



An Empirical Investigation into Benchmarking Model Multiplicity for Trustworthy Machine Learning: A Case Study on Image Classification

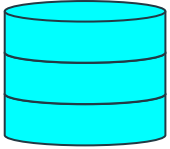
Prakhar Ganesh



What is Model Multiplicity?

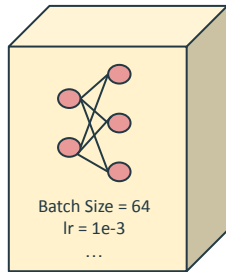
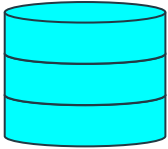
What is Model Multiplicity?

Train & Val
Dataset



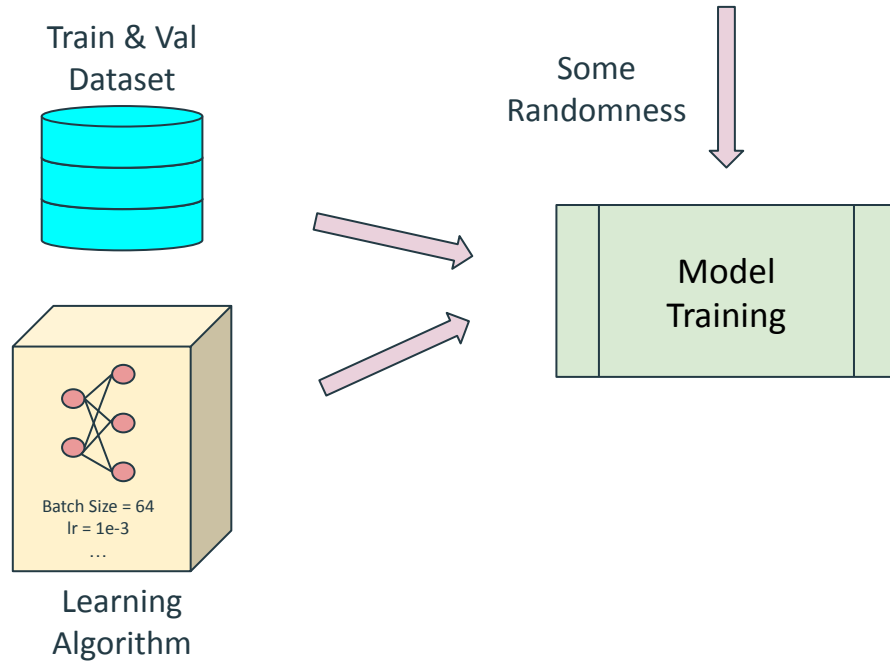
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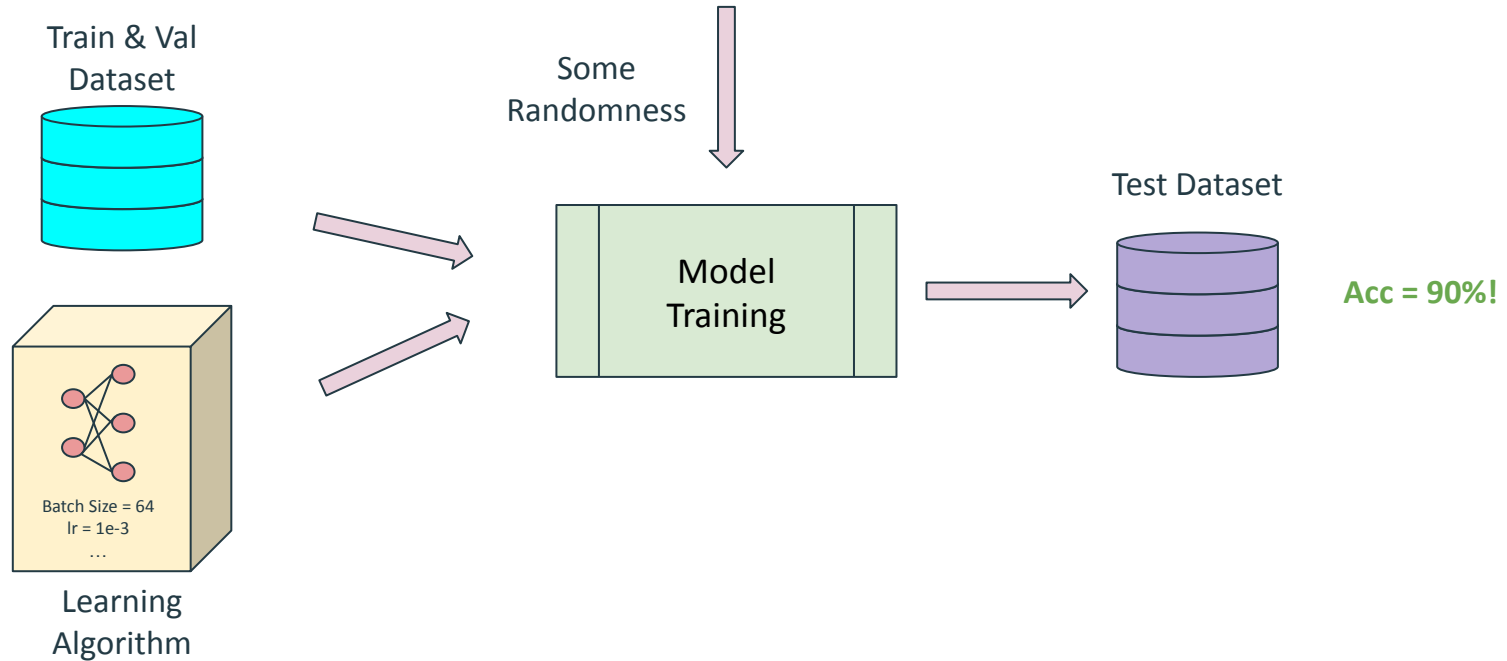


