# DLCV 2021 Final Project Challenge 3:

# Fine-grained long-tailed food image classification

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## **Problem**

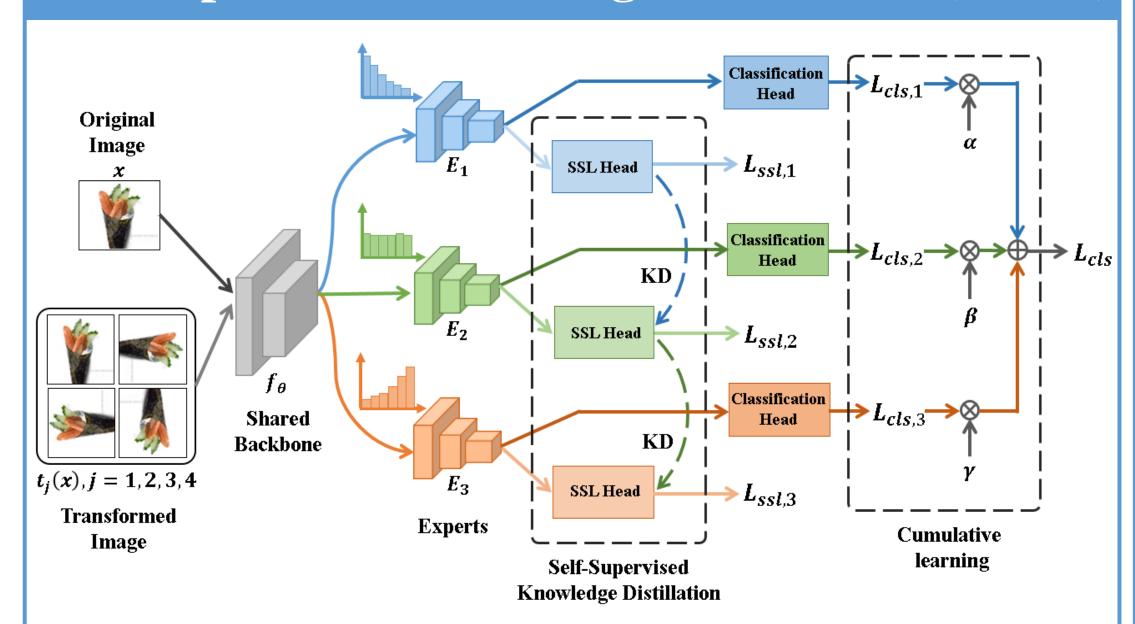
- The original TADE [1] suffers the degradation of head's accuracy due to the direct average of the outputs from different experts.
- Tail class experts tend to overfit on small classes, which is harmful to the representation ability of the backbone.

[1] "Test-agnostic long-tailed recognition by test-time aggregating diverse experts with self-supervision." arXiv (2021).

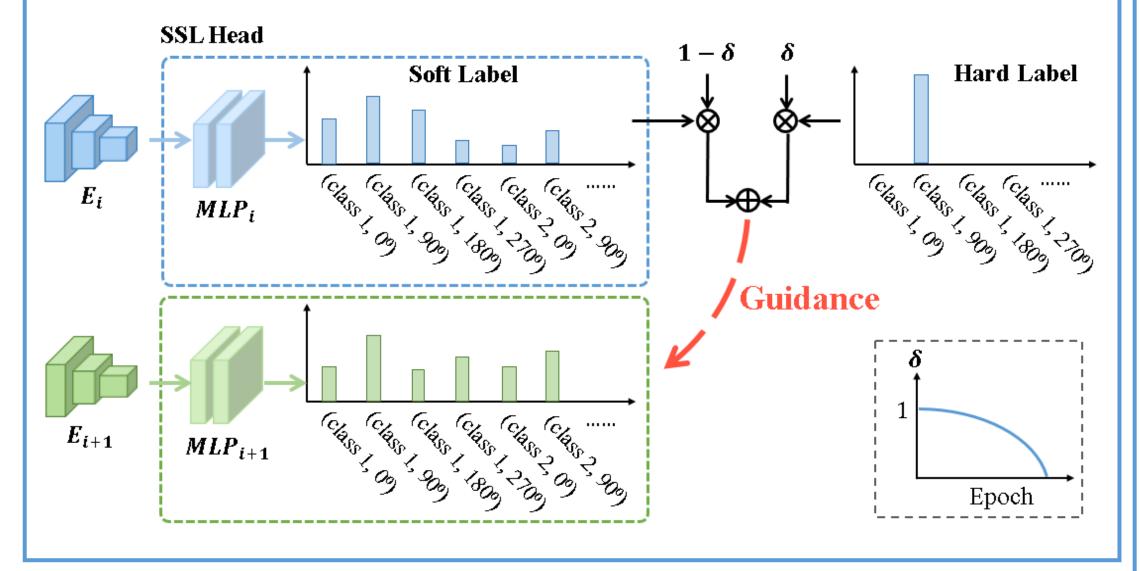
### Contribution

- Bring the *Self-Supervised Knowledge Distillation (SSKD)* into the long-tailed problem to have a better representation learning.
- Cumulative learning (CL) is adopted to avoid damaging the learned universal features when emphasizing the tail classes.
- Propose *Image-wise Test-time Aggregation (ITA)* to learn the aggregation weights of three experts image-wisely in test-time.

