## VisDA Classification Challenge:

## Runner-Up Talk

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## VisDA Classification Challenge

Our Approach

- Assembly of Shallow Domain Adaptation Networks (SDAN)
- We extract deep features
- Off-the-shelf models pre-trained on ImageNet
- Inception ResNet v2, Inception v3 and v4, ResNet 152

- We test two groups of methods
- Discrepancy-based SDANs and their multiple variants
- Adversarial Learning SDANs


## Discrepancy-based SDANs

- Max Mean Discrepancy (MMD) with multiple kernels [Gretton'12]

- MMD in DAN for learning transferable features [Long'15]



## SDAN Variants

- Residual SDAN: $W_{t}=W_{s}+W_{n}$, where $W_{n} \sim N\left(0, \frac{2}{n_{d}}\right)$
- Random Kernels: in any batch, 5 MMD kernel bandwidths are drawn from a Gamma distribution
- Adversarial SDAN: non-shared version
- Define $D_{p}$ be a projection (analog to Discriminator) in the kernel space

- For the loss $L_{M M D}$, alternate two optimization steps:
- $\min _{W_{s}, W_{t}} L_{M M D}$ and $\max _{D_{p}} L_{M M D}$


## Adversarial Learning

## Extend InfoGAN to Domain Adaptation

- Source/Target Mappers, Classifier and Discriminator like in ADDA [Tzeng'17]
- Latent variables $c$ : class labels for Source, random for Target
- Shared layers of Discriminator and $Q$-function



## Assembling approach

- Average the softmax values or by the majority voting
- Combine methods with a good expected accuracy and important variability
- Different deep features
- Shared weights and residual SDANs
- MMD kernels with fixed or random kernel bandwidths
- Subsets of training data to adapt SDANs
- Different $D$ and $Q$ layer sharing in adversarial SDANs


## Averaging: 2 levels

- Averaging SDAN versions with different parameters yields $\sim 3 \%$
- ResNet features:

| SDAN | Resid | RndKrnls | Resid+RndKrnls | $\ldots$ | All |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 77.6 | 77.95 | 79.1 | 77.55 | $\ldots$ | $\mathbf{8 1 . 6}$ |

- Averaging of 'Averaged' SDANs by features yields again $\sim 3 \%$

| ResNet | Incep-ResNet | InceptV3 | InceptV4 | All |
| :---: | :---: | :---: | :---: | :---: |
| 81.6 | 84.9 | 84.1 | 84.1 | $\mathbf{8 7 . 4}$ |


| Leaderboard: ImageNet Pretraining |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| \# | User | Team Name | Per Category Accuracy |  |  |  |  |  |  |  |  |  |  |  | MeanAcc |
|  |  |  | plane | beycl | bus <br> A | car $\boldsymbol{A}$ | horse | horse A | mcycl | person A | plant <br> A | sktbd | train | truck |  |
| 1 | GF_ColourLab_UEA |  | 96.9 | 92.4 | 92.0 | 97.2 | 95.2 | 98.8 | 86.3 | 75.3 | 97.7 | 93.3 | 94.5 | 93.3 | 92.8 |
| 2 | NLE_DA |  | 94.3 | 86.5 | 86.9 | 95.1 | 91.1 | 90.0 | 82.1 | 77.9 | 96.4 | 77.2 | 86.6 | 88.0 | 87.7 |
| 3 | BUPT_OVERFIT | BUPT_OVERFIT | 95.7 | 67.0 | 93.4 | 97.2 | 90.6 | 86.9 | 92.0 | 74.2 | 96.3 | 66.9 | 95.2 | 69.2 | 85.4 |