Learning
Algorithm

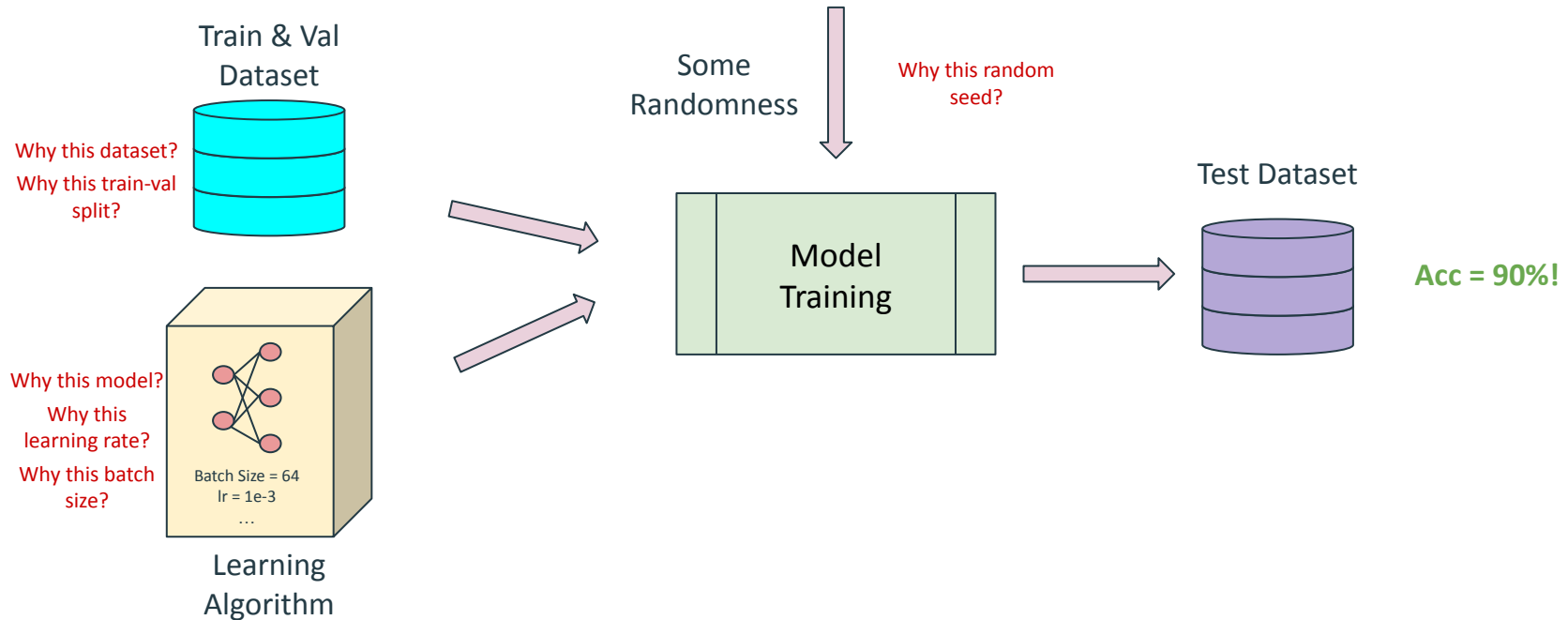
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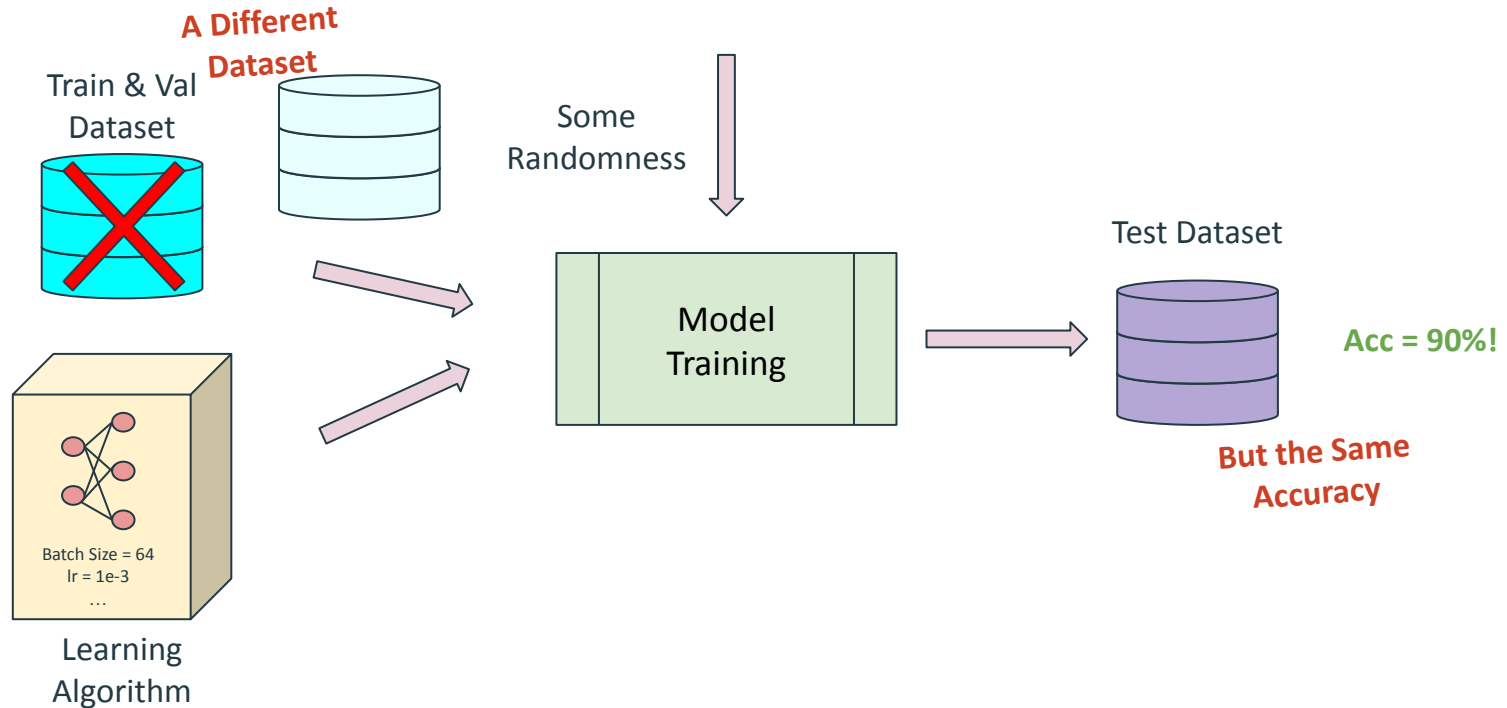
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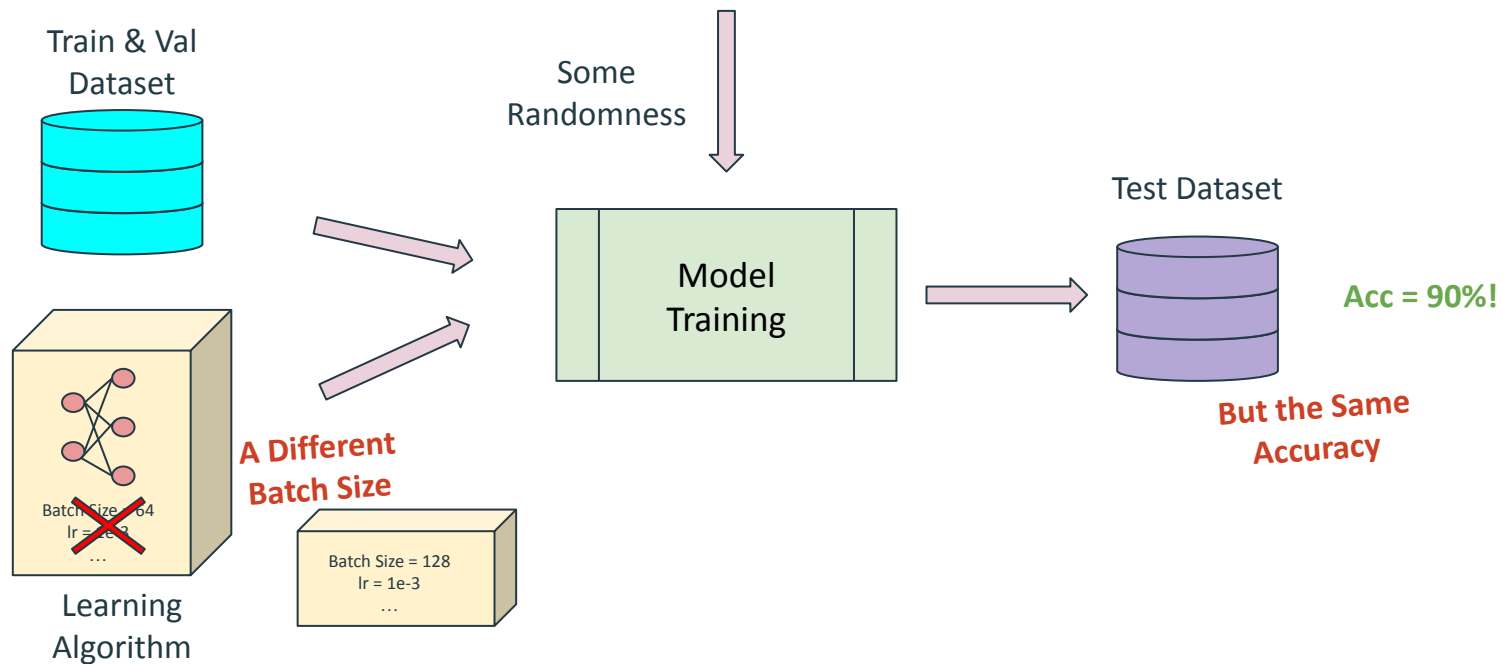
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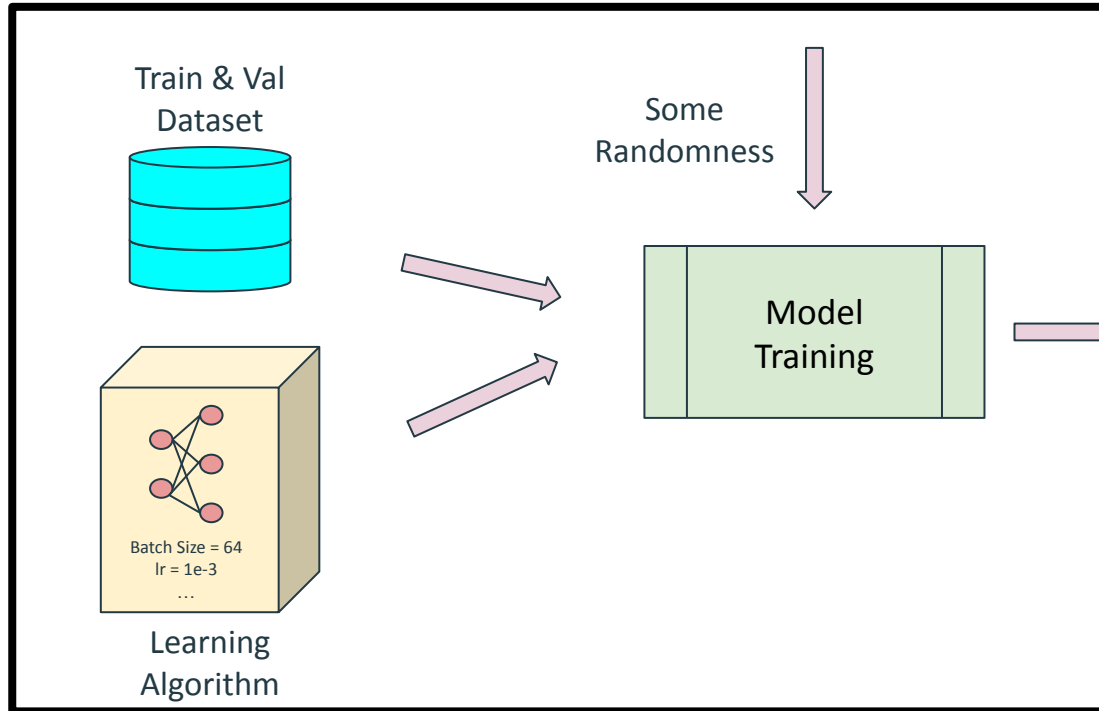


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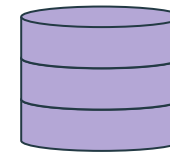


What is Model Multiplicity?

