

AMMAI Final

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Results

Models	minilImageNet	EuroSAT	ISIC
Baseline	67.87 \pm 0.71% 68.10% \pm 0.67%	/ 76.16 \pm 0.72% 75.69% \pm 0.66% / 79.08% \pm 0.61%	/ 48.22 \pm 0.61% 43.56% \pm 0.60% / 48.11% \pm 0.64%
Protonet	68.42 \pm 0.63% 66.33% \pm 0.65%	/ 81.79 \pm 0.62% 77.45% \pm 0.56% / 81.45% \pm 0.63%	/ 46.12 \pm 0.59% 41.73% \pm 0.56% / 46.72% \pm 0.59%
Compact	61.51 \pm 0.66%	59.94 \pm 0.81% / 79.56 \pm 0.64%	41.76 \pm 0.64% / 45.90 \pm 0.61%
Sphere	58.04 \pm 0.67%	67.76 \pm 0.71% / 59.90 \pm 0.82%	37.68 \pm 0.54% / 42.01 \pm 0.61%
Feature Generation	66.54 \pm 0.68%	/ 80.78 \pm 0.58%	/ 46.17 \pm 0.59%

*number in blue is the result in README of sample code

Protonet + Compact Loss

- Protonet + additional loss
- Each support image is mapped as close to their mean (prototype).
- Not so effective.

Models	miniImageNet	EuroSAT	ISIC
Baseline	$67.87 \pm 0.71\%$	$/ 76.16 \pm 0.72\%$	$/ 48.22 \pm 0.61\%$
Protonet	$68.42 \pm 0.63\%$	$/ 81.79 \pm 0.62\%$	$/ 46.12 \pm 0.59\%$
Compact	$61.51 \pm 0.66\%$	$59.94 \pm 0.81\% / 79.56 \pm 0.64\%$	$41.76 \pm 0.64\% / 45.90 \pm 0.61\%$

Large Margin on Sphere

- Each image is embedded on a 512-dim sphere.
- Distances are measured using cosine.
- Training with margin proves hard, not applicable in protonet-like training.

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Protonet	$68.42 \pm 0.63\%$	/ $81.79 \pm 0.62\%$	/ $46.12 \pm 0.59\%$
Sphere	$58.04 \pm 0.67\%$	$67.76 \pm 0.71\%$ / $59.90 \pm 0.82\%$	$37.68 \pm 0.54\%$ / $42.01 \pm 0.61\%$

Feature Generation

- Implementation of “Diversity Transfer Network for Few-Shot Learning” on AACL 2019
 - Chen, Mengting, et al. "Diversity Transfer Network for Few-Shot Learning." *arXiv preprint arXiv:1912.13182* (2019).
- Source code released by Authors
 - <https://github.com/Yuxin-CV/DTN>.
- Source code used on the final project
 - <https://github.com/shiannn/AMMAI-Final-FG>
 - Vectorize computing classifiers from support samples
 - Extra dataloader for referenced images

