

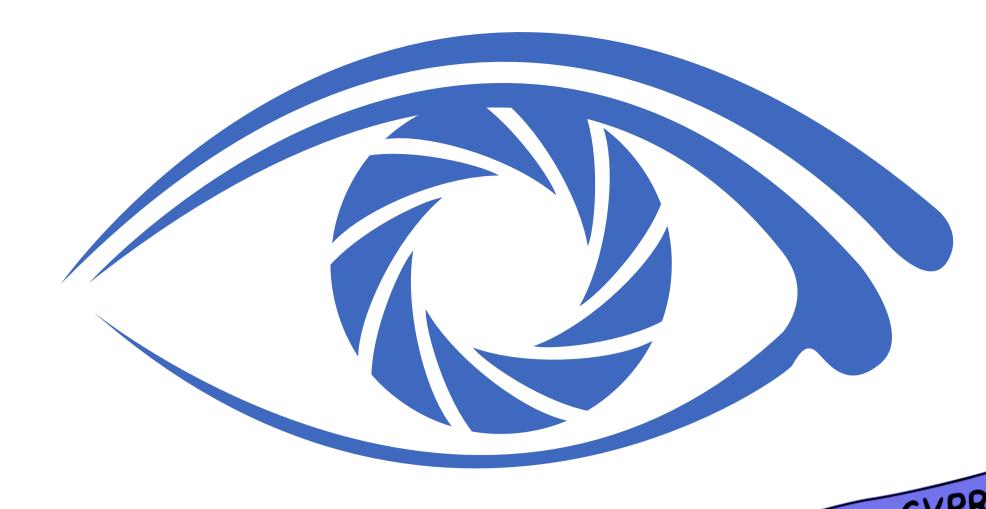


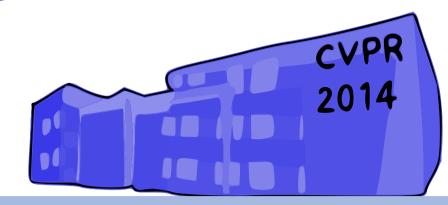
Nonparametric Part Transfer for Fine-grained Recognition

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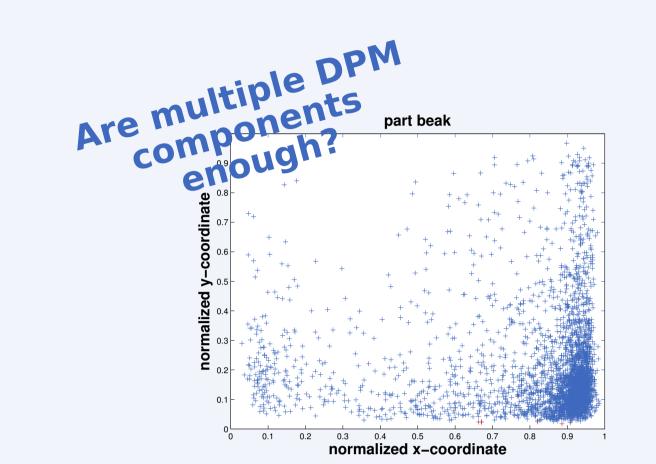
http://www.inf-cv.uni-jena.de

