

# Can Finetuning Overcome Racial Bias In Face Recognition Models?

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## MOTIVATION

Most face datasets are **racially unbalanced**.

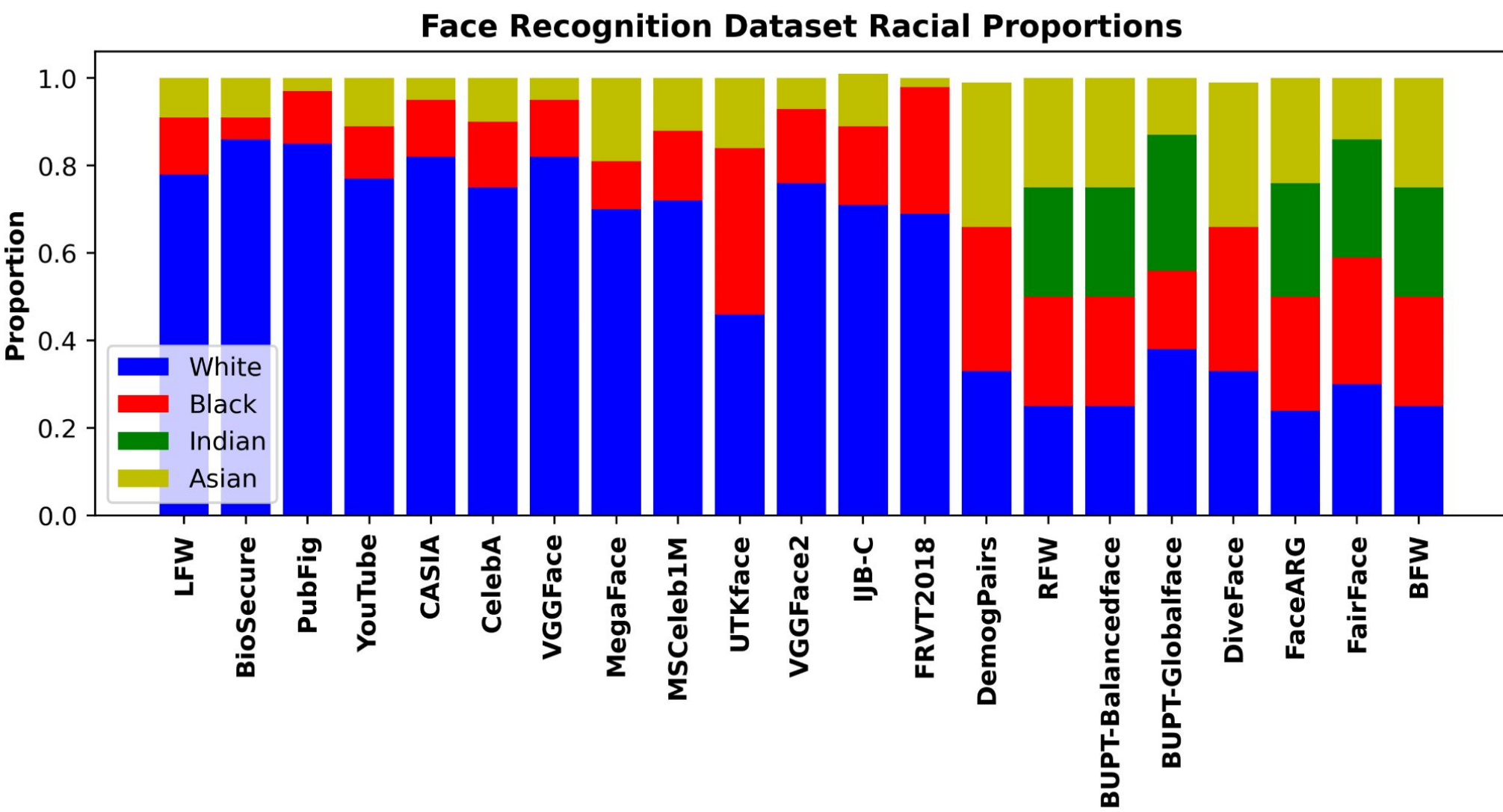


Figure 1. Dataset Racial Distributions in Popular Datasets [1]

Face recognition models are **racially-biased**, even with **balanced training data** (Table 1).