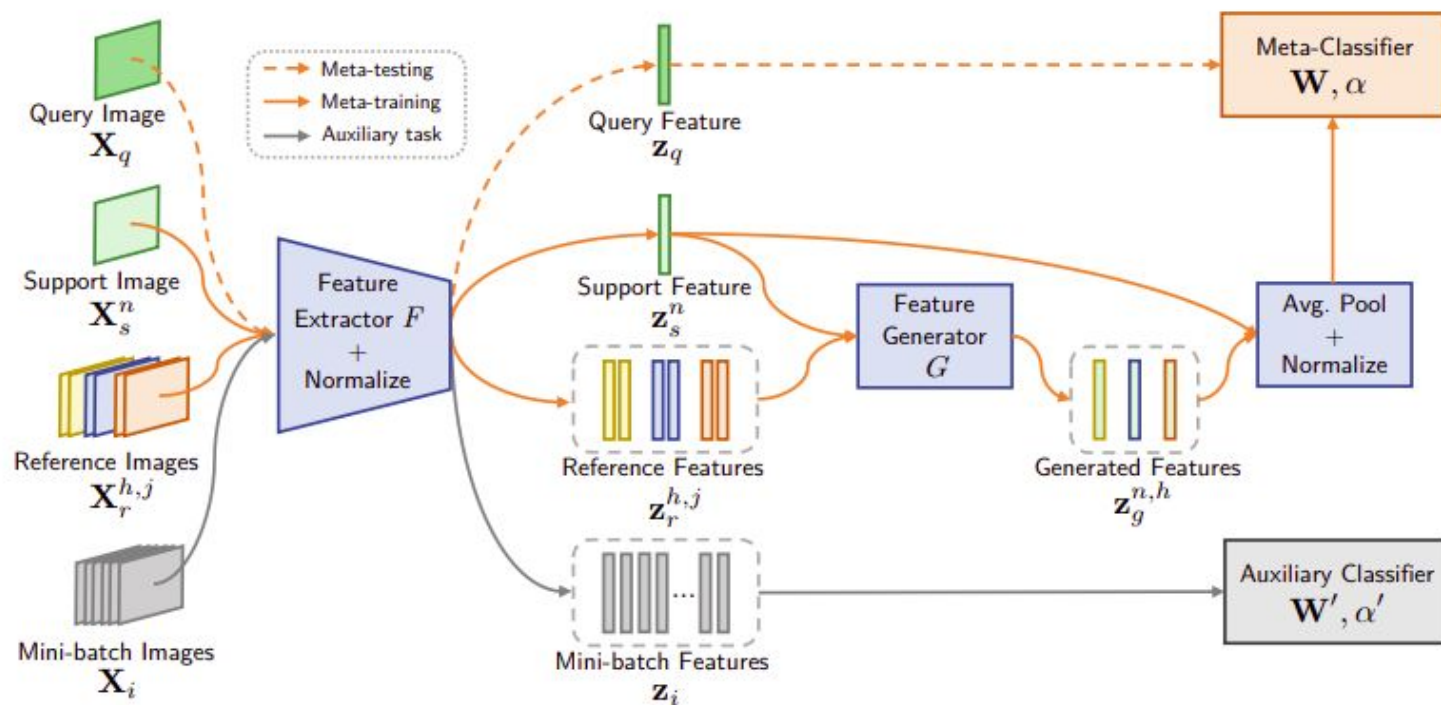
Feature Generation

- Using random seed
 - `np.random.seed(10)` & `torch.manual_seed(20)`
- For each support/query sample, we generate 6 reference pairs
- Little bits exceed the result of Protonet on miniimageNet of README
- Little bits exceed the result of Protonet on ISIC of re-run
- Exceed the result of Baseline on EuroSAT with $\sim 1.7\%$

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Baseline	$67.87 \pm 0.71\%$ $68.10\% \pm 0.67\%$	$/ 76.16 \pm 0.72\%$ $75.69\% \pm 0.66\% / 79.08\% \pm 0.61\%$	$/ 48.22 \pm 0.61\%$ $43.56\% \pm 0.60\% / 48.11\% \pm 0.64\%$
Protonet	$68.42 \pm 0.63\%$ $66.33\% \pm 0.65\%$	$/ 81.79 \pm 0.62\%$ $77.45\% \pm 0.56\% / 81.45\% \pm 0.63\%$	$/ 46.12 \pm 0.59\%$ $41.73\% \pm 0.56\% / 46.72\% \pm 0.59\%$
Feature Generation	$66.54 \pm 0.68\%$	$/ 80.78 \pm 0.58\%$	$/ 46.17 \pm 0.59\%$