Training Setup



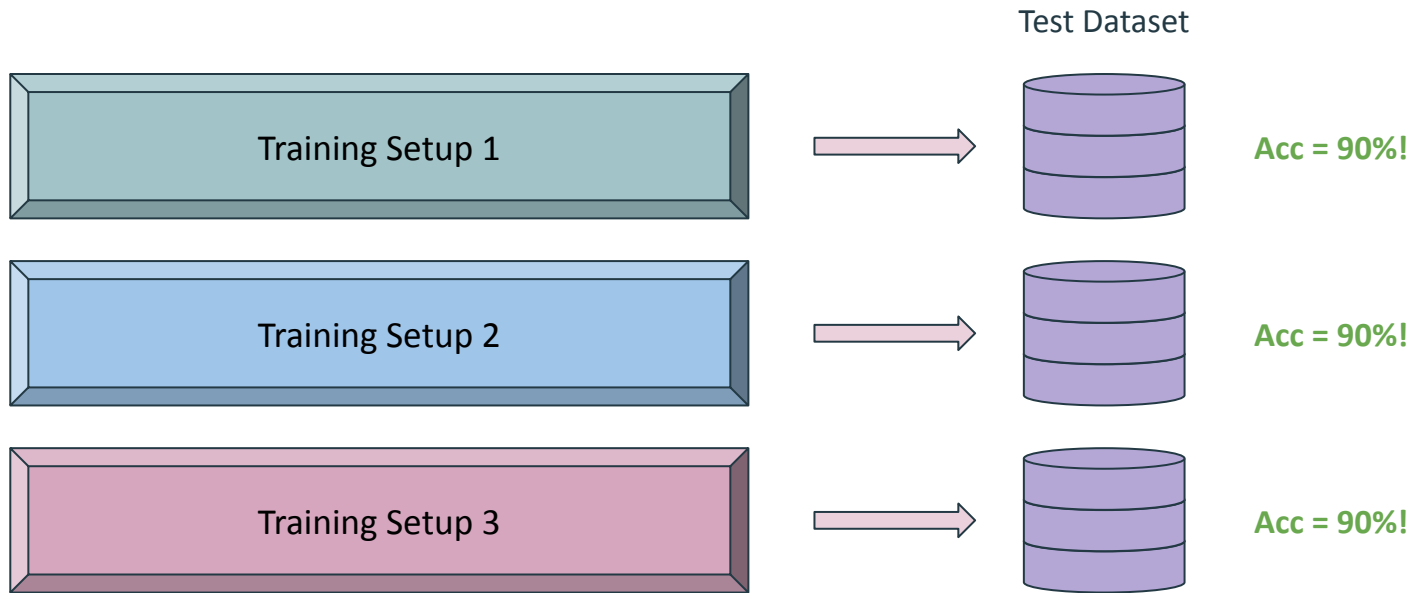
Test Dataset



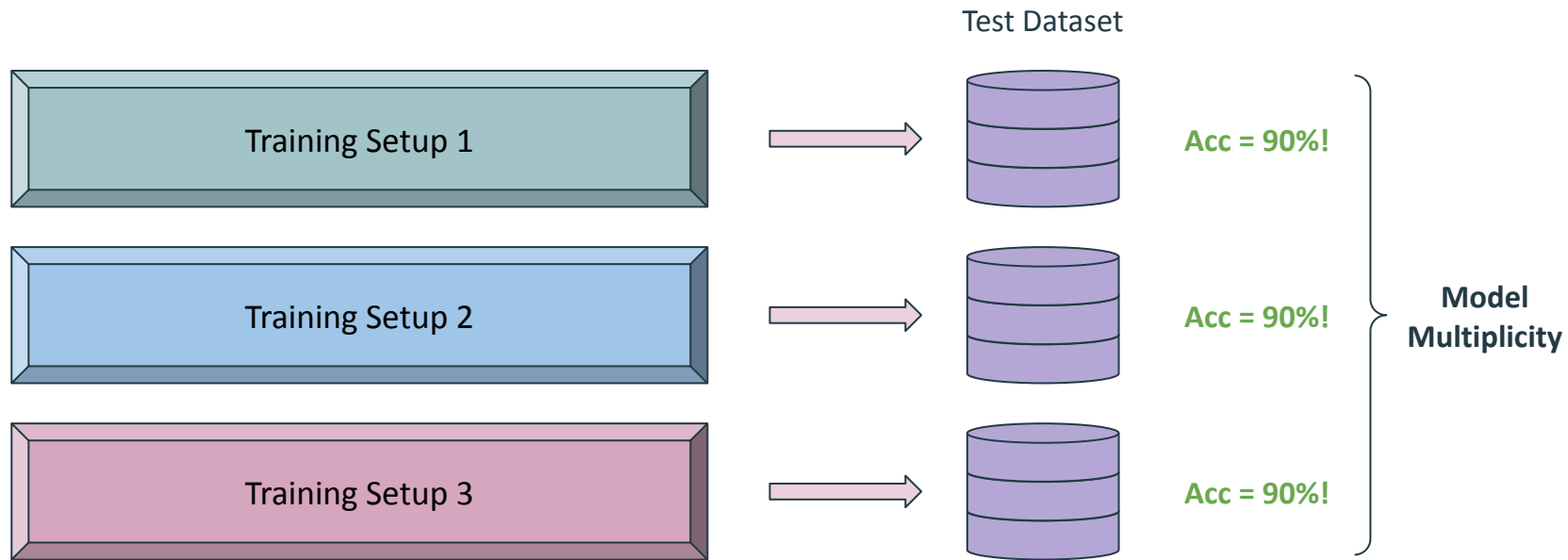
Acc = 90%

But the Same Accuracy

What is Model Multiplicity?

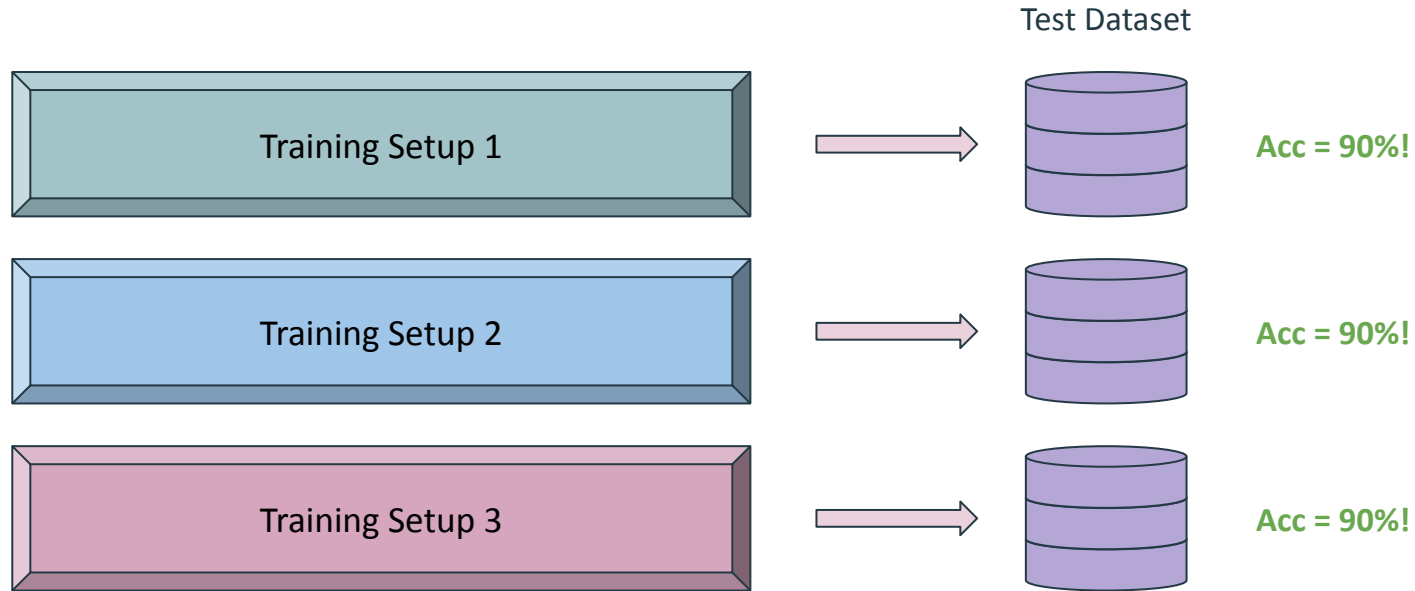


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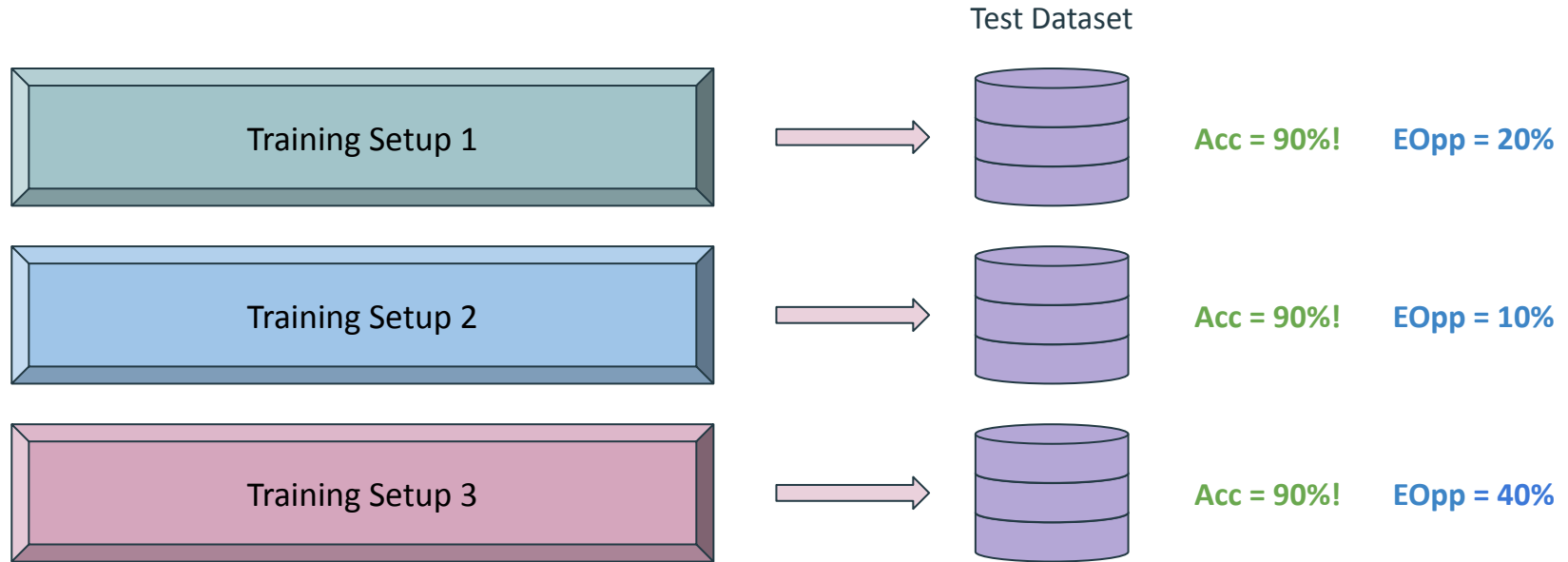


How does Model Multiplicity Impact Trustworthy ML?

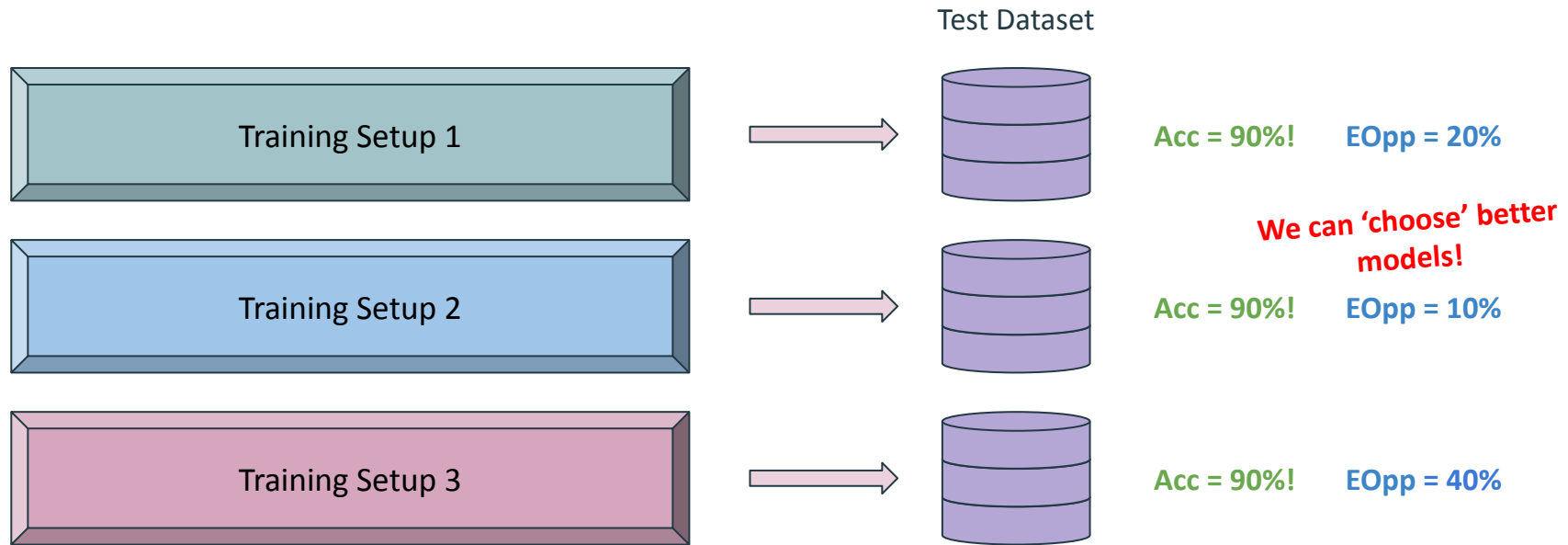
How does Model Multiplicity Impact Trustworthy ML?



