

Few-Class Arena: A Benchmark for Efficient Selection of Vision Models and Dataset Difficulty Measurement

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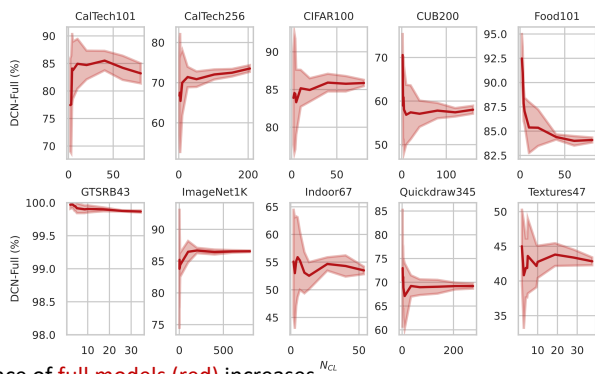
Goal

Few-Class Arena (FCA) is a benchmark tool designed to facilitate the selection of efficient neural network and research in the **few-class regime**, i.e., **number of classes < 10** ($N_{CL} < 10$).

- Propose a **one-time** sub-class dataset difficulty measurement.
- Unveil new insights in the **few-class regime**.

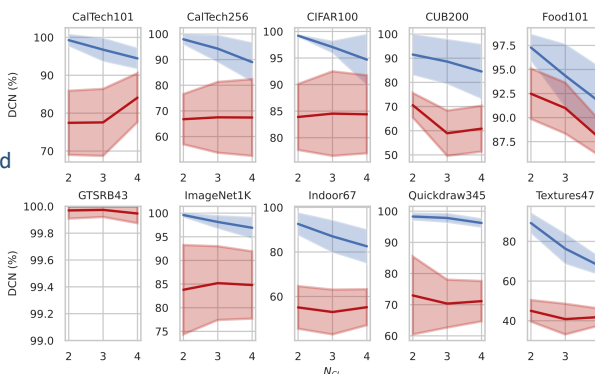
Motivation

Limitations
of Existing
many-class
Benchmarks
(red)



- The variance of **full models** (red) increases.
- This large variance indicates high uncertainty in efficient model selection.

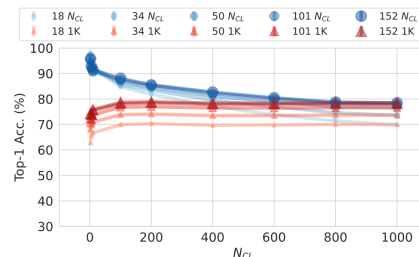
Our proposed
Few-Class
Arena (FCA)
(blue)



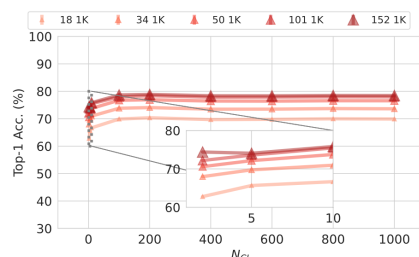
- The variance of **sub-models** (blue) decreases.
- Sub-model** accuracy increases.
- Lack of few-class benchmarks designed for sub-models.**

DCN: dataset classification difficulty number (DCN) is the empirically highest accuracy achieved in a dataset. Acc.: Accuracy. Var.: Variance. Corr.: correlation.

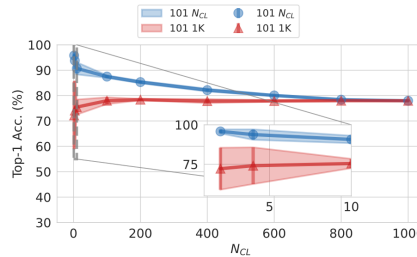
Higher Acc. & Lower Var. of Sub-models



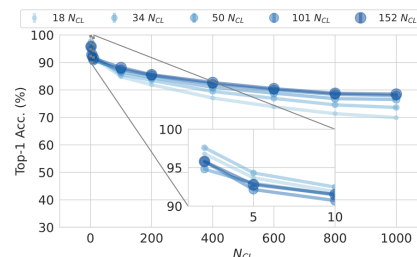
Sub-models attain higher upper-bound accuracy than full models.



Full models follow the scaling law in model size.



Full model accuracy range widens. Sub-models narrow the accuracy range.



The scaling law is violated for sub-models.

Proposed Similarity-Based Silhouette Score (SimSS)

$$SimSS^{(D)} = \frac{1}{|L| \times |C_l|} \sum_{i \in C_l} \frac{S_{\alpha}(i) - S'_{\beta}(i)}{\max(S_{\alpha}(i), S'_{\beta}(i))}$$

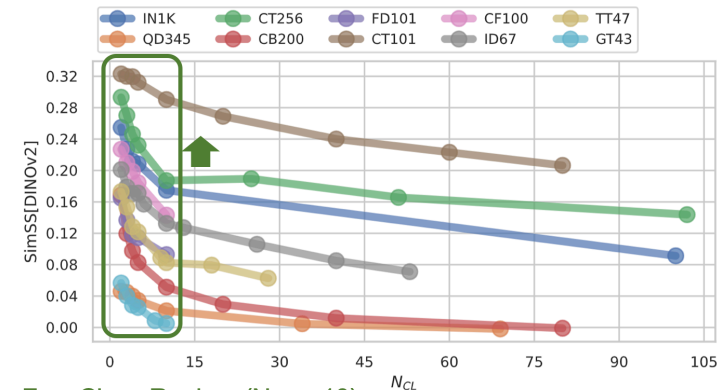
SimSS quantifies the cluster-like characteristics of a dataset to facilitate efficient vision model design and selection.

S_{α} : Intra-Class Similarity represents the **tightness** of the class

S'_{β} : Nearest Inter-Class Similarity represents the **closeness** to other classes.

MT: Model type. N_{CL} : Number of classes. Det: Object Detection. Seg: Segmentation. F (Full model): models pre-trained on the full many classes. S (Sub-model): models trained on the sub classes ($N_{CL} < 10$).

SimSS in the Few-Class Regime

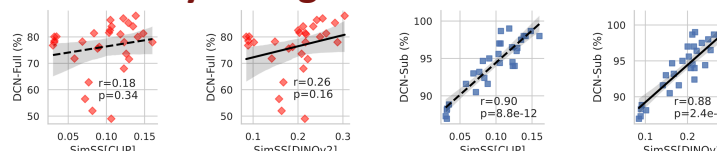


Few-Class Regime ($N_{CL} < 10$)
The importance of image similarity increases.

Extension to Detection & Segmentation

MT	N_{CL}	Det mAP@50↑	Det STDEV↓	Seg mAP@50↑	Seg STDEV↓
F	80	0.405	0.195	0.378	0.200
F	5	0.456	0.090	0.435	0.098
FT	5	0.488	0.069	0.465	0.082
S	5	0.503	0.069	0.474	0.084
F	2	0.488	0.161	0.475	0.180
FT	2	0.505	0.127	0.457	0.159
S	2	0.538	0.106	0.482	0.152

Summary — High Corr. SimSS & Acc.



Our Similarity-based score (SimSS) displays **high correlation** ($r \geq 0.88$) with model accuracy for the purpose of a practitioner quickly comparing models.