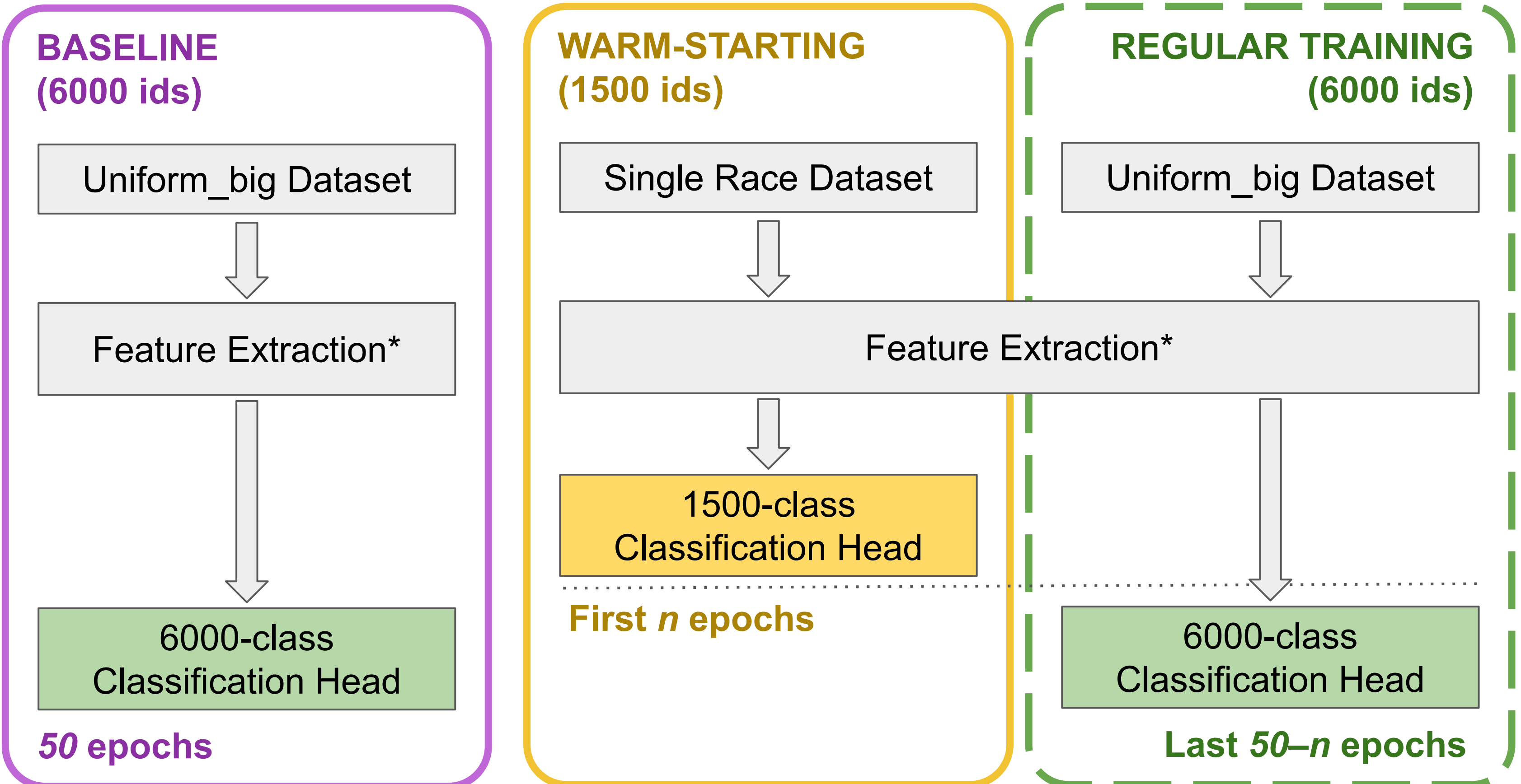
## OVERVIEW

- **Critical period**: the initial training period when model learns exceptionally quickly [2].
- **Warm-starting**: when training, model is introduced to a subset of the training data before seeing all training data [3].