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Contributions

- A standardized framework to measure and study model multiplicity.
- An empirical benchmark study of model multiplicity in image classification.
- Investigating the impact of model selection on unseen failure cases.

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Accuracy Under Intervention

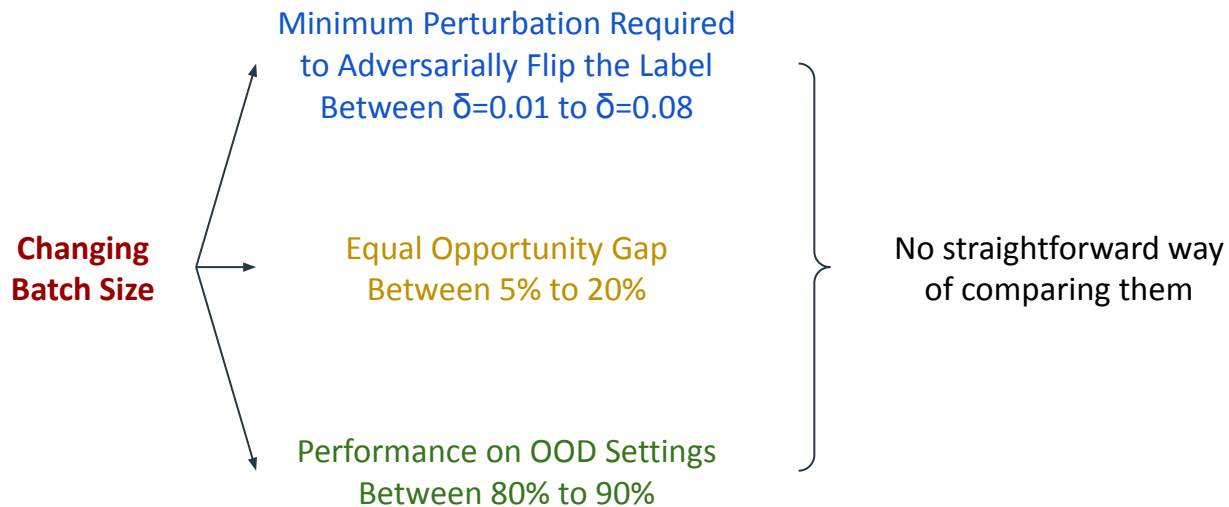
**Changing
Batch Size**

Minimum Perturbation Required
to Adversarially Flip the Label
Between $\delta=0.01$ to $\delta=0.08$

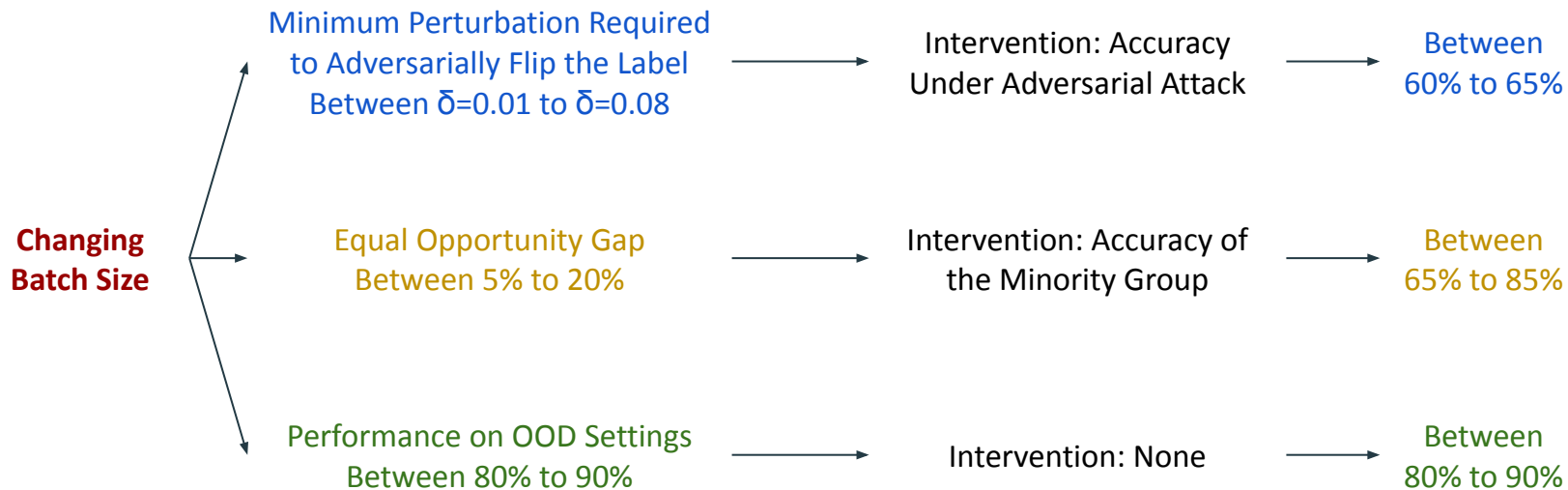
Equal Opportunity Gap
Between 5% to 20%

Performance on OOD Settings
Between 80% to 90%

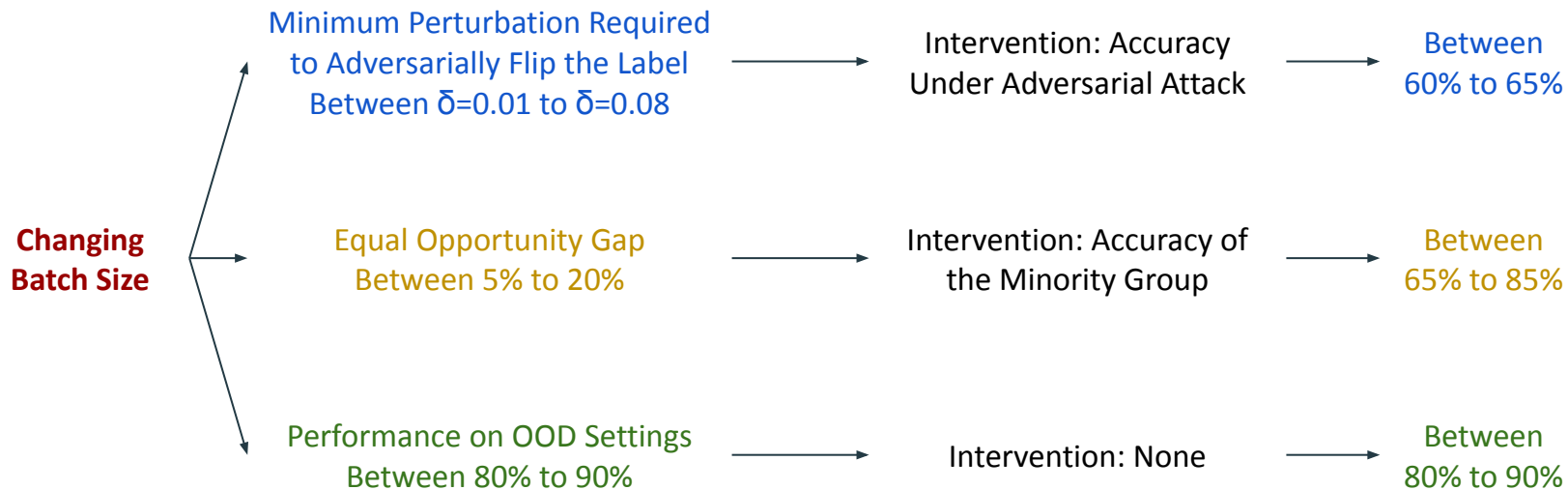
Accuracy Under Intervention



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Accuracy Under Intervention



The impact of changing batch size is a lot more severe on fairness, a little less on OOD robustness, and not that severe on adversarial robustness

Multiplicity Sheets

Multiplicity Sheets

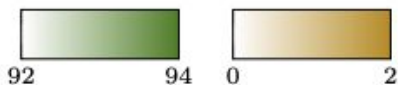
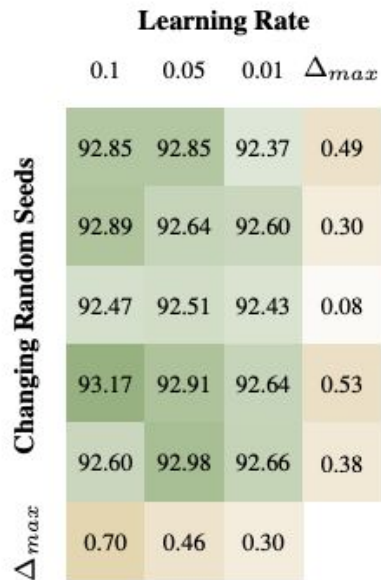
Learning Rate

