Supplementary material for submission When Naïve Bayes Nearest Neighbors Meet Convolutional Neural Networks

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1. Supplementary Experiments

Here, additional experimental results are provided both for Scene Recognition and Domain Adaptation.

1.1. Scene Recognition Experiments

Figure 1 contains, from top-left proceeding clockwise, results for NBNN on:

- Scene 15 [2] dataset, sparse sampling and Hybrid [5] features
- Scene 15 [2] dataset, sparse sampling and Imagenet [4] features
- UIUC Sports [3] dataset, sparse sampling and Places [5] features
- UIUC Sports [3] dataset, sparse sampling and Imagenet [4] features

1.2. Domain Adaptation Experiments

Tables 1 to 4 contain our full NBNN results on the Office + Caltech setting [1], both unsupervised and semi-supervised.

References

- [1] B. Gong, Y. Shi, F. Sha, and K. Grauman. Geodesic flow kernel for unsupervised domain adaptation. In *Computer Vision and Pattern Recognition (CVPR)*, *IEEE Conference on*, 2012.
- [2] S. Lazebnik, C. Schmid, and J. Ponce. Beyond bags of features: Spatial pyramid matching for recognizing natural scene categories. In *Computer Vision and Pattern Recognition*, *IEEE Conference on.* 2006.
- [3] L.-J. Li and L. Fei-Fei. What, where and who? classifying events by scene and object recognition. In *Computer Vision* (ICCV), IEEE International Conference on, 2007. 1

Table 1: Source only NBNN - non ReLU

| | | Amazon | Webcam | DSLR | Caltech |
|-----------------|---------|------------------|-------------------|-------------------|-------------------|
| 16px - 1 level | Amazon | 88.23 ± 1.17 | 42.7 ± 7.2 | 38.4 ± 5.8 | 63.12 ± 1.16 |
| | Webcam | 48.67 ± 2.13 | 98.41 ± 1.21 | 88.59 ± 2.9 | 44.29 ± 1 |
| | DSLR | 26.25 ± 0.48 | 75.59 ± 0.84 | 93.75 ± 3.21 | 34.75 ± 0.68 |
| | Caltech | 78.5 ± 2.29 | 44.64 ± 4.8 | 41.65 ± 11.19 | 72.38 ± 1.37 |
| 16px - 2 levels | Amazon | 87.67 ± 1.17 | 26.8 ± 7.85 | 35.09 ± 5.07 | 62.68 ± 1.49 |
| | Webcam | 48.14 ± 1.8 | 97.44 ± 1.78 | 88.9 ± 3.1 | 44.22 ± 1.22 |
| | DSLR | 34.9 ± 1.81 | 71.96 ± 0.99 | 92.18 ± 2.32 | 42.9 ± 0.749 |
| | Caltech | 74.05 ± 1.9 | 38.06 ± 5.6 | 41.84 ± 7.5 | 72.09 ± 1.54 |
| 16px - 3 levels | Amazon | 89.43 ± 1.12 | 50.5 ± 1.81 | 58.91 ± 4.24 | 70.12 ± 1.61 |
| | Webcam | 52.48 ± 2.69 | 97.79 ± 1.89 | 96.17 ± 1.82 | 52.55 ± 1.55 |
| | DSLR | 60.76 ± 1.5 | 90.64 ± 0.62 | 94.37 ± 1.5 | 56.1 ± 0.91 |
| | Caltech | 75.9 ± 2.08 | 46.33 ± 4.64 | 57.89 ± 6.27 | 78.29 ± 1.45 |
| 32 px - 1 level | Amazon | 87.44 ± 1.01 | 34.033 ± 5.68 | 35.15 ± 5.27 | 63.74 ± 1.26 |
| | Webcam | 50.63 ± 1.15 | 98.27 ± 1.78 | 89.93 ± 1.52 | 45.19 ± 1.38 |
| | DSLR | 29.85 ± 0.89 | 73.49 ± 1.27 | 92.34 ± 2.7 | 39.11 ± 0.7 |
| | Caltech | 75.78 ± 2.52 | 38.71 ± 6.77 | 41.97 ± 7.99 | 72.29 ± 1.87 |
| 32px - 2 levels | Amazon | 88.98 ± 0.89 | 48.2 ± 3.21 | 55.66 ± 5.37 | 70.14 ± 1.67 |
| | Webcam | 54.06 ± 1.92 | 98.2 ± 1.08 | 95.79 ± 1.09 | 51.7 ± 1.8 |
| | DSLR | 43.69 ± 1.19 | 87.01 ± 0.76 | 95 ± 2.05 | 50.5 ± 0.78 |
| | Caltech | 74.79 ± 1.84 | 48.5 ± 6.38 | 59.04 ± 9.26 | 76.92 ± 1.9 |
| 32px - 3 levels | Amazon | 89.85 ± 0.64 | 59.89 ± 2.55 | 68.28 ± 4.95 | 75.76 ± 1.33 |
| | Webcam | 66.53 ± 1.33 | 98.48 ± 1.29 | 98.21 ± 0.98 | 62.53 ± 1.25 |
| | DSLR | 67.27 ± 1.14 | 94 ± 0.57 | 97.18 ± 1.77 | 65.35 ± 0.66 |
| | Caltech | 78.81 ± 0.84 | 61.25 ± 4.4 | 66.75 ± 6.26 | 80.5 ± 1.42 |
| 64px - 1 level | Amazon | 89.03 ± 0.92 | 49.83 ± 3.84 | 56.36 ± 5.3 | 69.1 ± 1.59 |
| | Webcam | 57.33 ± 1.98 | 98.55 ± 1.54 | 96.3 ± 1.26 | 51.42 ± 1.57 |
| | DSLR | 42.56 ± 0.87 | 84.88 ± 0.89 | 92.81 ± 2.91 | 47.7 ± 0.64 |
| | Caltech | 74.37 ± 1.69 | 49.32 ± 4.2 | 60.06 ± 7.2 | 76.37 ± 1.7 |
| 64px - 2 levels | Amazon | 89.84 ± 0.81 | 60.54 ± 3.59 | 68.66 ± 4.95 | 75.33 ± 1.05 |
| | Webcam | 65.55 ± 1.41 | 98.55 ± 1.54 | 98.15 ± 1.14 | 62.57 ± 1.03 |
| | DSLR | 64.46 ± 0.92 | 94.74 ± 0.55 | 95.93 ± 1.67 | 63.57 ± 0.404 |
| | Caltech | 78.04 ± 1.04 | 60.16 ± 5.29 | 66.49 ± 6.17 | 80.23 ± 1.21 |
| 64px - 3 levels | Amazon | 89.76 ± 0.95 | 60.23 ± 3.29 | 68.98 ± 4.54 | 75.2 ± 1.08 |
| | Webcam | 66.87 ± 1.2 | 98.48 ± 1.48 | 98.53 ± 0.73 | 63.33 ± 1.2 |
| | DSLR | 67.4 ± 0.94 | 93.93 ± 0.74 | 97.18 ± 2.05 | 65.22 ± 0.64 |
| | Caltech | 79.03 ± 0.93 | 61.28 ± 5.53 | 67.83 ± 2.0 | 80.46 ± 1.35 |