Hypothesis: **racial bias could be mitigated**:

1. Warm-starting the model with subset of data of under-performing races in the first few epochs.
2. Fine-tuning single-race model using uniform data.

## WARM-STARTING EXPERIMENT



(\*) Feature Extraction Architecture: VGGm-11 or ResNet-18

Results of Warm-starting on **underperforming races: Asian and African**

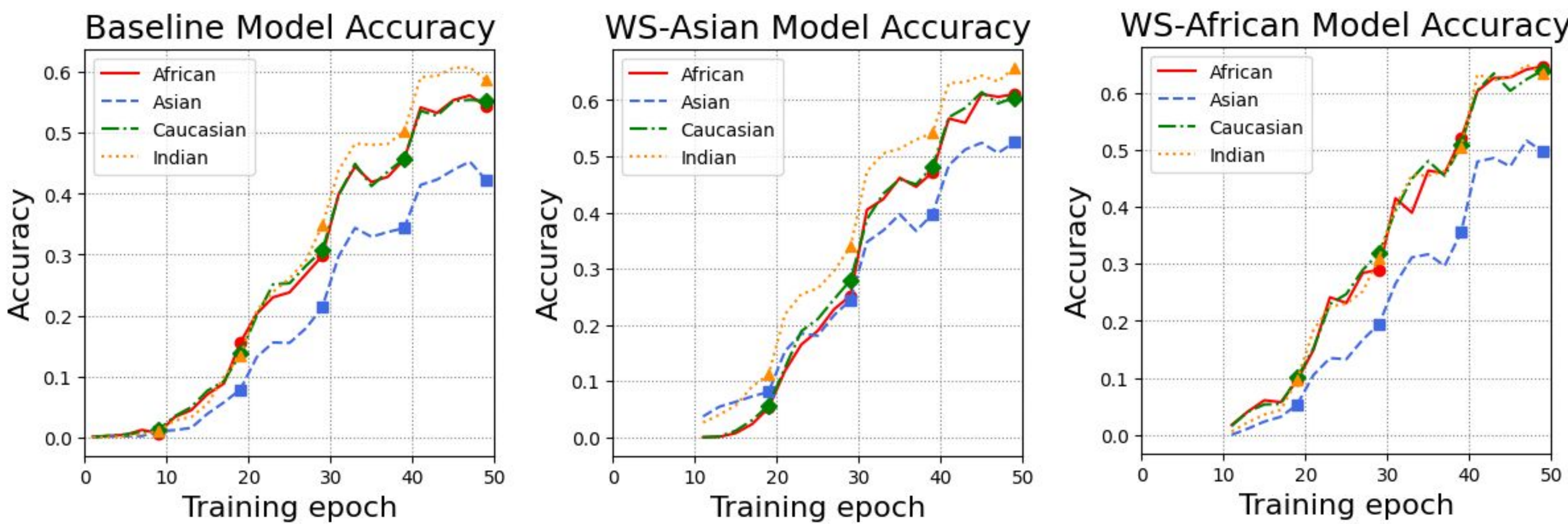


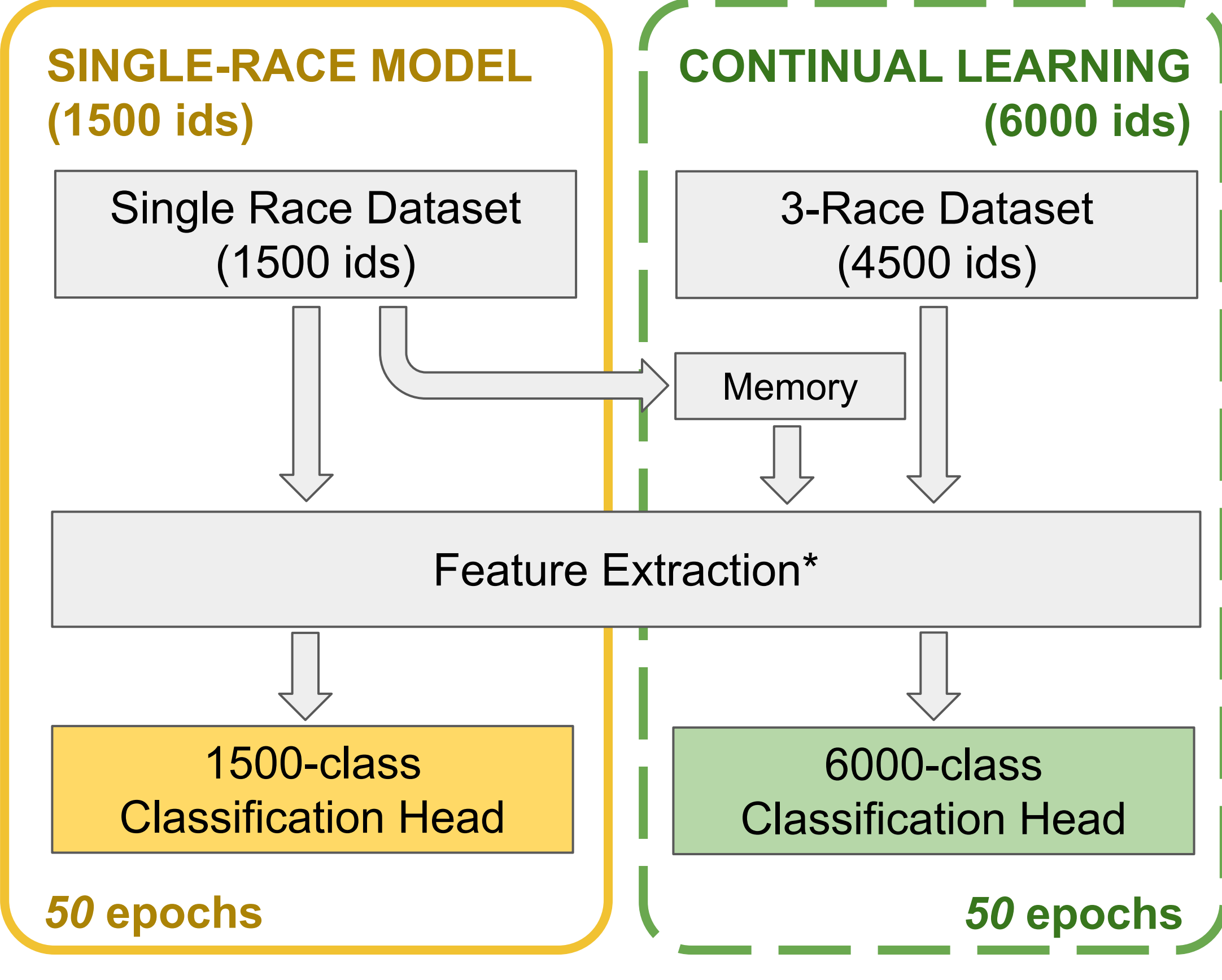
Figure 1. Model's accuracy on different races during training

Control	None	Race	Epoch	Race & Epoch
P-Values (Baseline vs WS-African)	0.8874	0.8462	0.3022	0.1157
P-Values (Baseline vs WS-Asian)	0.0408*	0.0412*	0.0482*	0.0485*

