Self-supervised learning in the 'wild'

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Task

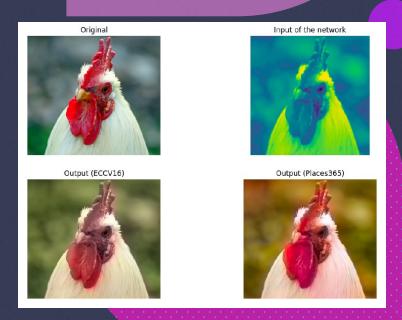
- Investigate various open-source self-supervised learning methods
- Select one or two classification datasets unrelated to ImageNet
- Utilize the selected SSL methods to pretrain models on your chosen datasets.
- Conduct a linear benchmark evaluation using a portion of ImageNet
- Compare the performance of your pretrained models against those publicly available and trained on ImageNet

Concept

- Two different methods:
 - Colorization
 - Jigsaw
- Various training techniques and hyperparameter settings
- Train a baseline model (VISSL framework [4])

Colorization

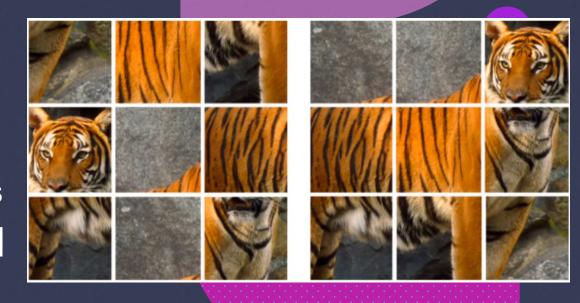
- Colorful Image Colorization [1]
- Places365 dataset [2]
- Resolution:128x128
- 35 hours of training, 10 epochs
- Split the model into backend and head, frozen the weights and merged with the ResNet50





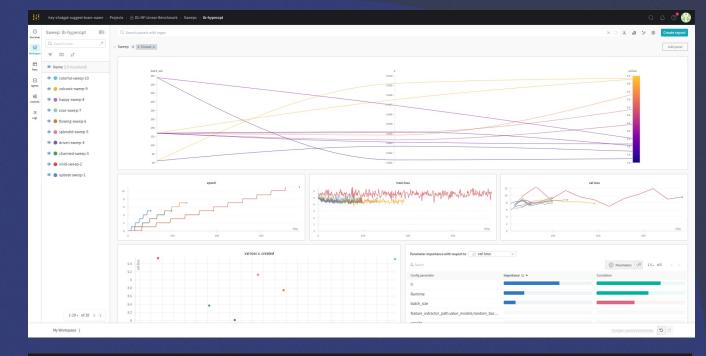
Jigsaw - 1

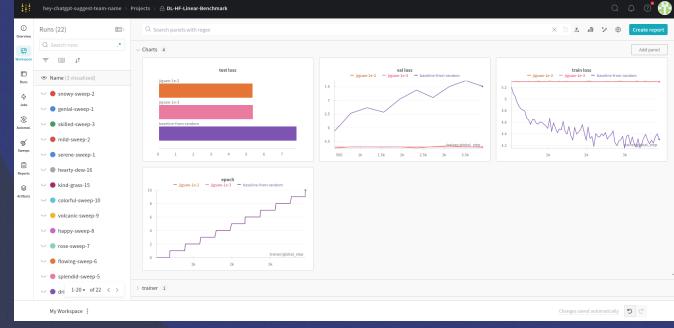
- Dividing into 3x3 tiles, random permutations
- Pretraining on non-ImageNet: Places365 [2]
- Linear benchmark evaluation:
 - Loading and freezing the pretrained weights
 - Fully connected classification layer
 - Tiny ImageNet dataset, 200 classes
- Initial experiments: VISSL (unsuccessful)



Jigsaw - 2

- Our own implementation loosely based on [5]:
 - Convolutional backbone for each tile
 - Fully connected layer per-tile
 - Global fully connected layer
 - Output: one-hot representation of a permutation
 - Additions: BN, poolings etc.
- Customizable: # of conv. filters, layer width
- Customizable: # of possible permutations, tiles (2x2 or 3x3)
- Implemented linear benchmark evaluation
- Torch, Lightning, WandB, hyperopt with Sweeps
- Heavy technical constraints due to HW limitations POC