*number in blue is the result in README of sample code

Feature Generation - via Diversity Transfer

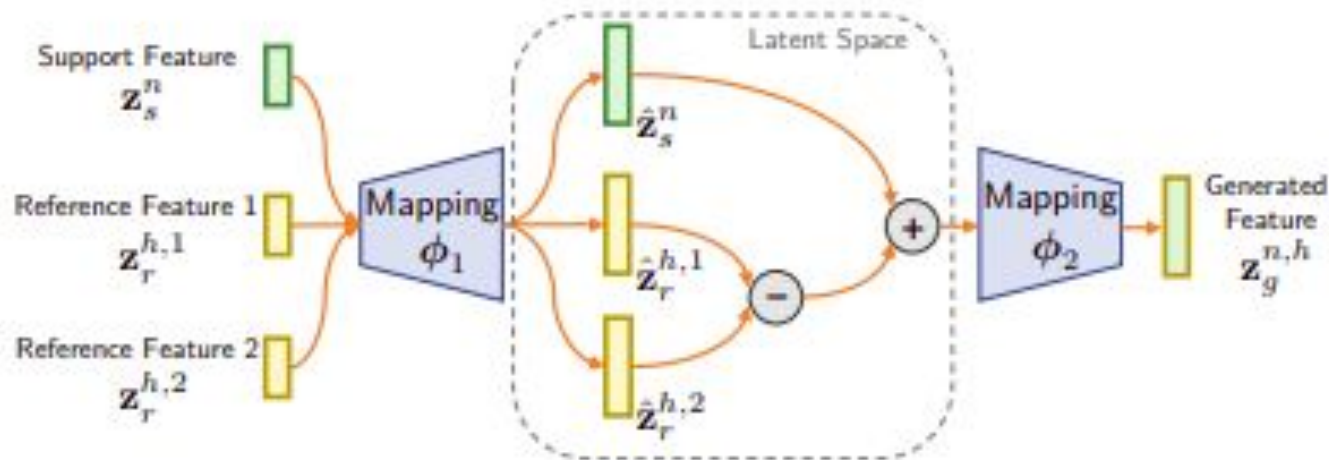


Feature Generation - via Diversity Transfer

- Two dataloaders
 - One for sampling support & query samples
 - Another for sampling referenced image pairs of same classes
- Feature Extractor
 - Using backbone as sample code (ResNet10) to extract 512 dimension embeddings
- Feature Generator
 - Get diversity from image pairs of other class and add to the current support sample to generate extra feature
 - Blend generated features with original support features to get center of each class in classifier
- Classifier
 - Calculate cosine similarity of each query sample with class centers
 - Fix with a trainable coefficient

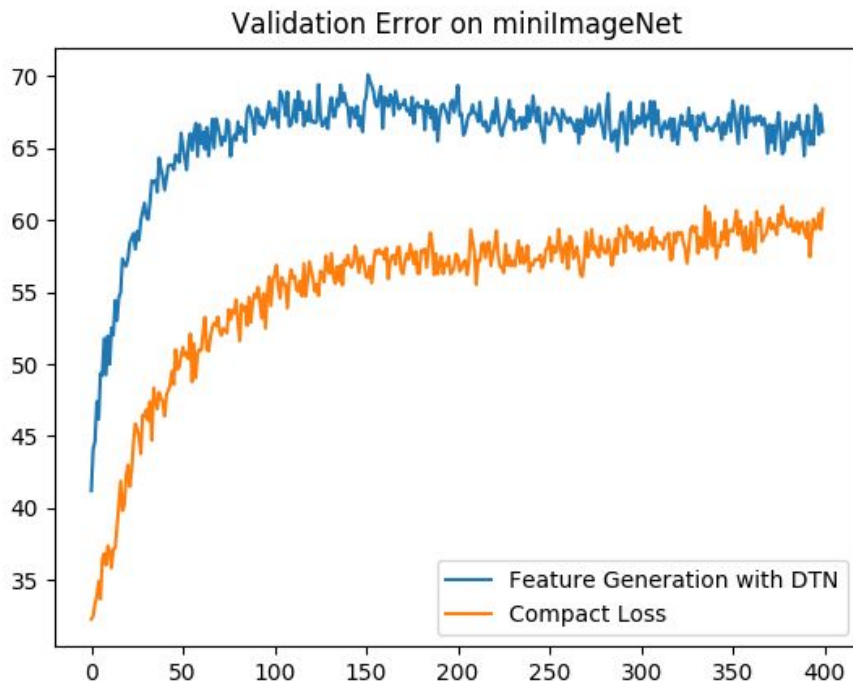
Feature Generator

- Three Input features are mapped into latent space
- Diversity (offset) is calculated in latent space and added to original support feature
 - To some degree, like data augmentation
- Mapping back to the same size as input, the result is supposed to a sample belonging to the same category as input support feature



Validation Error Observation

- Fast converge
 - Test Acc = 70.12% \pm 0.91% on 151 epoch
- Some Overfitting issue on Feature Generation



Future Work

- Try different number of generated pairs and compare the result
- Try Auxiliary Task Co-training

Q&A