

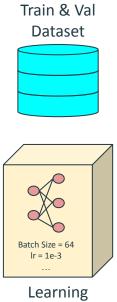


An Empirical Investigation into Benchmarking Model Multiplicity for Trustworthy Machine Learning: A Case Study on Image Classification

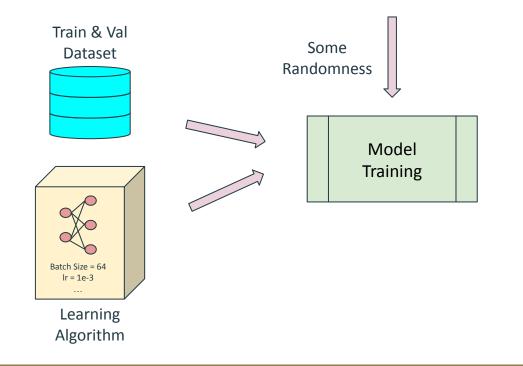
Prakhar Ganesh

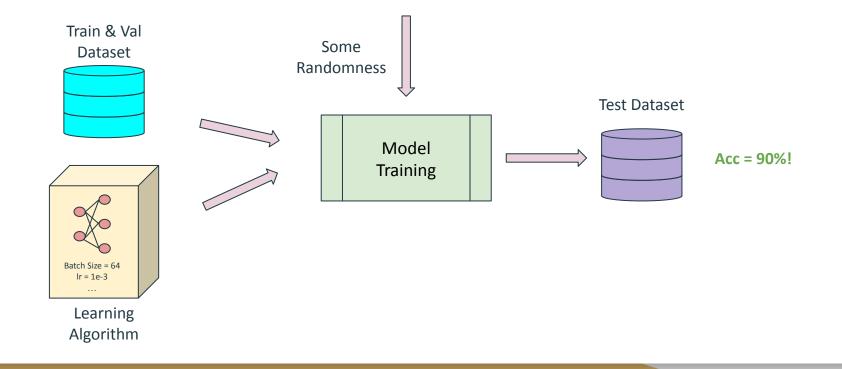


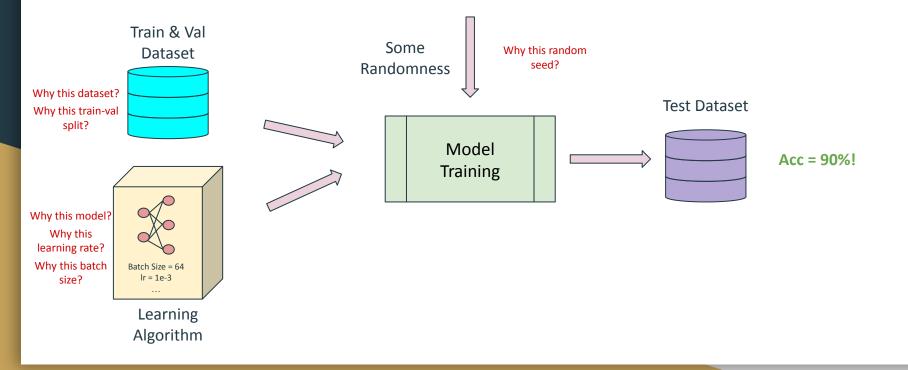


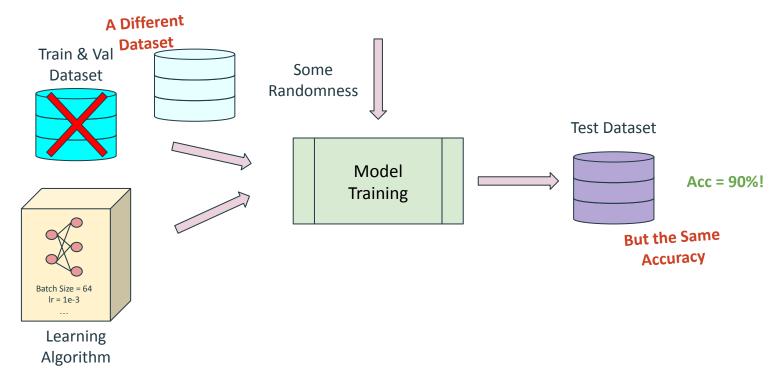