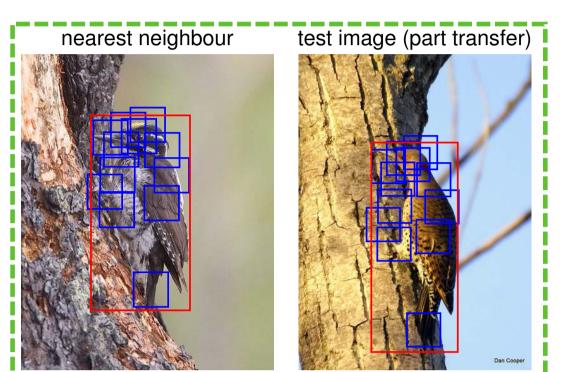


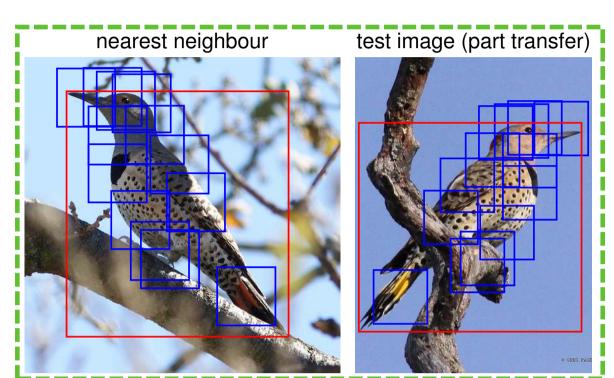


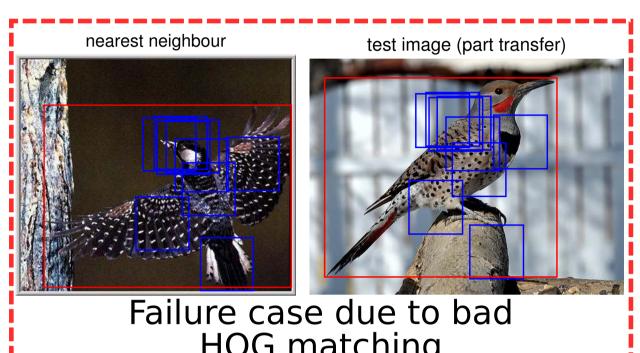
Motivation and method

- Fine-grained recognition: part-based approaches necessary but part locations are difficult to estimate
- Our approach: efficient non-parametric detection with part transfer
- Algorithm to classify a test image
 - 1. Find k nearest neighbours in the training set using HOG to get similar poses
 - 2. Transfer part locations from each of the neighbours (rescaling)
 - 3. Compute k feature representations by using the k part constellations
 - 4. Combine k classifier decisions by averaging



