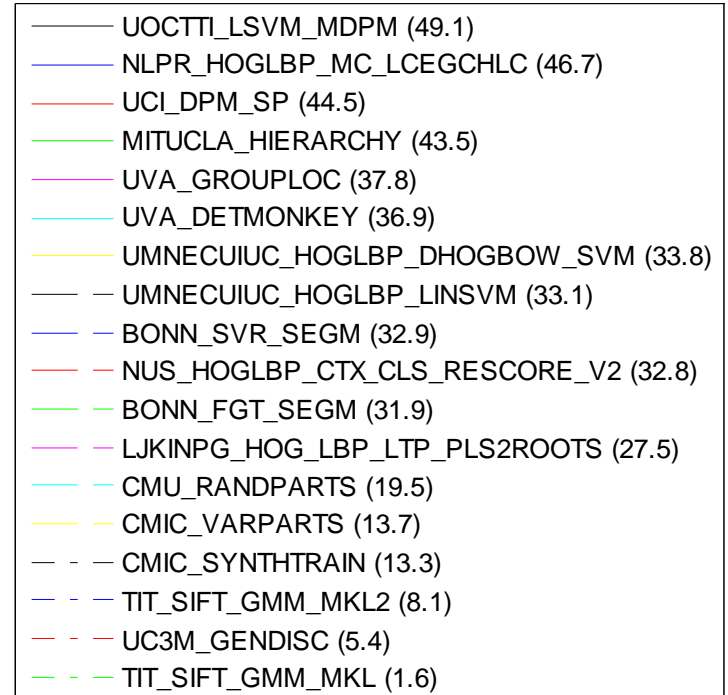
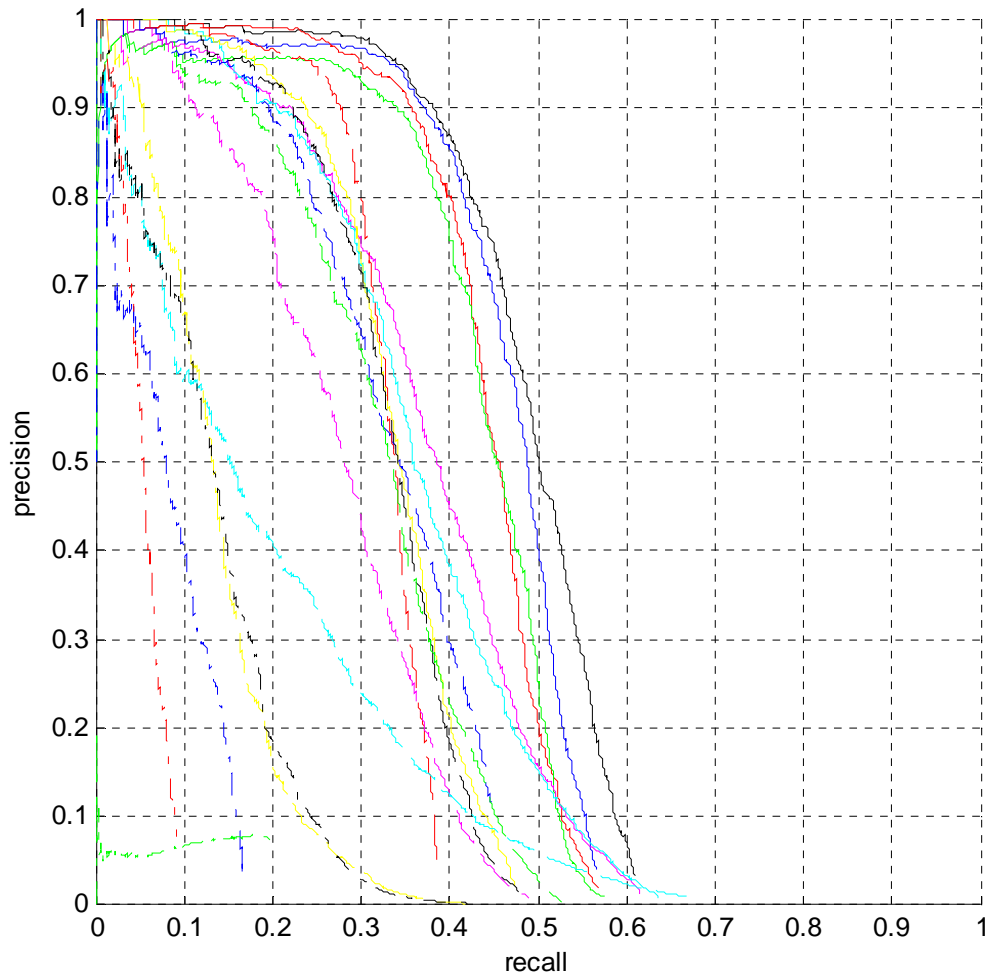
(1st, 2nd, 3rd place)