0.1 0.05 0.01 Δ_{max}

Δ_{max} Changing Random Seeds

Δ_{max}

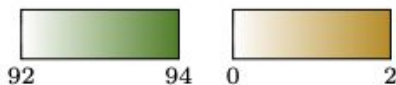
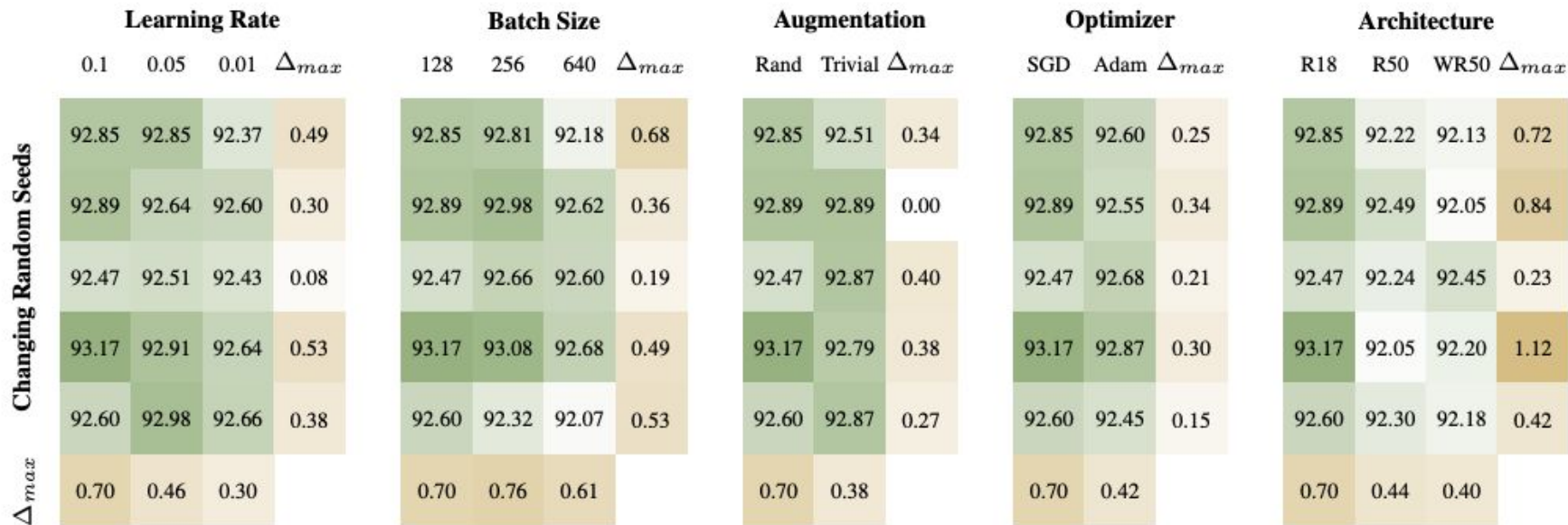
Multiplicity Sheets



Default Config: Learning Rate 0.1; Batch Size 128;
*Augmentation **Rand**; Optimizer **SGD**; Architecture **R18***

Metric: Accuracy
Dataset: UTKFace

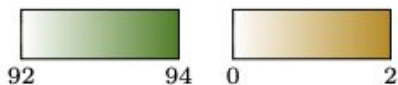
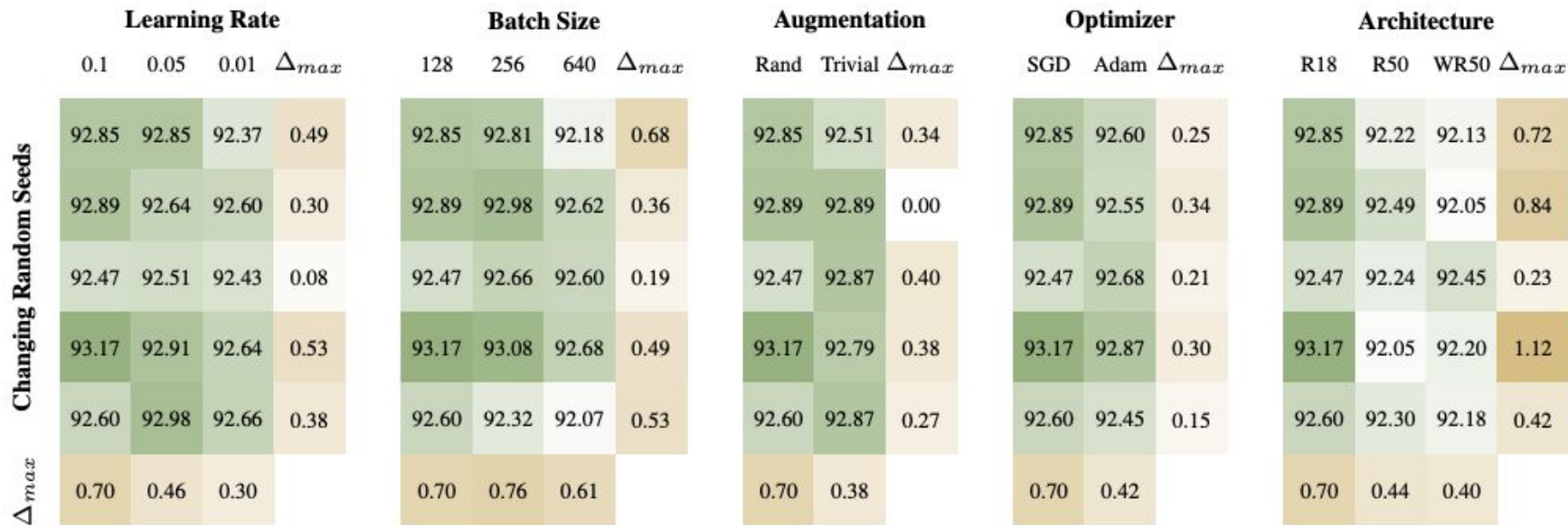
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Multiplicity Sheets



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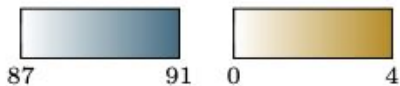
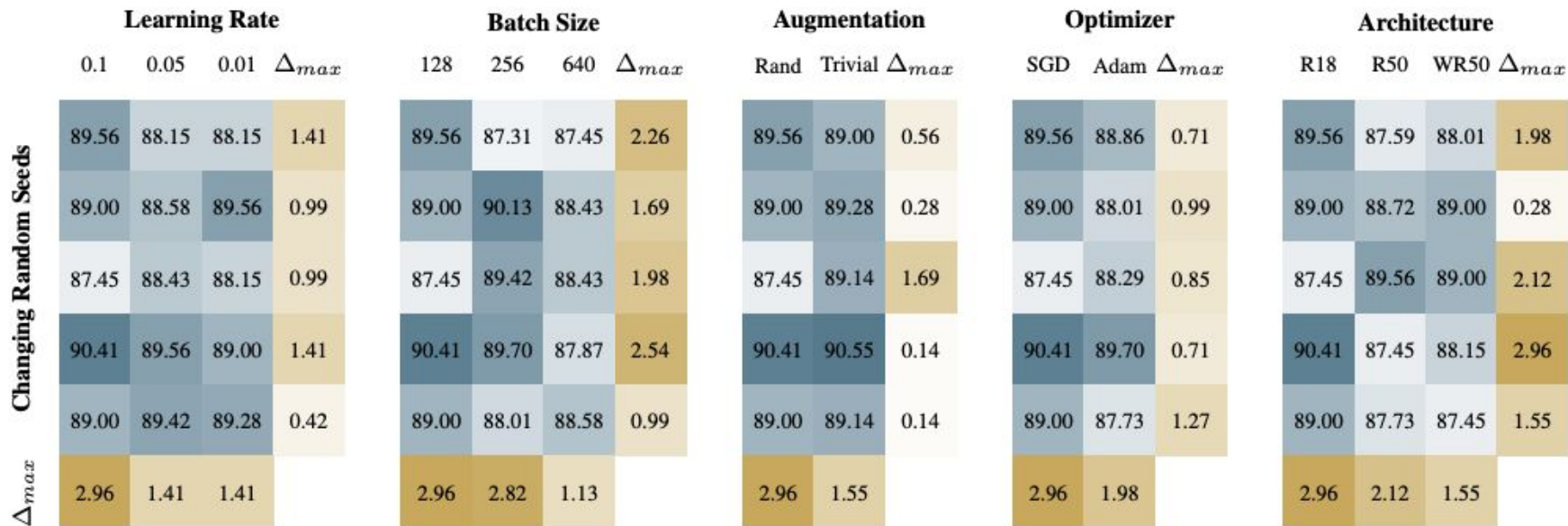
Metric: Accuracy
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Δ_{max}^{all} : **1.12**

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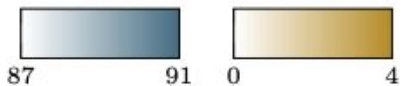
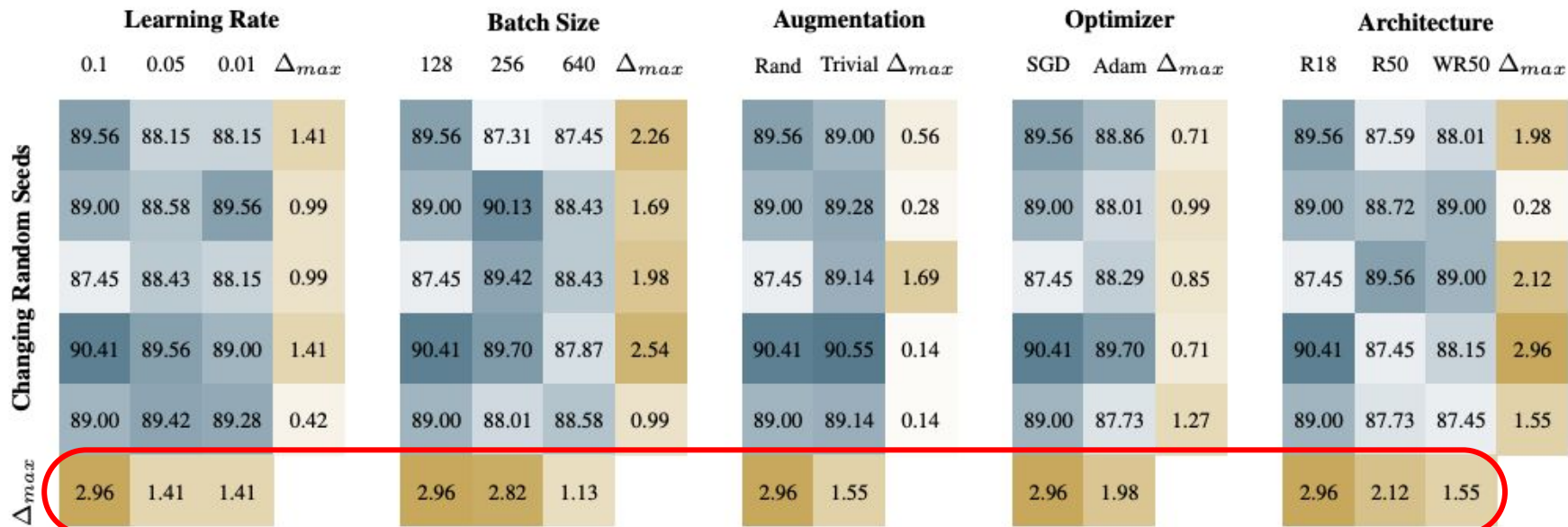
Multiplicity Sheet: Fairness