\*statistically significant at 5%

Table 2. Relative Accuracy to Caucasian by Epoch and Race

## CONTINUAL LEARNING EXPERIMENT



(\*) Feature Extraction Architecture: VGGm-11 or ResNet-18

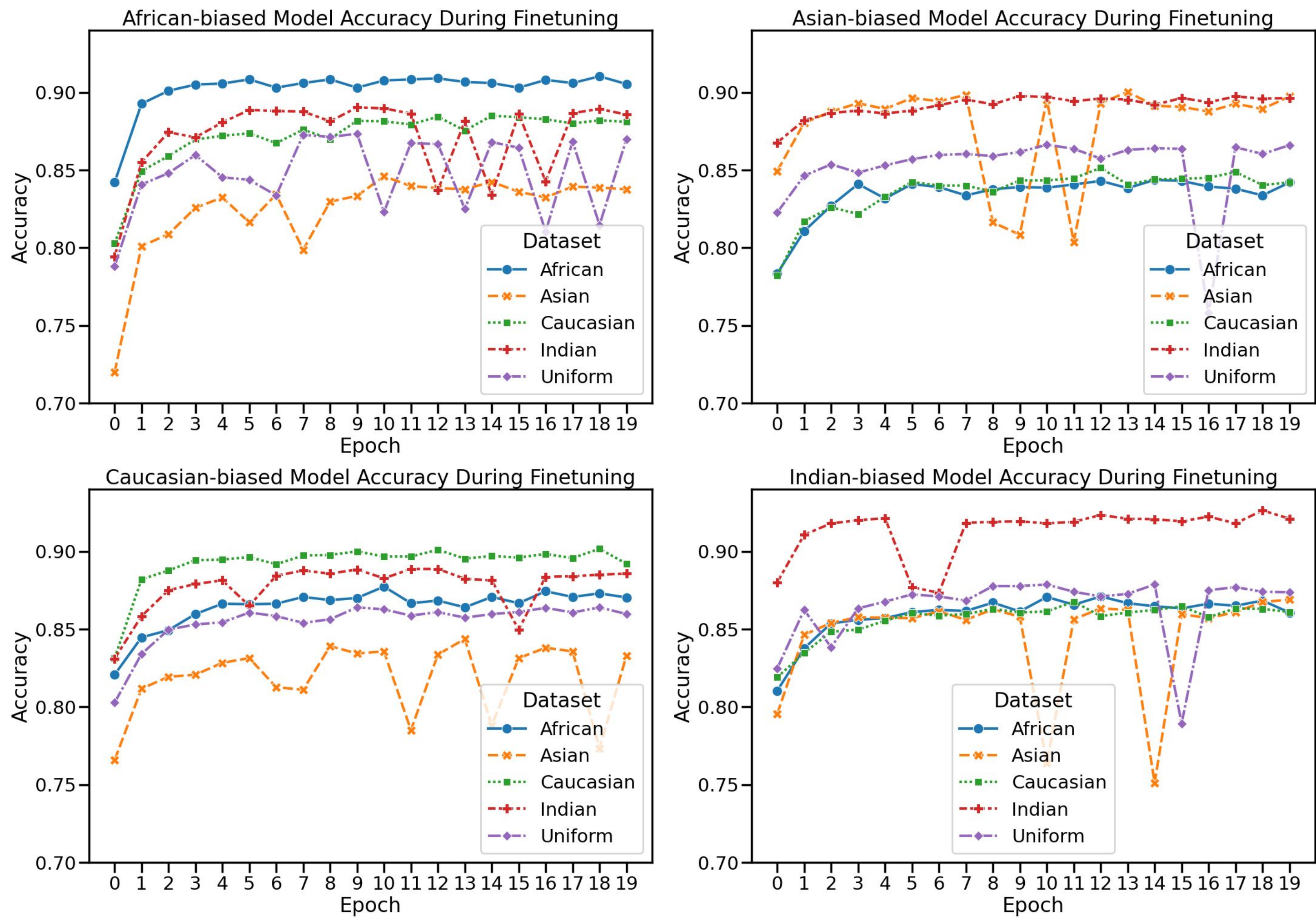


Figure 2. Model's accuracy on different races during fine-tuning