| | aero plane | bicycle | bird | boat | bottle | bus | car | cat | chair | cow | dining table | dog | horse | motor bike | person | potted plant | sheep | sofa | train | tv/ monitor |
|-----------------------------------|---------------|---------|------|------|--------|------|------|------|-------|------|-----------------|------|-------|---------------|--------|-----------------|-------|------|-------|----------------|
| BONN_FGT_SEG | 52.7 | 33.7 | 13.2 | 11.0 | 14.2 | 43.1 | 31.9 | 35.6 | 5.7 | 25.4 | 14.4 | 20.6 | 38.1 | 41.7 | 25.0 | 5.8 | 26.3 | 18.1 | 37.6 | 28.1 |
| BONN_SVR_SEG | 50.5 | 24.4 | 17.1 | 13.3 | 10.9 | 39.5 | 32.9 | 36.5 | 5.6 | 16.0 | 6.6 | 22.3 | 24.9 | 29.0 | 29.8 | 6.7 | 28.4 | 13.3 | 32.1 | 27.2 |
| CMIC_SYNTHTRAIN | - | 28.9 | - | - | - | 30.2 | 13.3 | - | - | - | - | - | 26.2 | 28.1 | 13.2 | - | - | - | 18.8 | 25.7 |
| CMIC_VARPARTS | - | 28.2 | - | - | - | 26.9 | 13.7 | - | - | - | - | - | 23.5 | 24.7 | 16.1 | - | - | - | 18.8 | 24.5 |
| CMU_RANDPARTS | 23.8 | 31.7 | 1.2 | 3.4 | 11.1 | 29.7 | 19.5 | 14.2 | 0.8 | 11.1 | 7.0 | 4.7 | 16.4 | 31.5 | 16.0 | 1.1 | 15.6 | 10.2 | 14.7 | 21.0 |
| CMU_RANDPARTS _MAXSCORE | - | - | 2.7 | - | - | - | - | 16.2 | - | 10.6 | 8.5 | - | - | - | 17.9 | - | - | - | 15.7 | - |
| LJKINPG_HOG_LBP _LTP_PLS2ROOTS | 32.7 | 29.7 | 0.8 | 1.1 | 19.8 | 39.4 | 27.5 | 8.6 | 4.5 | 8.1 | 6.3 | 11.0 | 22.9 | 34.1 | 24.6 | 3.1 | 24.0 | 2.0 | 23.5 | 27.0 |
| MITUCLA_HIERARCHY | 54.2 | 48.5 | 15.7 | 19.2 | 29.2 | 55.5 | 43.5 | 41.7 | 16.9 | 28.5 | 26.7 | 30.9 | 48.3 | 55.0 | 41.7 | 9.7 | 35.8 | 30.8 | 47.2 | 40.8 |
| NLPR_HOGLBP_MC _LCEGCHLC | 53.3 | 55.3 | 19.2 | 21.0 | 30.0 | 54.4 | 46.7 | 41.2 | 20.0 | 31.5 | 20.7 | 30.3 | 48.6 | 55.3 | 46.5 | 10.2 | 34.4 | 26.5 | 50.3 | 40.3 |
| NUS_HOGLBP_CTX _CLS_RESCORE_V2 | 49.1 | 52.4 | 17.8 | 12.0 | 30.6 | 53.5 | 32.8 | 37.3 | 17.7 | 30.6 | 27.7 | 29.5 | 51.9 | 56.3 | 44.2 | 9.6 | 14.8 | 27.9 | 49.5 | 38.4 |
| TIT_SIFT_GMM_MKL | 10.5 | 1.6 | 1.2 | 0.9 | 0.1 | 2.8 | 1.6 | 6.7 | 0.1 | 2.0 | 0.4 | 3.0 | 2.0 | 4.4 | 2.0 | 0.3 | 1.1 | 1.2 | 2.1 | 1.9 |
| TIT_SIFT_GMM_MKL2 | 20.0 | 14.5 | 3.8 | 1.2 | 0.5 | 17.6 | 8.1 | 28.5 | 0.1 | 2.9 | 3.1 | 17.5 | 7.2 | 18.8 | 3.3 | 0.8 | 2.9 | 6.3 | 7.6 | 1.1 |
| UC3M_GENDISC | 15.8 | 5.5 | 5.6 | 2.3 | 0.3 | 10.2 | 5.4 | 12.6 | 0.5 | 5.6 | 4.5 | 7.7 | 11.3 | 12.6 | 5.3 | 1.5 | 2.0 | 5.9 | 9.1 | 3.2 |
| UCI_DPM_SP | 46.1 | 52.6 | 13.8 | 15.5 | 28.3 | 53.2 | 44.5 | 26.6 | 17.6 | - | 16.1 | 20.4 | 45.5 | 51.2 | 43.5 | 11.6 | 30.9 | 20.3 | 47.6 | - |
| UMNECUIUC_HOGLBP _DHOGBOW_SVM | 40.4 | 34.7 | 2.7 | 8.4 | 26.0 | 43.1 | 33.8 | 17.2 | 11.2 | 14.3 | 14.4 | 14.9 | 31.8 | 37.3 | 30.0 | 6.4 | 25.2 | 11.6 | 30.0 | 35.7 |
| UMNECUIUC_HOGLBP _LINSVM | 37.9 | 33.7 | 2.7 | 6.5 | 25.3 | 37.5 | 33.1 | 15.5 | 10.9 | 12.3 | 12.5 | 13.7 | 29.6 | 34.5 | 33.8 | 7.2 | 22.9 | 9.9 | 28.9 | 34.1 |
| UOCTI_LSVM_MDPM | 52.4 | 54.3 | 13.0 | 15.6 | 35.1 | 54.2 | 49.1 | 31.8 | 15.5 | 26.2 | 13.5 | 21.5 | 45.4 | 51.6 | 47.5 | 9.1 | 35.1 | 19.4 | 46.6 | 38.0 |
| UVA_DETMONKEY | 56.7 | 39.8 | 16.8 | 12.2 | 13.8 | 44.9 | 36.9 | 47.7 | 12.1 | 26.9 | 26.5 | 37.2 | 42.1 | 51.9 | 25.7 | 12.1 | 37.8 | 33.0 | 41.5 | 41.7 |
| UVA_GROUPLOC | 58.4 | 39.6 | 18.0 | 13.3 | 11.1 | 46.4 | 37.8 | 43.9 | 10.3 | 27.5 | 20.8 | 36.0 | 39.4 | 48.5 | 22.9 | 13.0 | 36.8 | 30.5 | 41.2 | 41.9 |