Default Config: Learning Rate **0.1**; Batch Size **128**;
 Augmentation **Rand**; Optimizer **SGD**; Architecture **R18**

Metric: **Group Accuracy (Asian)**
Dataset: **UTKFace** Δ_{max}^{all} : **3.24**

Multiplicity Sheet: Fairness

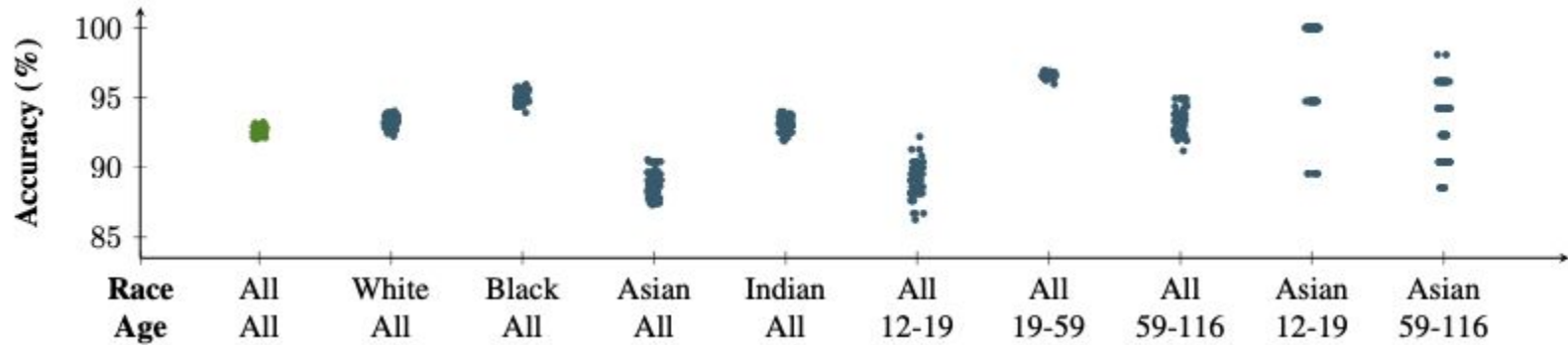


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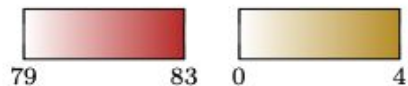
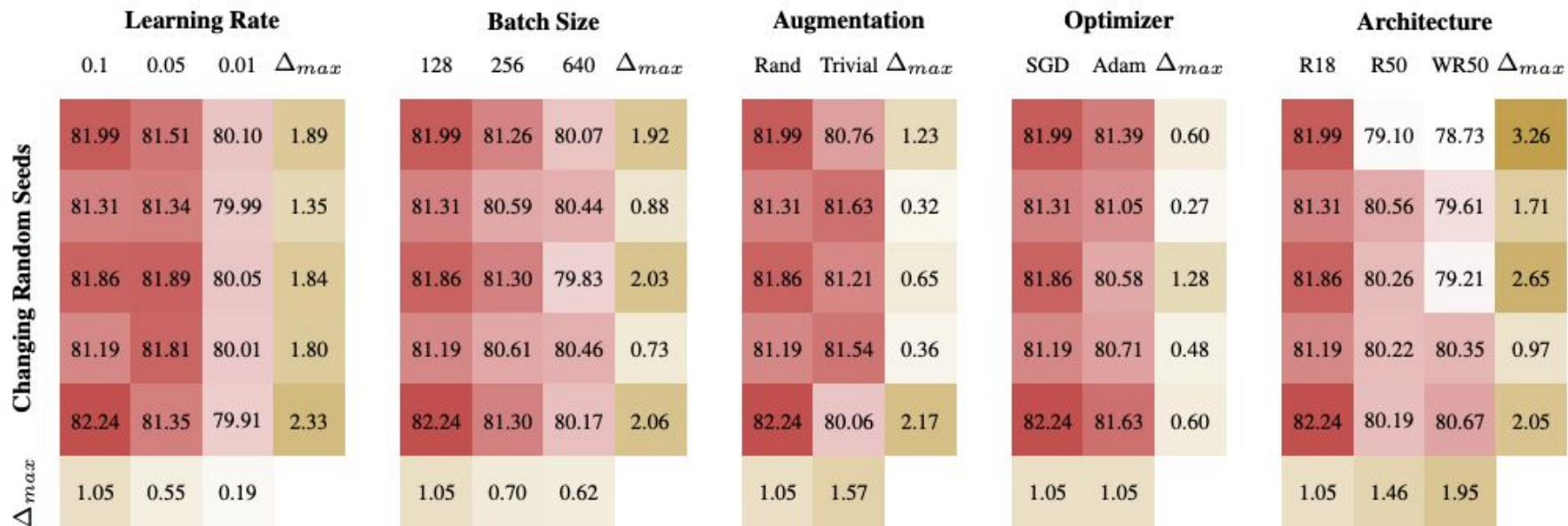
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Intersectionality and Multiplicity



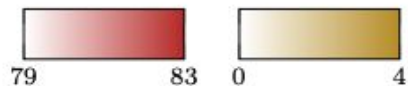
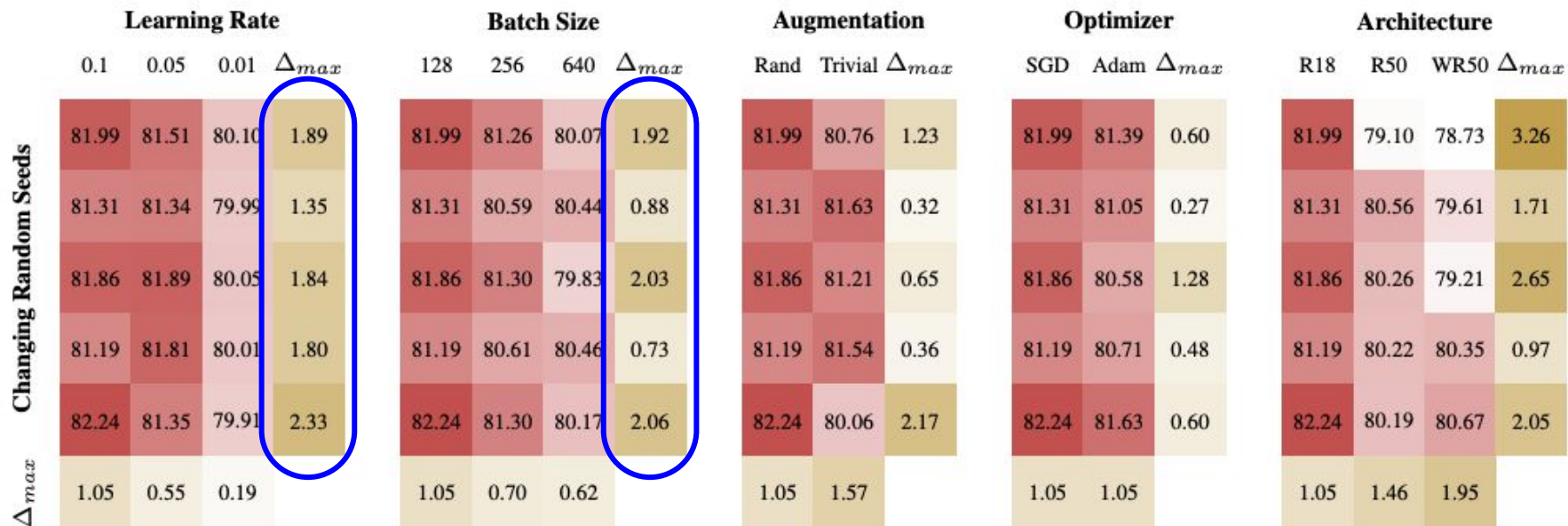
Multiplicity Sheet: Robustness



Default Config: Learning Rate **0.1**; Batch Size **128**;
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Metric: **OOD Accuracy (FairFace)**
Dataset: **UTKFace** Δ_{max}^{all} : **3.51**