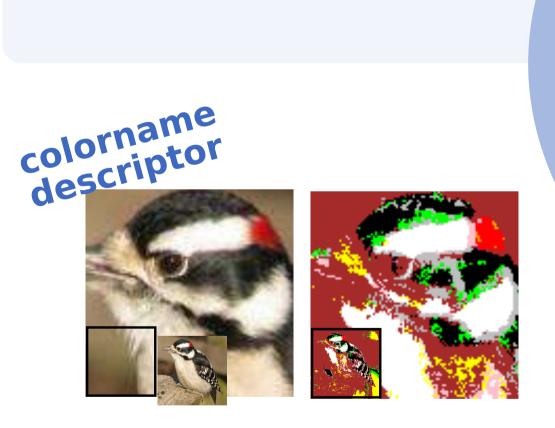






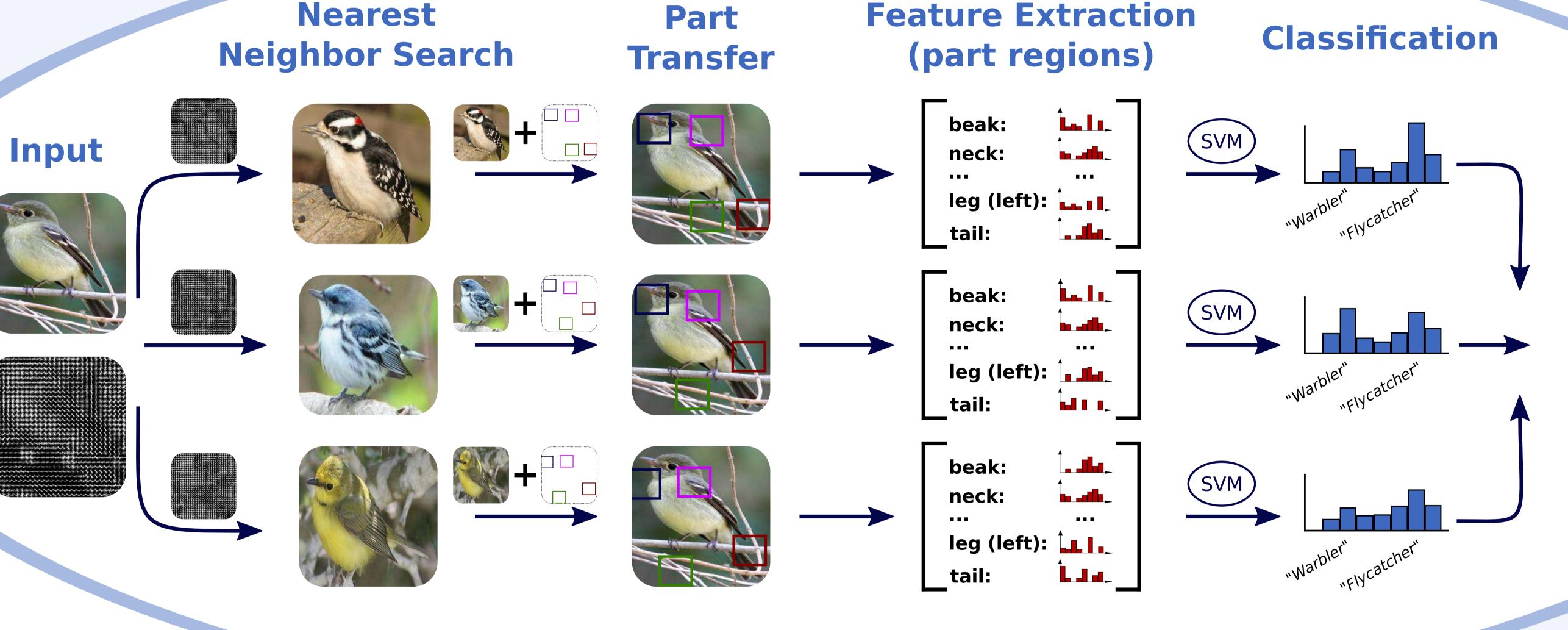
Part detection performance

back (Our app



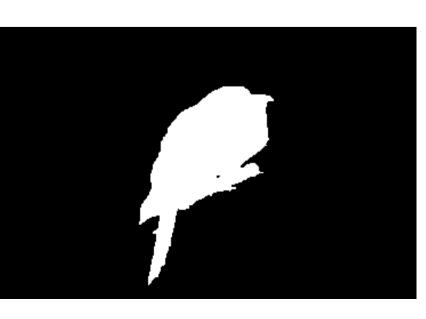
- GrabCut segmentation for masking
- Combination of SIFT and color descriptors (van de Weijer)
- Part-specific bag-of-words codebooks
- Linear SVM for classification with explicit kernel transforms
- Taking care of part constellation symmetry

Method	CUB-2011/14
NN part transfer	69.85%
without GrabCut masking only	68.09%
without part features only	66.08%
without global features only	61.56%

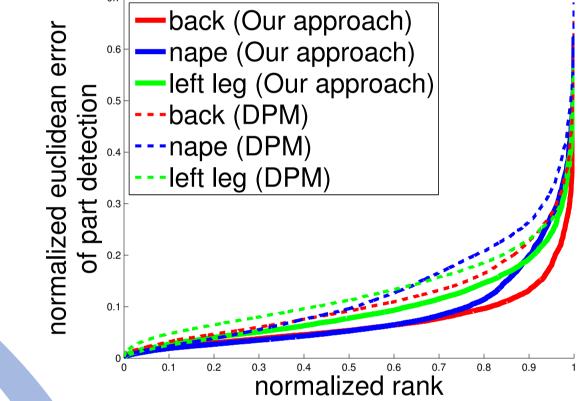








Approach	CUB-2010
Template (Yang, NIPS 2012)	28.20%
Segmentation (Angelova, CVPR 2013)	30.20%
Bubblebank (Deng, CVPR 2013)	32.50%
DPD (Zhang, ICCV 2013)	34.50%
Ours, part transfer	35.94 %





Classification result: "*Flycatcher*"



Approach	CUB-2011/14	CUB-2011/200
PDL (Jia, arXiV 2013)	-	38.91%
Template (Yang, NIPS 2012)	-	43.67%
DPD (Zhang, ICCV 2013)	-	50.98%
POOF (Berg, CVPR 2013)	70.10%	56.78%
Ours, DPM	67.59%	39.18%
Ours, part transfer	69.85%	54.76%
Ours, part transfer		
ensemble with $k = 5$	73.86 %	57.84 %

Classification performance

Implementation details