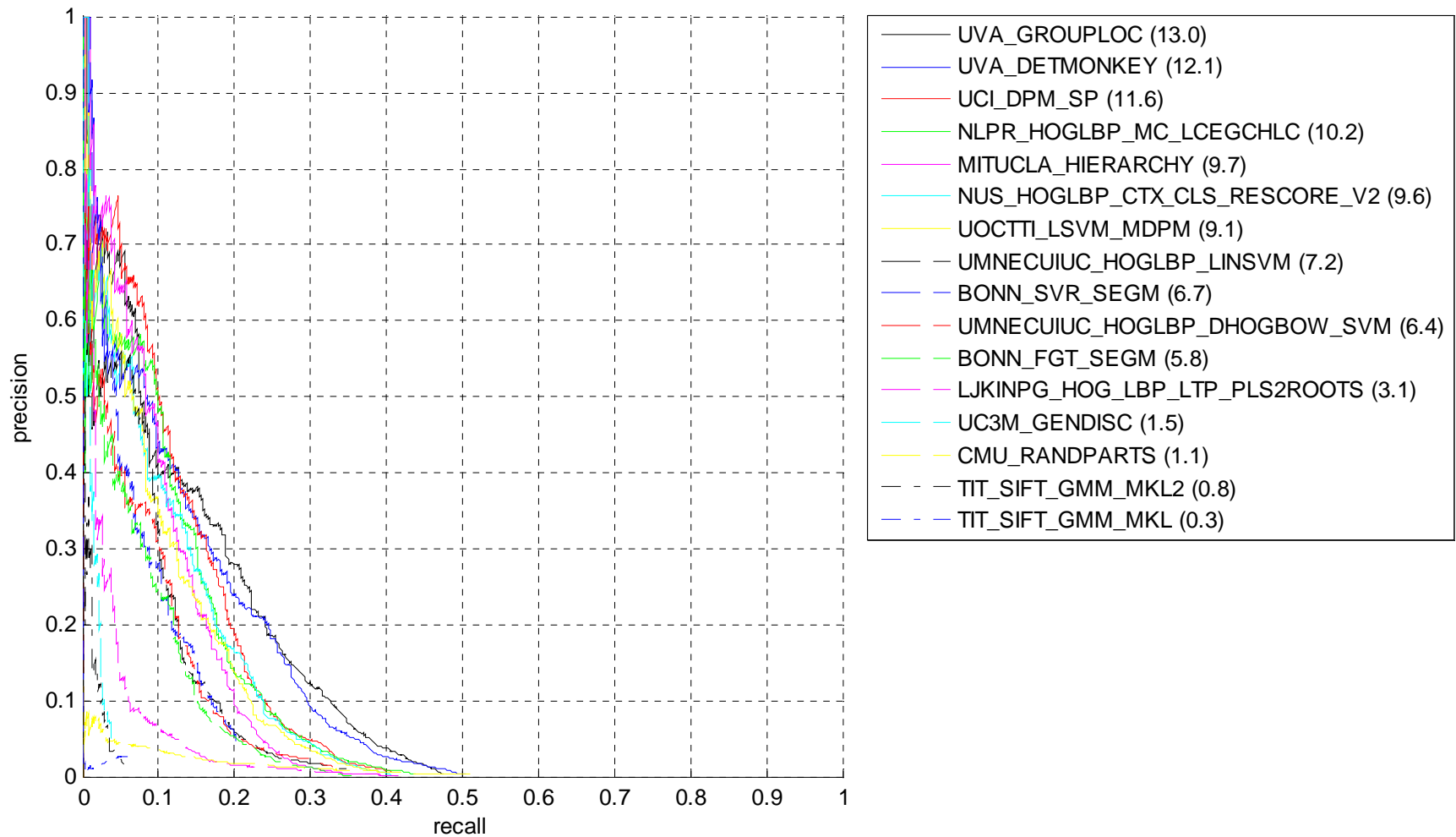
Precision/Recall - Bicycle



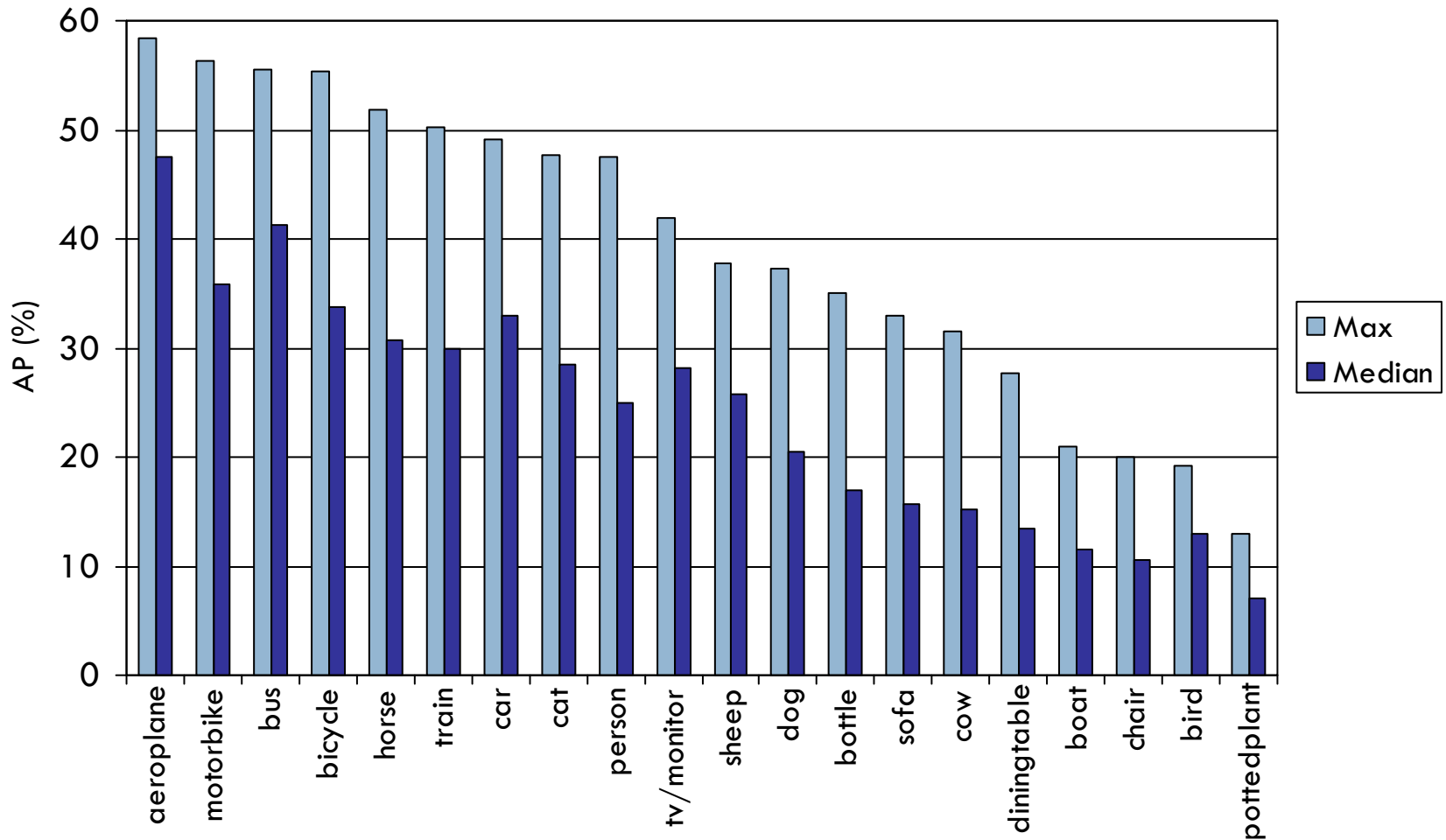
Precision/Recall - Car



Precision/Recall – Potted plant

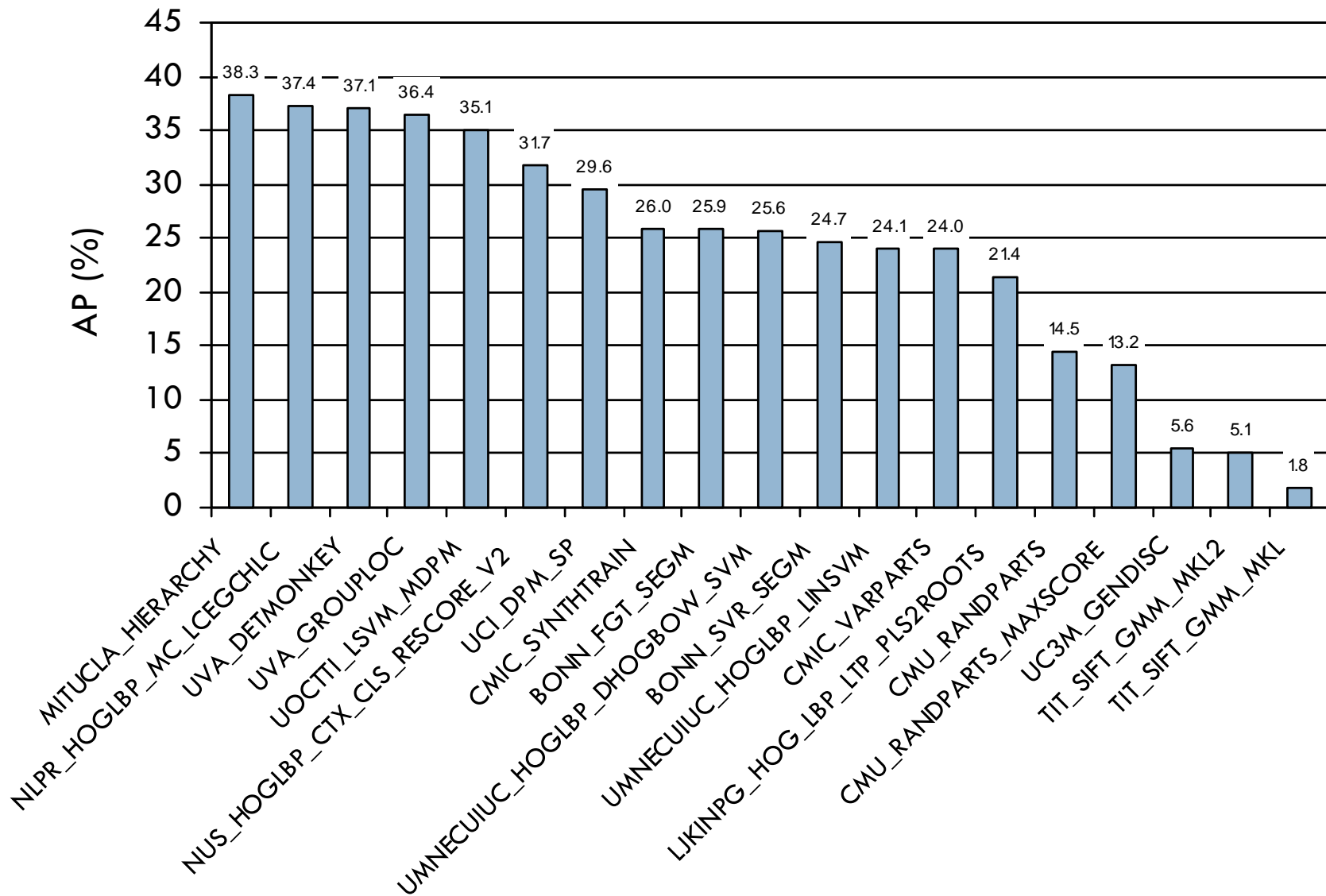


AP by Class



- Max AP: 58.4% (aeroplane) .. 13.0% (potted plant)

Median AP by Method



True Positives - Person

UOCTTI_LSVM_MDPM



NLPR_HOGLBP_MC_LCEGCHLC

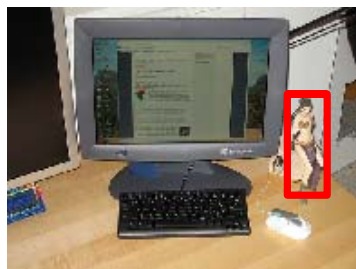


NUS_HOGLBP_CTX_CLS_RESCORE_V2

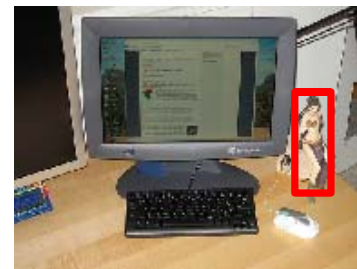
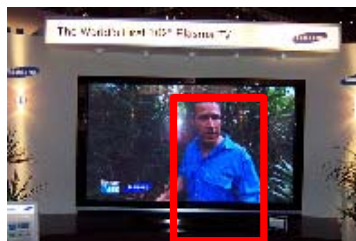