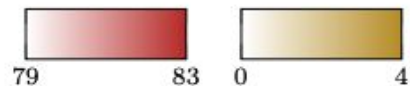
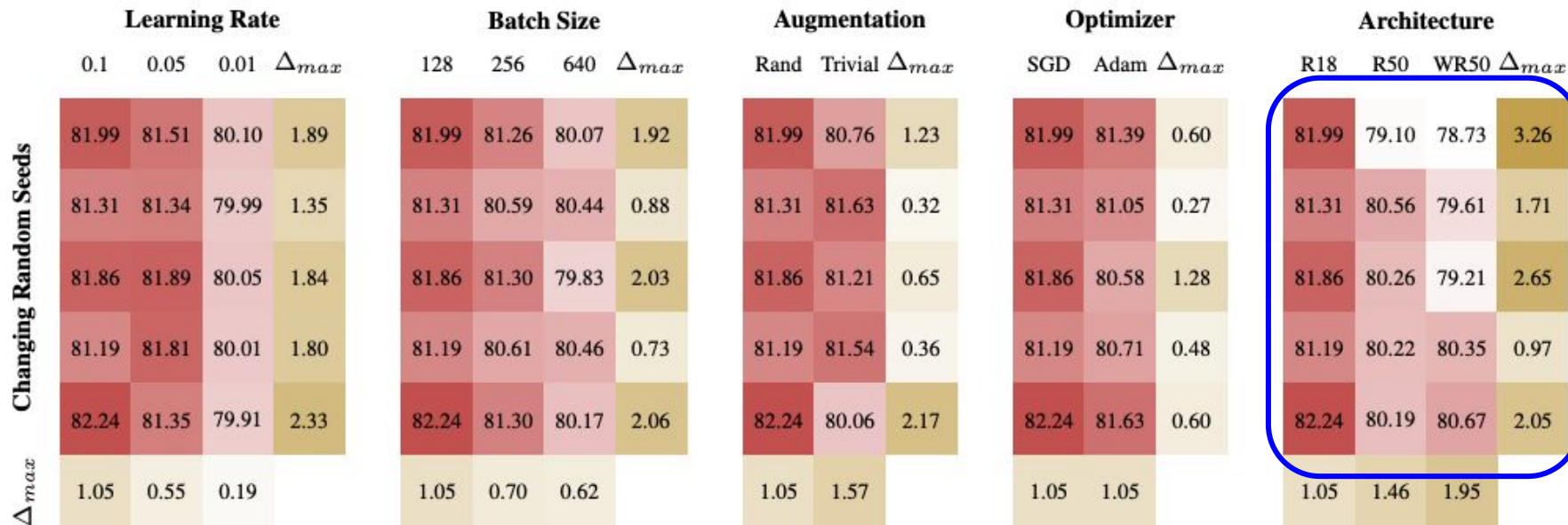
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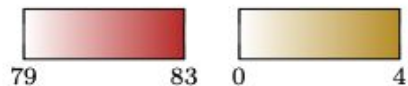
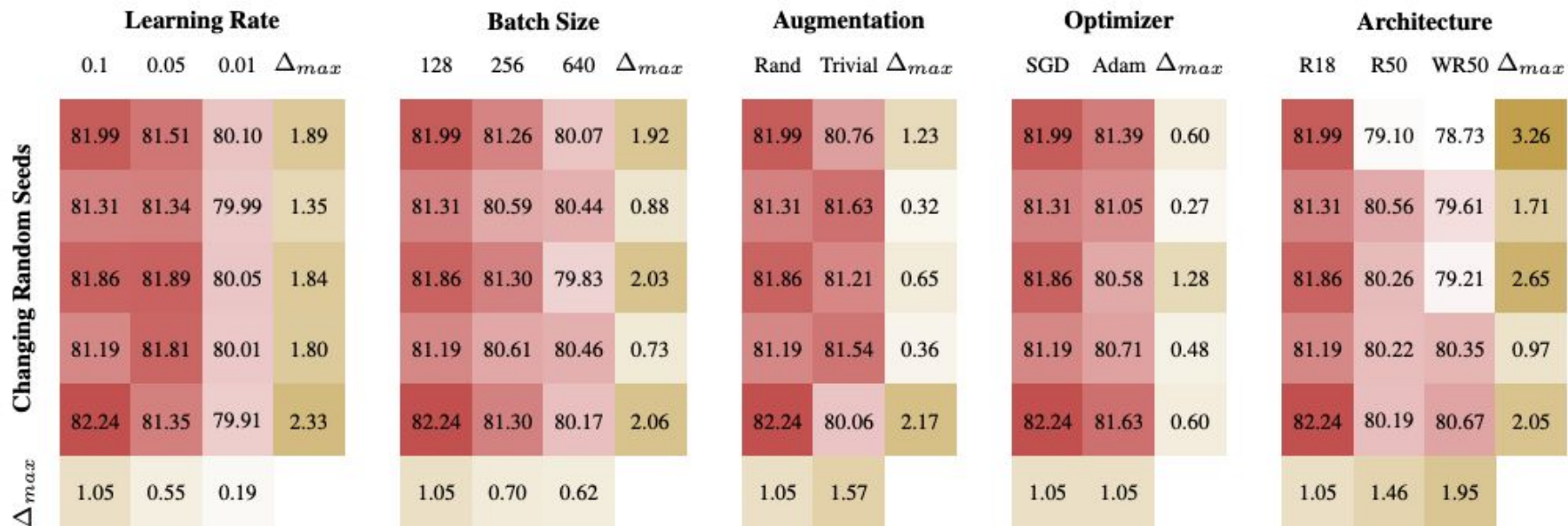
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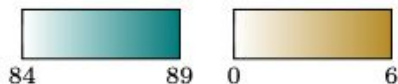
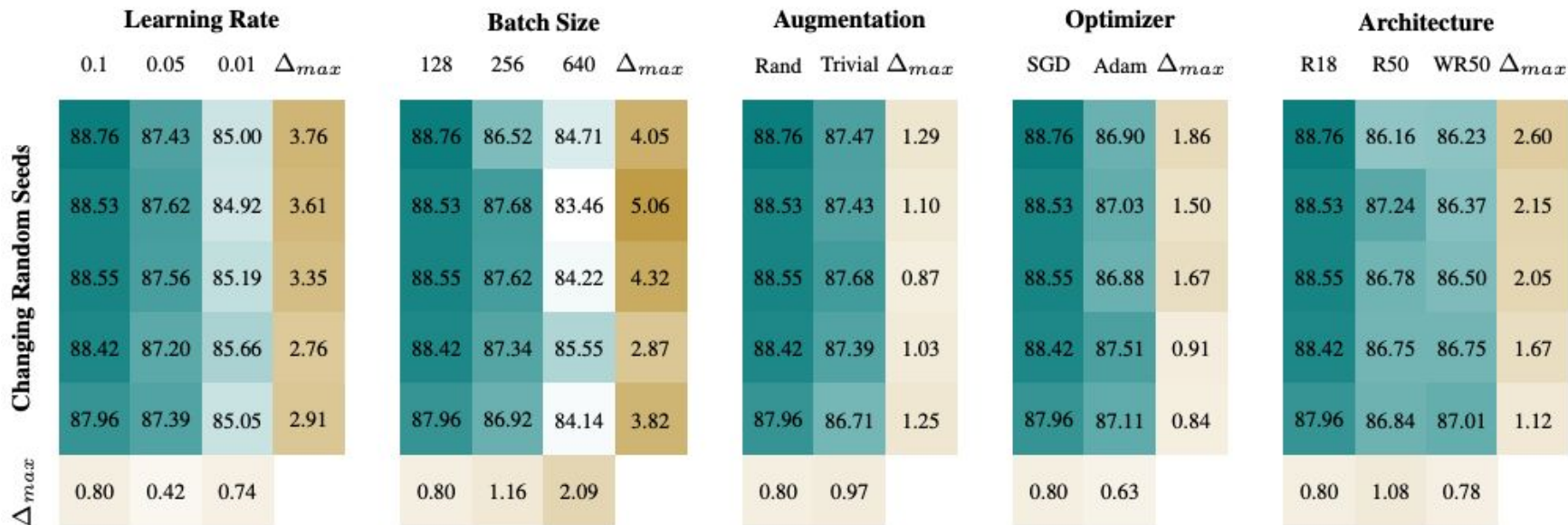
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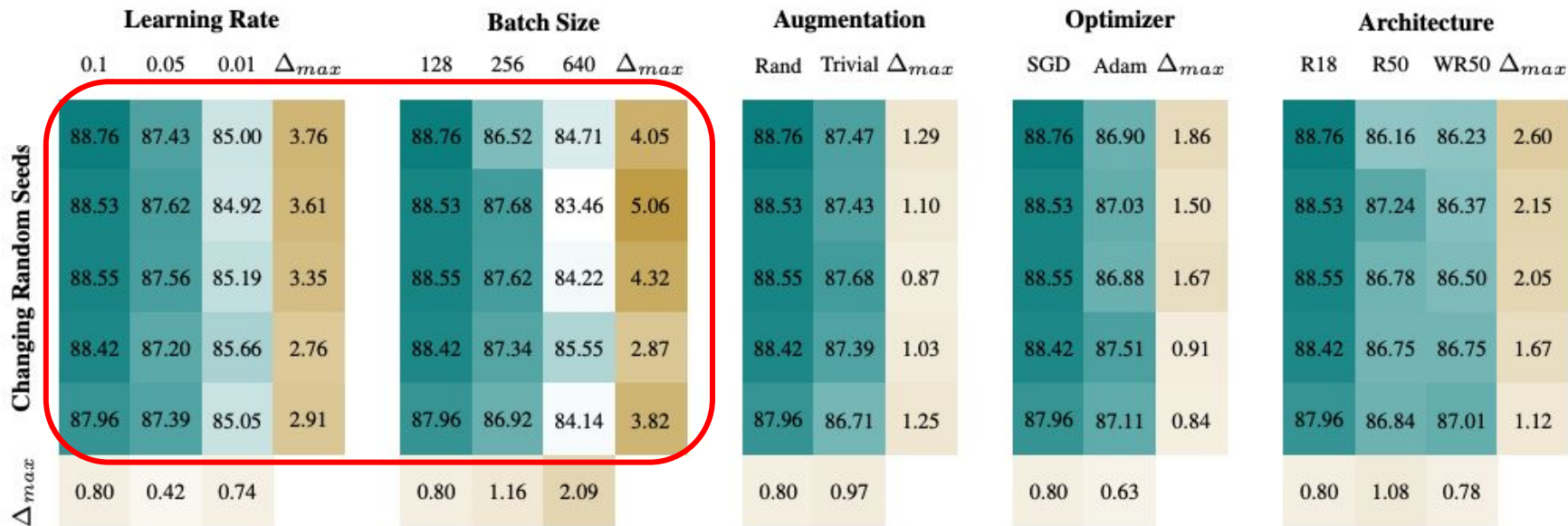
Multiplicity Sheet: Privacy



Default Config: Learning Rate **0.1**; Batch Size **128**;
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Metric: **Pert. Accuracy** ($\lambda = 5$)
Dataset: **UTKFace** Δ_{max}^{all} : **5.30**