## Self-Supervised Knowledge Distillation (SSKD)

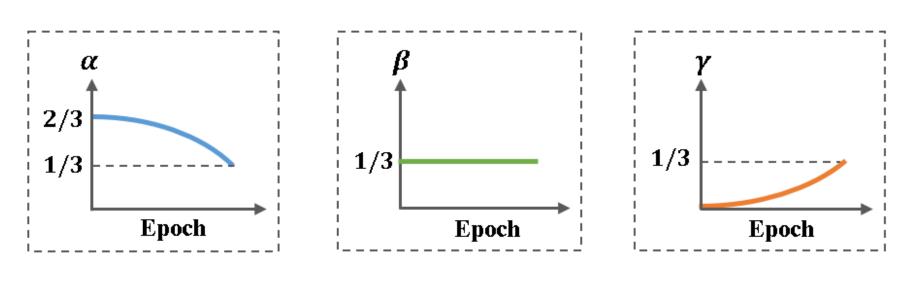


- Utilize self-supervised learning as an auxiliary task (SSL Head) for each expert to mine the dark knowledge.
- Transfer the representational knowledge of head class experts into tail class experts by knowledge distillation (KD).
- Adopt the SSL method of jointly learning the original classification task and self-supervised auxiliary task



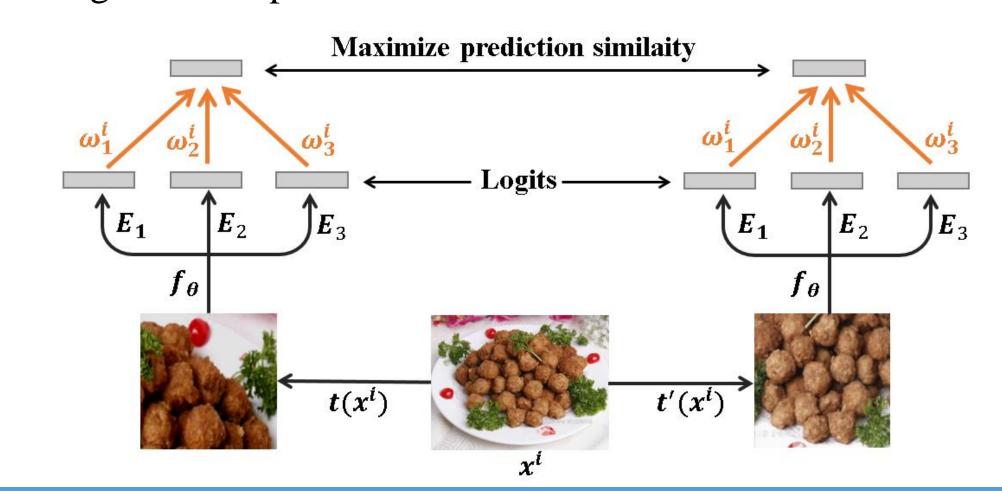
# **Cumulative Learning (CL)**

• Adjust the weights of loss function from different experts in a way that model learns the universal representations first, then transition to focus on the tail classes.



## Image-wise Test-time Aggregation (ITA)

- Propose that testing images have to be assigned their own aggregation weights of three experts.
- Assign larger aggregation weight for the expert such that it can produce similar predictions for two different augmented inputs.



# Experiment

#### Comparison with SOTA

|      | All            | Frequent       | Common        | Rare           |
|------|----------------|----------------|---------------|----------------|
| TADE | 0.745          | 0.754          | 0.762         | 0.582          |
| Ours | 0.792 (+0.047) | 0.821 (+0.067) | 0.792 (+0.03) | 0.587 (+0.005) |

#### Ablation Study

|            | All            | Frequent       | Common         | Rare           |
|------------|----------------|----------------|----------------|----------------|
| TADE       | 0.745          | 0.754          | 0.762          | 0.582          |
| +SSKD      | 0.780 (+0.035) | 0.799 (+0.045) | 0.792 (+0.03)  | 0.587 (+0.005) |
| +CL        | 0.790 (+0.01)  | 0.819 (+0.02)  | 0.791 (-0.001) | 0.588 (+0.001) |
| +Test-time | 0.792 (+0.002) | 0.821 (+0.002) | 0.792 (+0.001) | 0.587 (-0.001) |

### Effectiveness of SSKD

|      | All            | Frequent       | Common         | Rare           |
|------|----------------|----------------|----------------|----------------|
| TADE | 0.745          | 0.754          | 0.762          | 0.582          |
| SSL  | 0.771 (+0.026) | 0.783 (+0.029) | 0.791 (+0.029) | 0.572 (-0.01)  |
| SSKD | 0.780 (+0.035) | 0.799 (+0.045) | 0.792 (+0.03)  | 0.587 (+0.005) |

#### Choices of KD

|                | All            | Frequent       | Common         | Rare  |
|----------------|----------------|----------------|----------------|-------|
| From $E_1$     | 0.769          | 0.780          | 0.789          | 0.588 |
| From $E_{i-1}$ | 0.790 (+0.021) | 0.819 (+0.039) | 0.791 (+0.002) | 0.588 |