False Positives - Person

UOCTTI_LSVM_MDPM



NLPR_HOGLBP_MC_LCEGCHLC



NUS_HOGLBP_CTX_CLS_RESCORE_V2



“Near Misses” - Person

UOCTTI_LSVM_MDPM



NLPR_HOGLBP_MC_LCEGCHLC



NUS_HOGLBP_CTX_CLS_RESCORE_V2

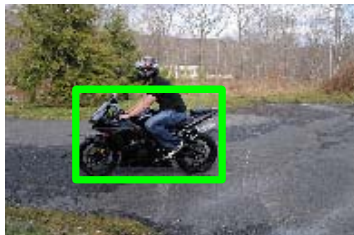


True Positives - Motorbike

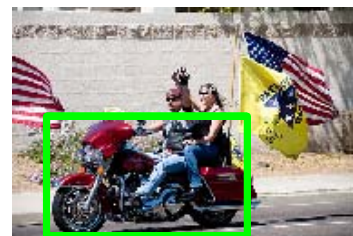
MITUCLA_HIERARCHY



NLPR_HOGLBP_MC_LCEGCHLC



NUS_HOGLBP_CTX_CLS_RESCORE_V2



False Positives - Motorbike

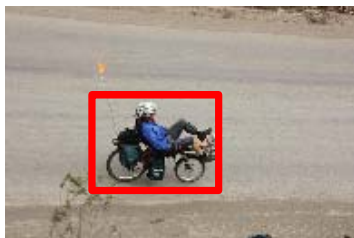
MITUCLA_HIERARCHY



NLPR_HOGLBP_MC_LCEGCHLC

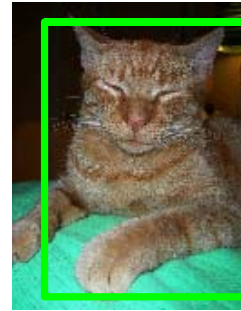


NUS_HOGLBP_CTX_CLS_RESCORE_V2



True Positives - Cat

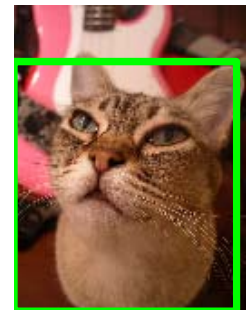
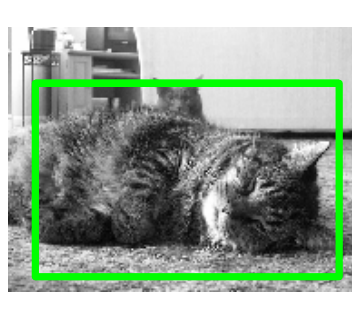
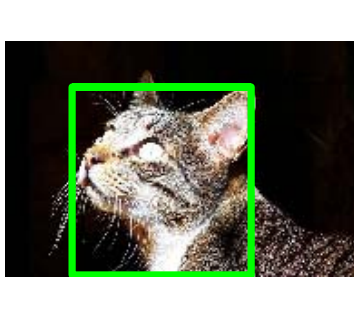
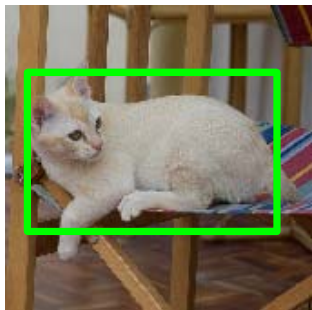
UVA_DETMONKEY



UVA_GROUPLOC



MITUCLA_HIERARCHY



False Positives - Cat

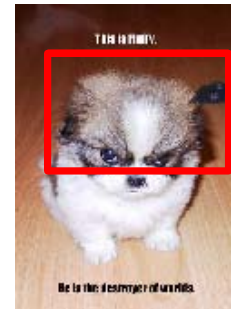
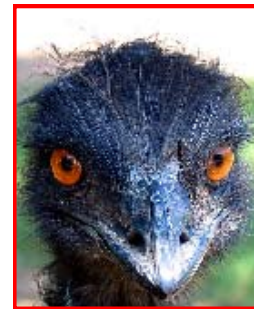
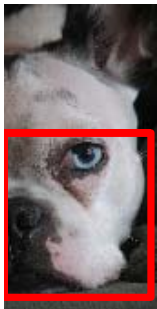
UVA_DETMONKEY