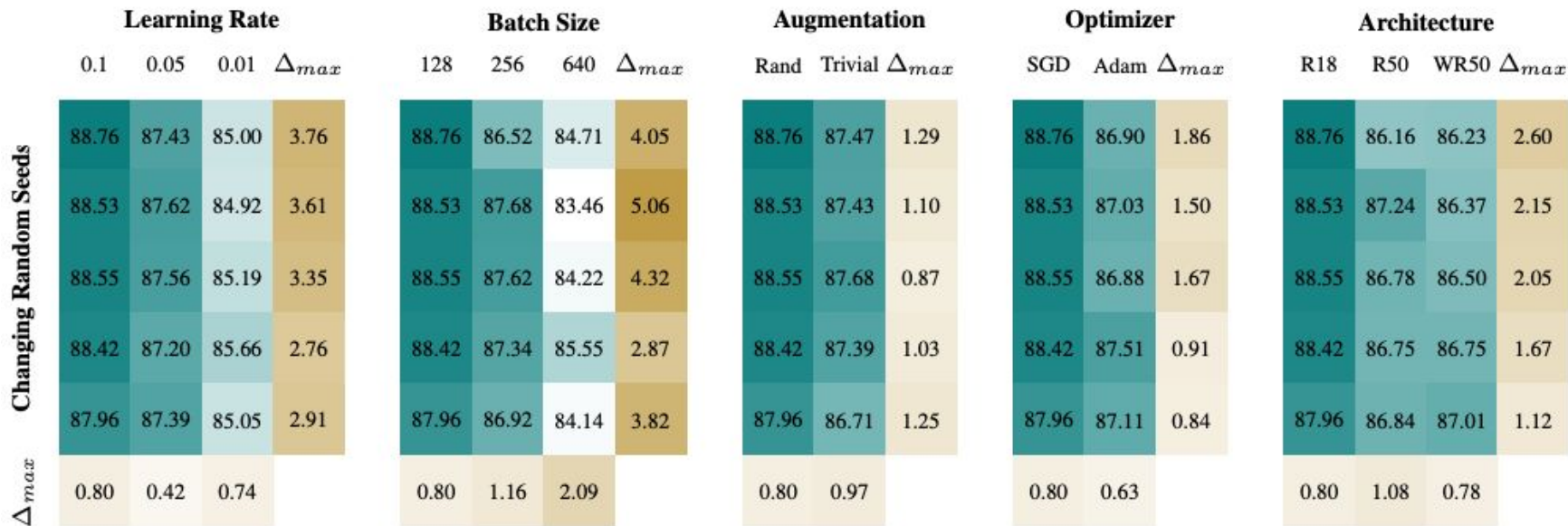
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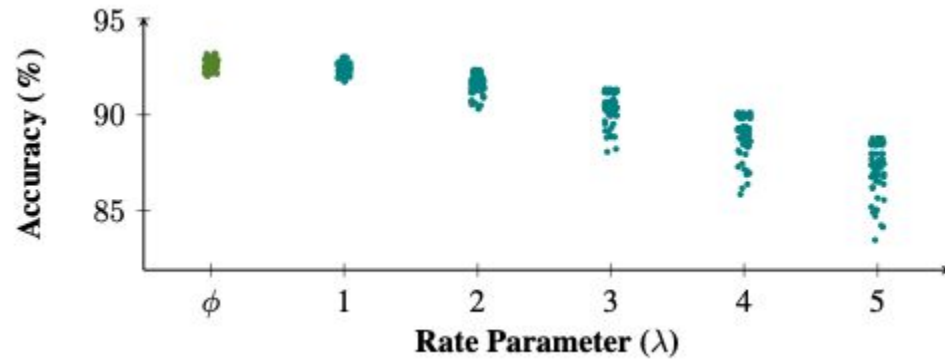
Multiplicity Sheet: Privacy



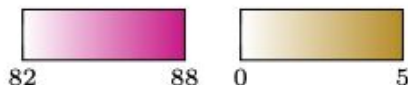
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Accuracy Under Intervention: Privacy



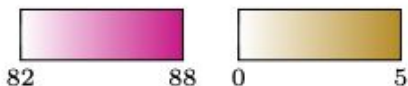
Multiplicity Sheet: Security



Default Config: Learning Rate **0.1**; Batch Size **128**;
 Augmentation **Rand**; Optimizer **SGD**; Architecture **R18**

Metric: **PGD Accuracy** ($\delta = 0.005$)
Dataset: **UTKFace** Δ_{max}^{all} : **5.53**

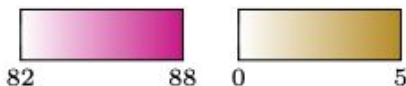
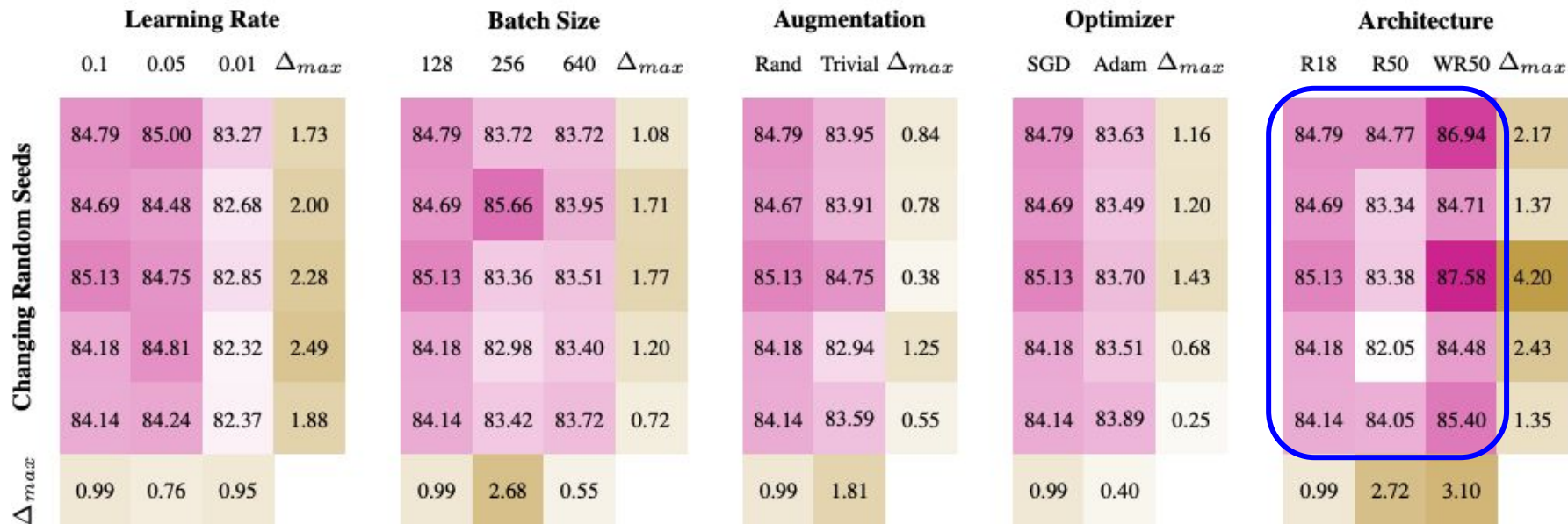
Multiplicity Sheet: Security



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Metric: **PGD Accuracy** ($\delta = 0.005$)
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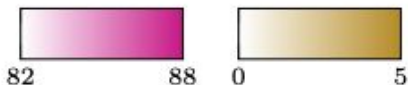
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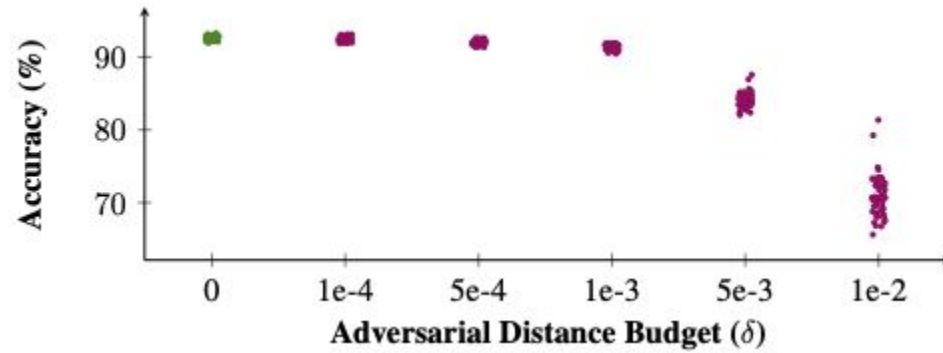
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Accuracy Under Intervention: Security



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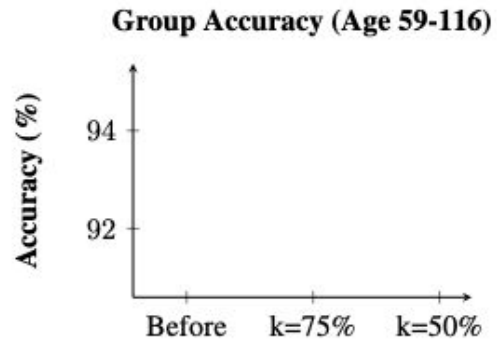
Model Selection to Counter Multiplicity

- We collect all 45 models in previous multiplicity sheets.
- We only select models which rank in the top k% of each metric, i.e.,
 - Accuracy for racial group 'Asian'
 - Accuracy on OOD dataset 'FairFace'
 - Accuracy under Output Perturbations for Privacy
 - Accuracy under PGD Adversarial Attacks $\delta=0.005$

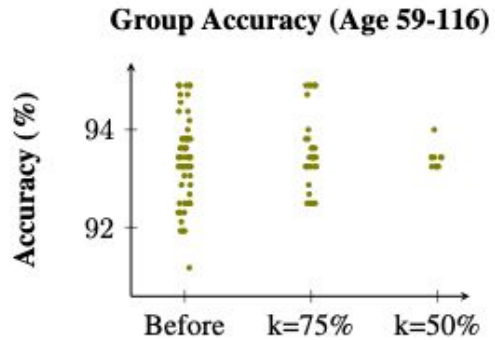
Model Selection to Counter Multiplicity: Unseen Metrics

- To 'simulate' unseen failure cases
 - Accuracy for **age group '59-116'**
 - Accuracy on OOD dataset '**CelebA**'
 - Accuracy under **Input** Perturbations for Privacy
 - Accuracy under PGD Adversarial Attacks **$\delta=0.01$**

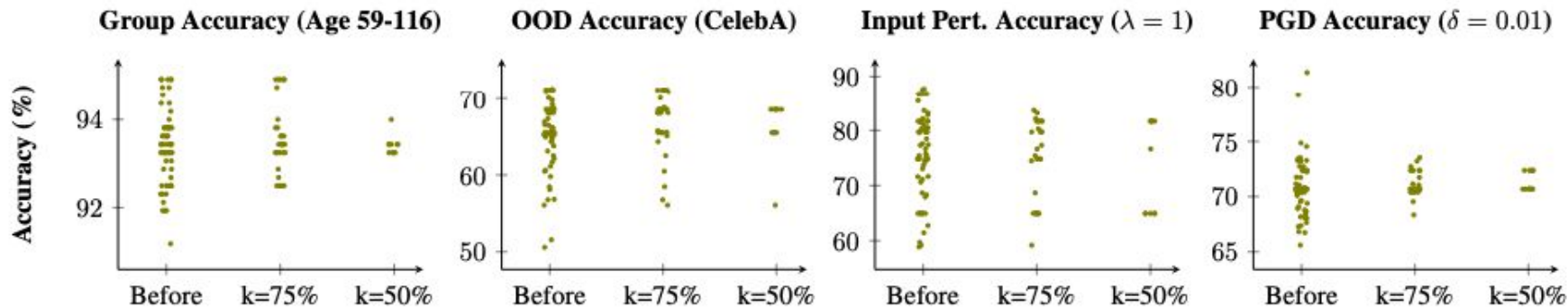
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- We performed a detailed case study and benchmarked model multiplicity of various trustworthy ML metrics for image classification.
- We showed empirically that the concerns of model multiplicity persist even beyond model selection.