- [4] O. Russakovsky, J. Deng, H. Su, J. Krause, S. Satheesh, S. Ma, Z. Huang, A. Karpathy, A. Khosla, M. Bernstein, A. C. Berg, and L. Fei-Fei. Imagenet large scale visual recognition challenge. *International Journal of Computer Vision*, pages 1–42, 2015. 1
- [5] B. Zhou, A. Lapedriza, J. Xiao, A. Torralba, and A. Oliva. Learning deep features for scene recognition using places database. In *Advances in Neural Information Processing Sys*tems, NIPS, 2014. 1

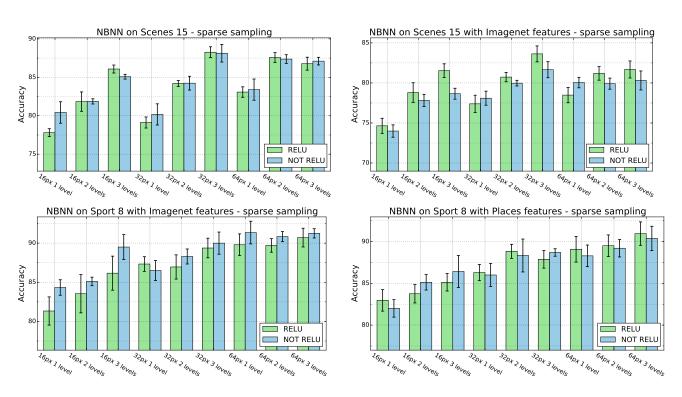


Figure 1: Results obtained by NBNN using CNN activations, based on different networks.