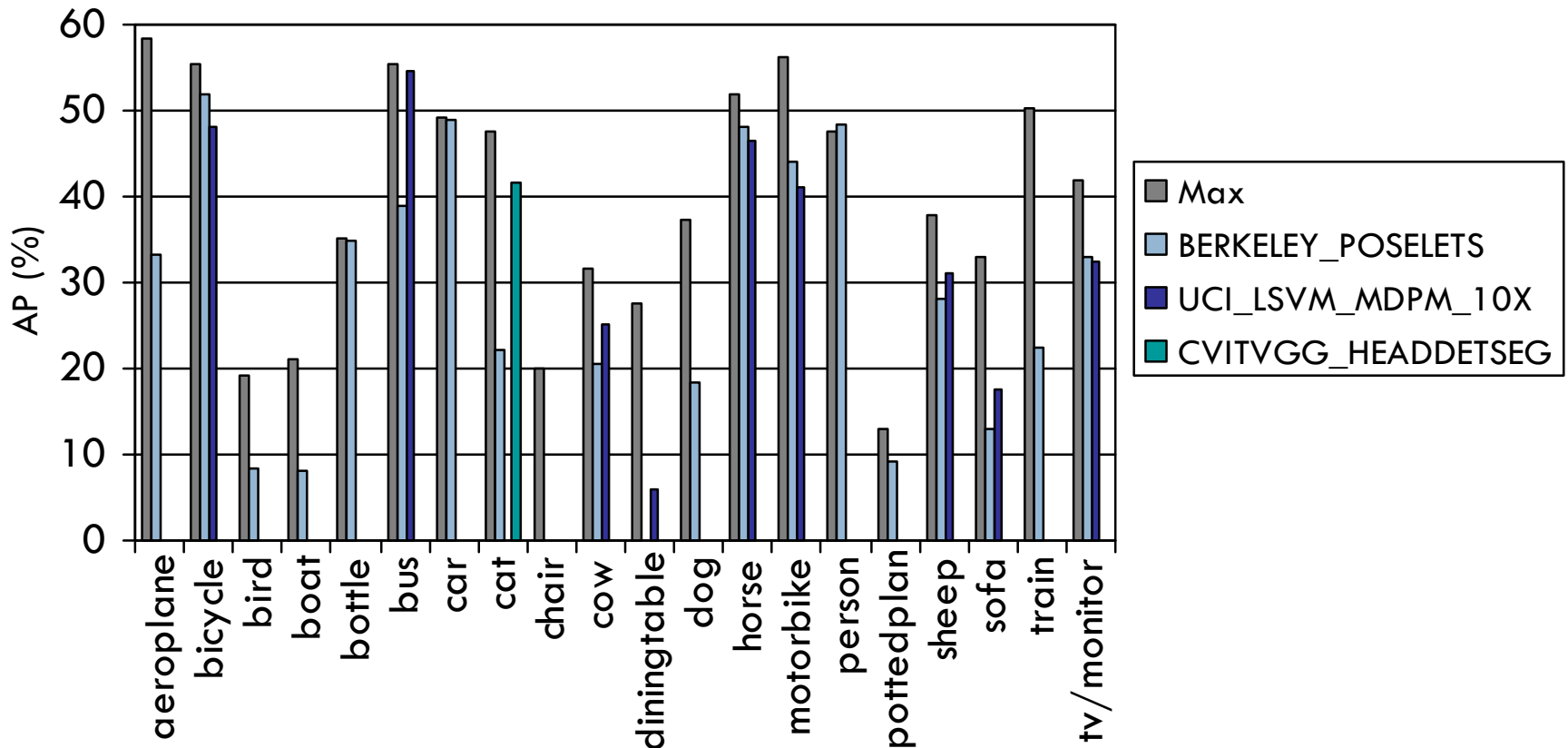
UVA_GROUPLOC



MITUCLA HIERARCHY

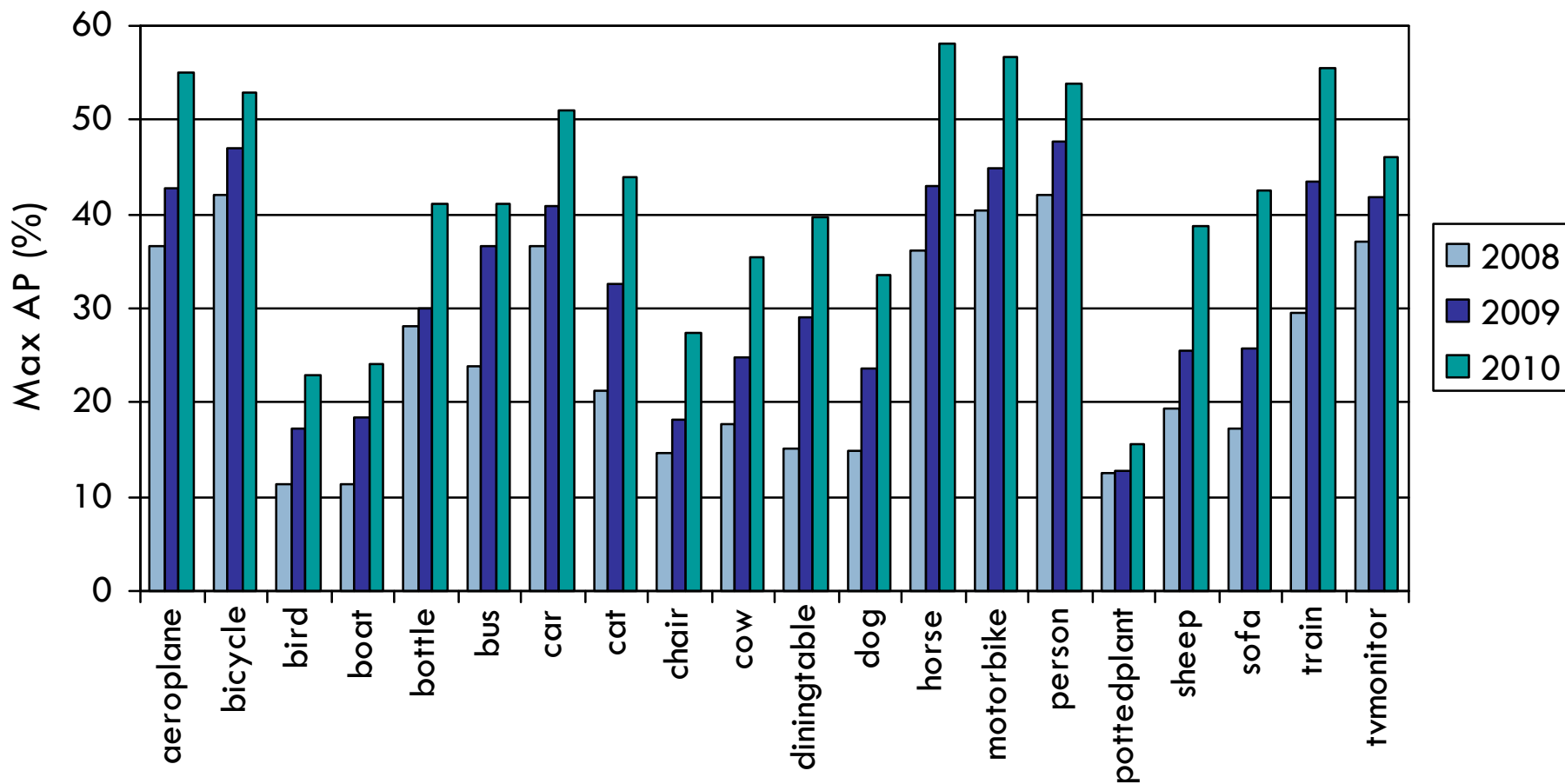


Additional Training Data/Annotation



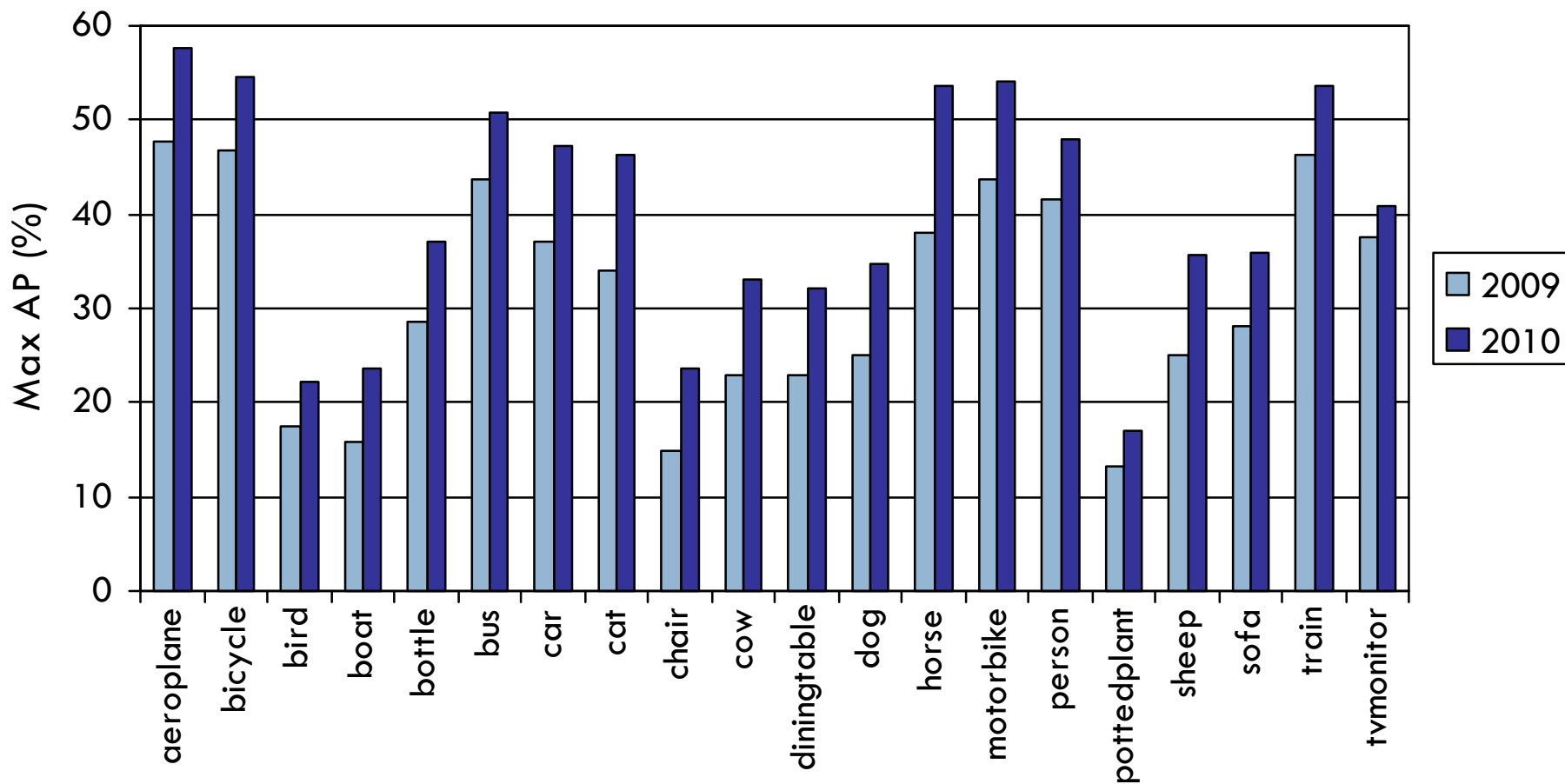
- BERKELEY use additional training data and annotation (manually labelled key points). Improves on best results for “person” class alone (1%).
- UCI annotated 10 times bigger training sets for some classes using Mechanical Turk. Does not improve on best results.
- CVIT/VGG used proprietary “cat” training set. Does not improve on best result.

Progress 2008-2010



- Results on 2008 data improve for best 2009 and 2010 methods for all categories, by over 100% for some categories