## DATASETS

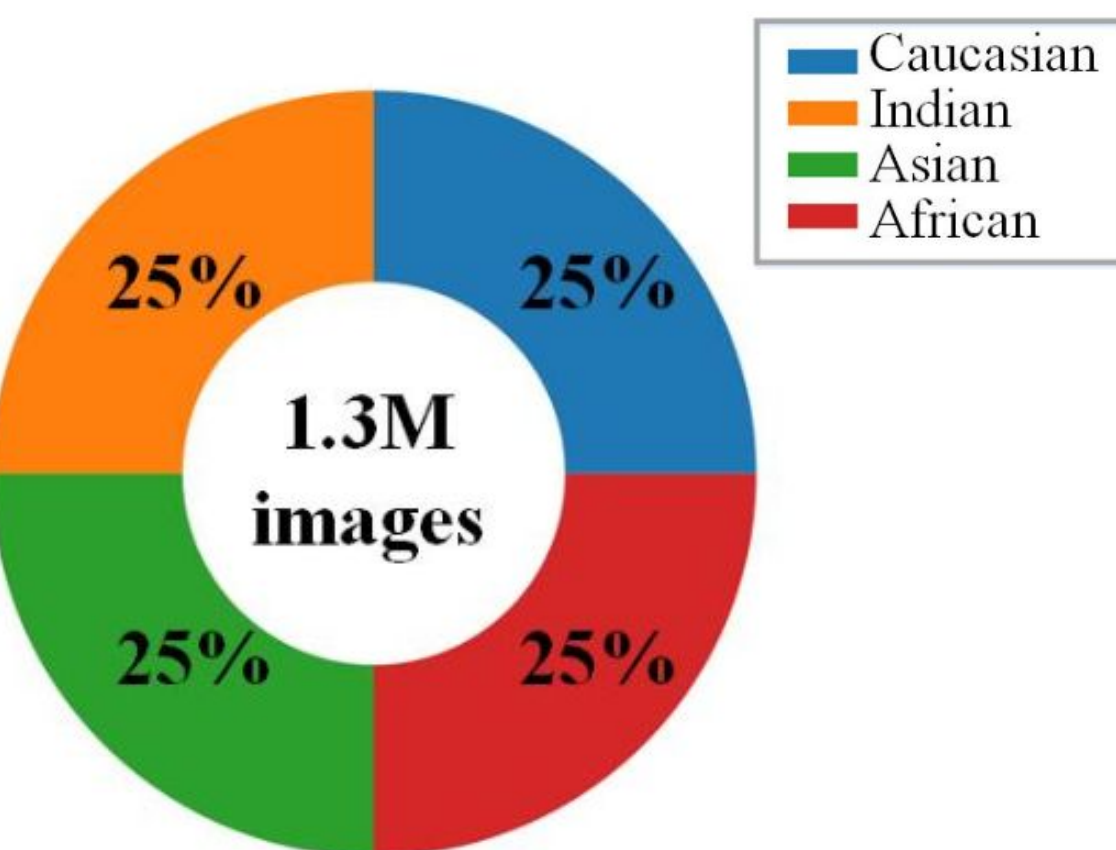


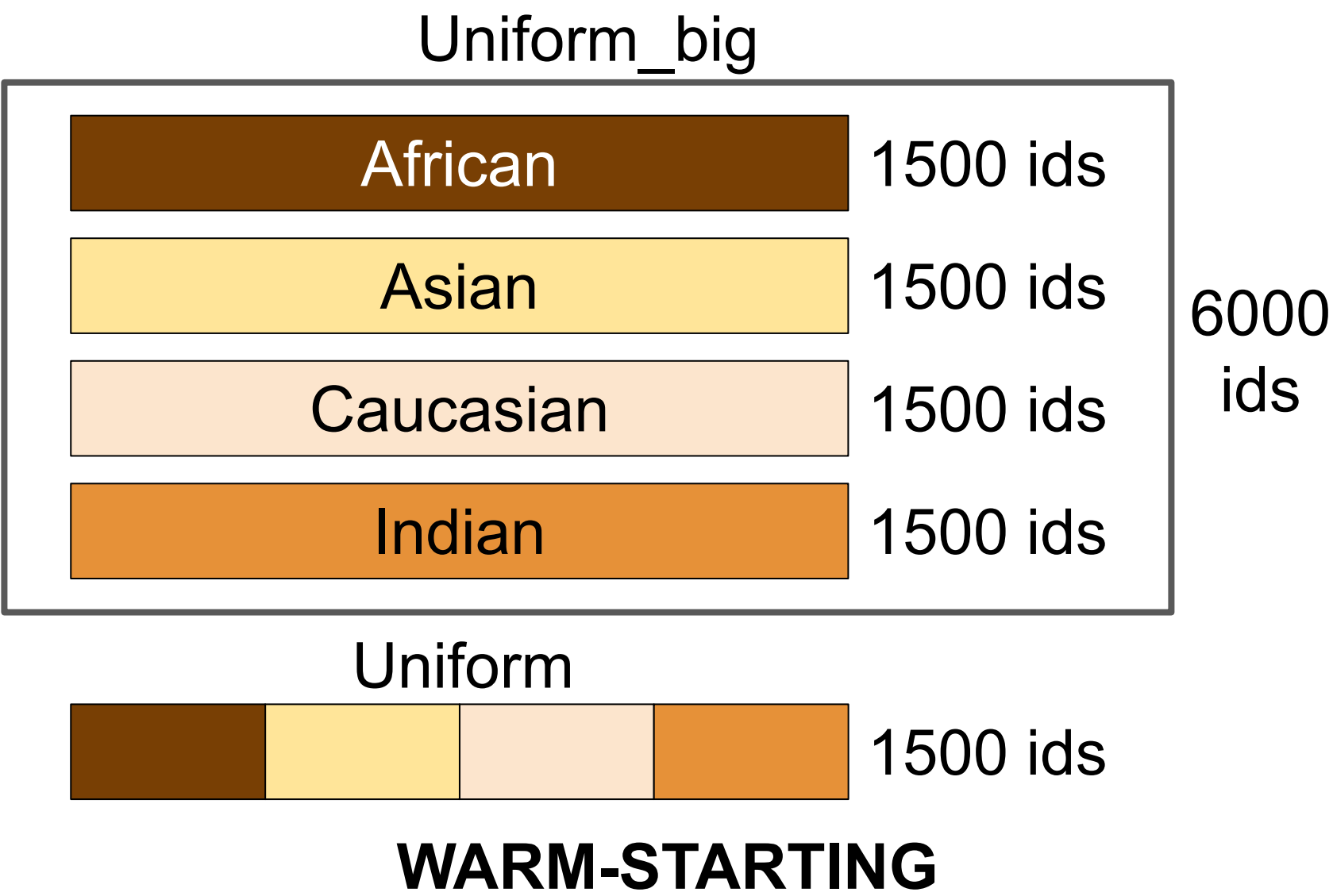
Fig 2. BUPT-BalancedFace [5]