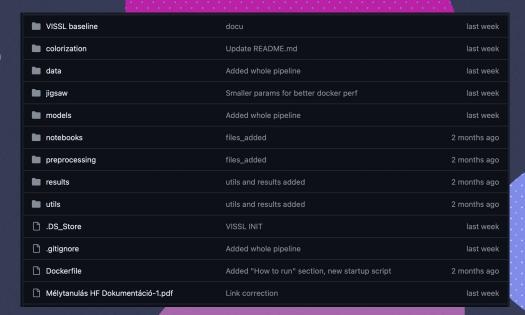
VISSL Baseline

- Facebook AI Research
- Model ZOO
- JIGSAW problem
- Configfile based setting
- Error during training

```
config:
VERBOSE: True
LOG_FREQUENCY: 200
TEST_ONLY: False
TEST_EVERY_NUM_EPOCH: 1
TEST_MODEL: True
SEED VALUE: 1
MULTI PROCESSING METHOD: forkserver
HOOKS:
  PERF_STATS:
    MONITOR_PERF_STATS: True
DATA:
  NUM DATALOADER WORKERS: 5
   TRAIN:
     DATA SOURCES: [data]
     LABEL_SOURCES: [data/labels]
     DATASET_NAMES: [imagenet1k_folder]
     BATCHSIZE_PER_REPLICA: 32
     TRANSFORMS:
      - name: RandomResizedCrop
         size: 224
       - name: RandomHorizontalFlip
       - name: ToTensor
      - name: Normalize
         mean: [0.485, 0.456, 0.406]
         std: [0.229, 0.224, 0.225]
    MMAP_MODE: True
     COPY_TO_LOCAL_DISK: False
     COPY_DESTINATION_DIR: /tmp/imagenet1k/
   TEST:
    DATA_SOURCES: [data]
     LABEL_SOURCES: [data/labels]
```

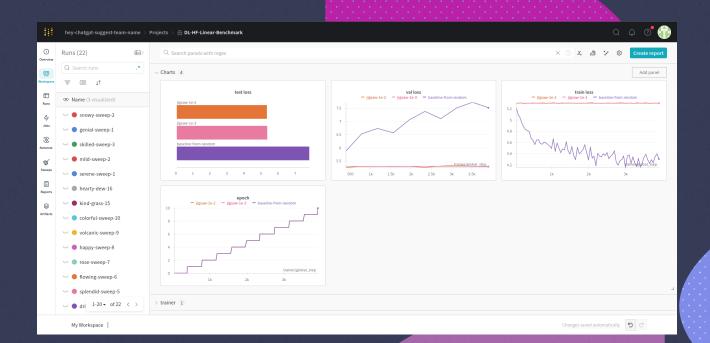
Structure of the project

- Each project in different folders (Colorization, JIGSAW, VISSL)
 - Seperate:
 - Dataloading and pre-processing
 - Training
 - Evaluation
 - Environement (Docker)
- Some global features are in the root



Results

- Descending loss
- Proof of concept pipeline
- Trained models available for inference
- Containerization, reproducibility



Summary

- Gained experience in SSL
- Two different type of frameworks
- Future possibilities:
 - Integrating both methods
 - Concatenate both feature extractors: better "ensemble" features

Refrences

- 1. Zhang, R., Isola, P. and Efros, A.A., 2016. Colorful Image Colorization. Available at: arXiv:1603.08511
- 2. http://places.csail.mit.edu
- 3. Goyal, P., Mahajan, D., Gupta, A. and Misra, I., 2019. Scaling and Benchmarking Self-Supervised Visual Representation Learning.: *Available at: <https://arxiv.org/pdf/1905.01235.pdf>*
- 4. VISSL Team, n.d. VISSL: A library for state-of-the-art self-supervised learning from images. Available at: https://vissl.ai
- 5. Mehdi Noroozi, Paolo Favaro: Unsupervised Learning of Visual Representations by Solving Jigsaw Puzzles. Available at: https://arxiv.org/abs/1603.09246v3