 - Caveat: Better methods or more training data?

Progress 2009-2010



- Best 2010 methods improve on 2009 results for all categories
 - Caveat: Better methods or more training data?

Prizes



- **Winner:**

- **NLPR_HOGLBP_MC_LCEGCHLC**

- Yinan Yu, Junge Zhang, Yongzhen Huang, Shuai Zheng, Weiqiang Ren, Chong Wang, Kaiqi Huang, Tieniu Tan
National Laboratory of Pattern Recognition, Institute of Automation, Chinese Academy of Sciences

- **Honourable Mentions:**

- **MITUCLA_HIERARCHY**

- Long Zhu, Yuanhao Chen, William Freeman, Alan Yuille, Antonio Torralba
MIT, UCLA

- **NUS_HOGLBP_CTX_CLS_RESCORE_V2**

- Zheng Song, Qiang Chen, Shuicheng Yan
National University of Singapore

- **UVA_GROUPLOC/UVA_DETMONKEY**

- Jasper Uijlings, Koen van de Sande, Theo Gevers, Arnold Smeulders, Remko Scha
University of Amsterdam

“Lifetime Achievement” Prize



- **UoC/TTI Chicago**

Pedro Felzenszwalb¹, Ross Girshick¹,
David McAllester², Deva Ramanan³

¹*University of Chicago*

²*Toyota Technological Institute at Chicago*

³*University of California, Irvine*

- First submitted to the PASCAL VOC challenge in 2007, this detector has now become a core component of many classification, segmentation, person layout and action classification submissions
- We honour the contribution made to the community by the innovation and success of the method and its free distribution