### Training:

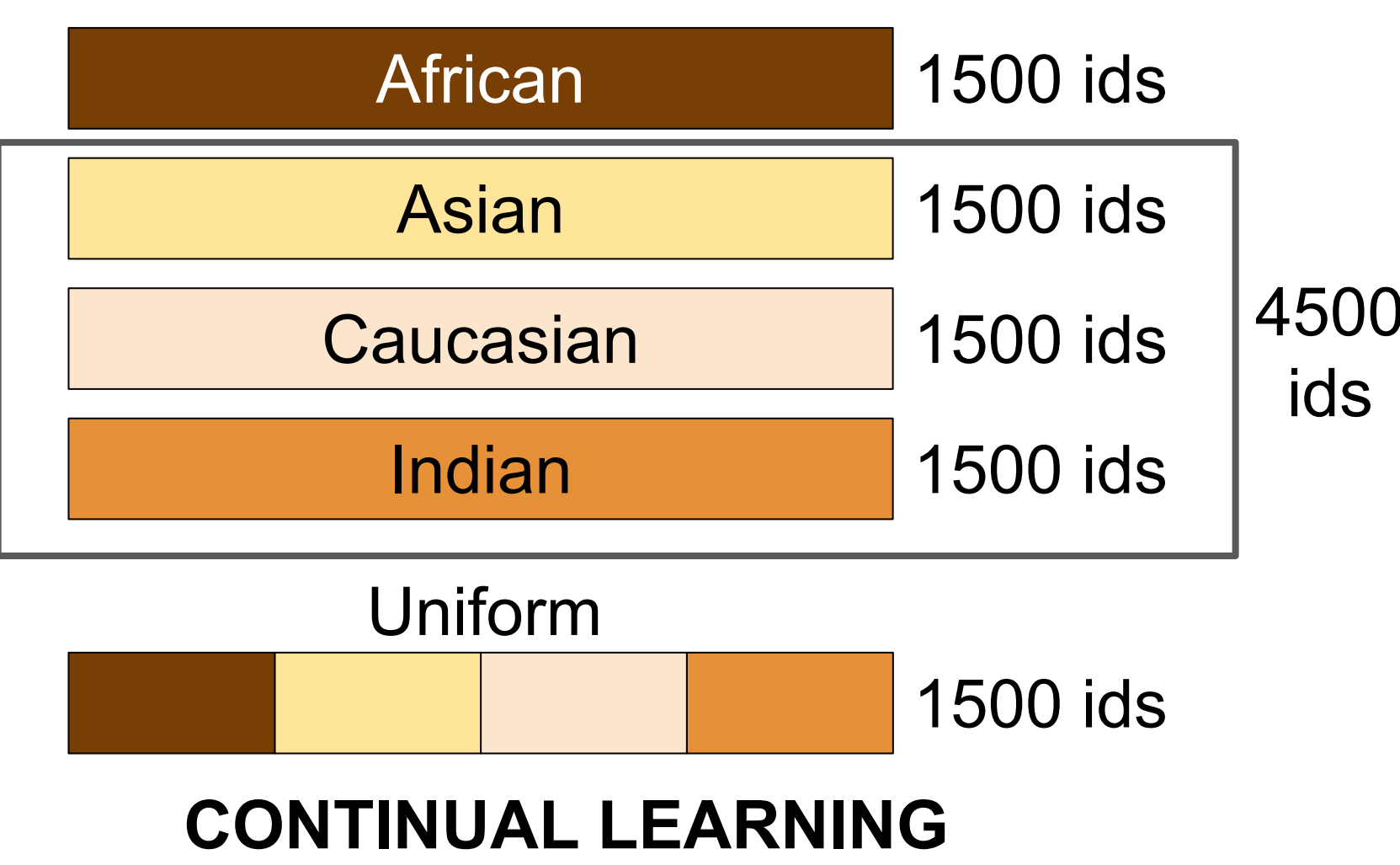
BUPT-BalancedFace [5]:  
- 1.3M images  
- 7000 identities/race

### Testing:

RFW: Racial Faces in The Wild [6]:  
- 3000 identities/race



Singe Race  
3-Race



## REFERENCES

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4. German I. Parisi, Ronald Kemker, Jose L. Part, Christopher Kanan, and Stefan Wermter. Continual lifelong learning with neural networks: A review. Neural Networks, 113:54–71, 2019. ISSN 0893-6080. doi: <https://doi.org/10.1016/j.neunet.2019.01.012>. URL <https://www.sciencedirect.com/science/article/pii/S0893608019300231.298>
5. Mei Wang, Weihong Deng. Mitigating Bias in Face Recognition using Skewness-Aware Reinforcement Learning. CVPR2020.
6. Mei Wang, Weihong Deng, Jiani Hu, Xunqiang Tao, Yaohai Huang. Racial Faces in the Wild: Reducing Racial Bias by Information Maximization Adaptation Network. ICCV2019.