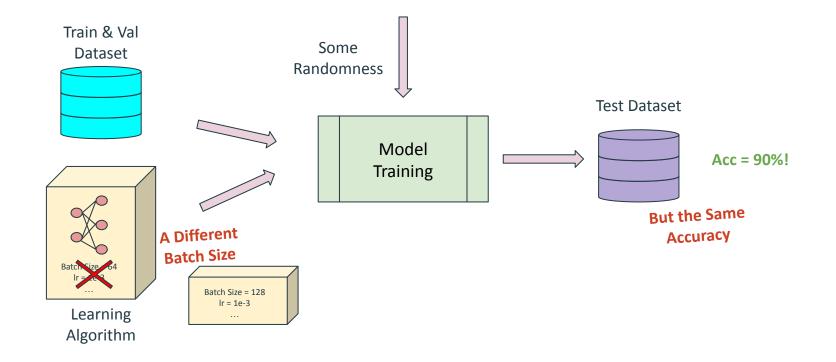
Algorithm

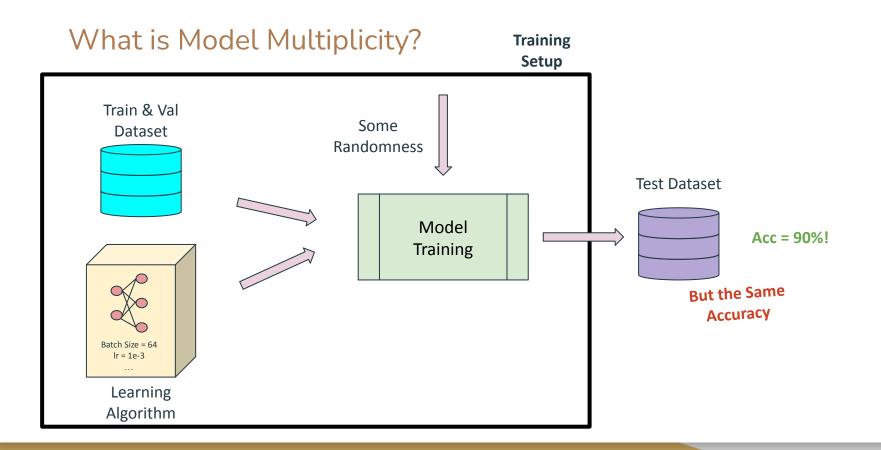


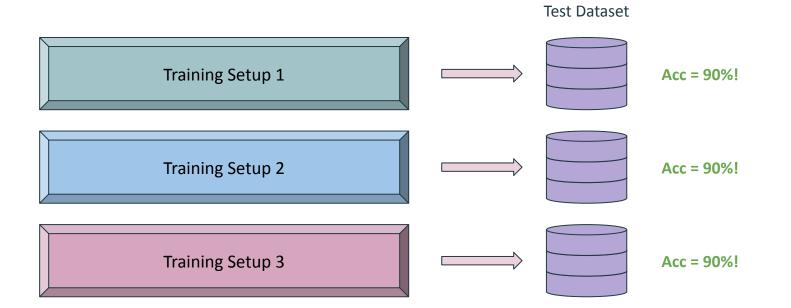


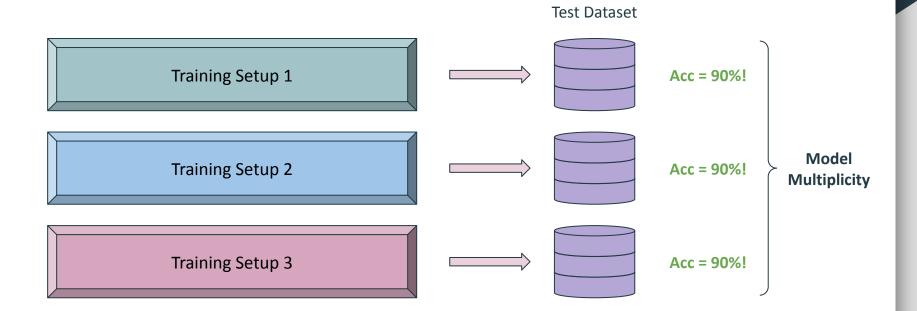


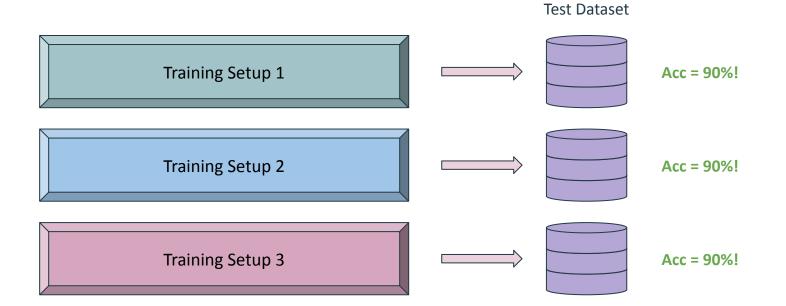


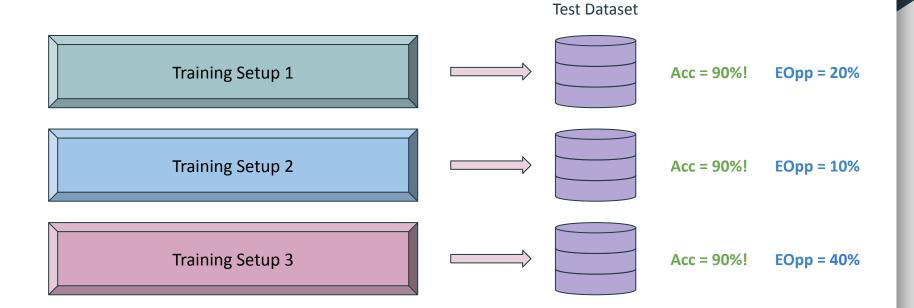


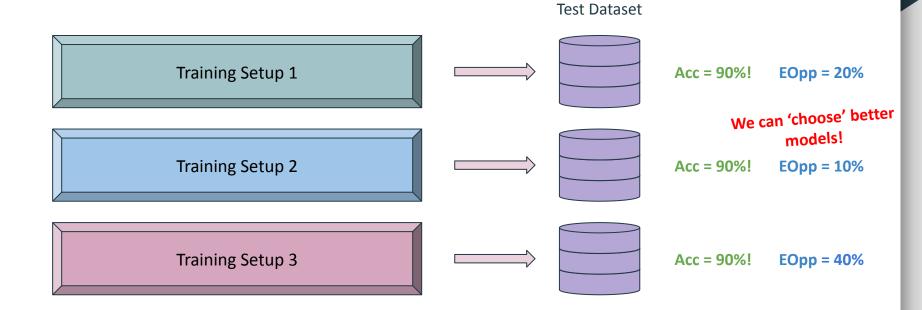












Contributions

- 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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Changing Batch Size Between δ =0.01 to δ =0.08

Minimum Perturbation Required to Adversarially Flip the Label

Equal Opportunity Gap Between 5% to 20%

Performance on OOD Settings Between 80% to 90%

to Adversarially Flip the Label Between δ=0.01 to δ=0.08

Minimum Perturbation Required

Changing Batch Size Equal Opportunity Gap Between 5% to 20%

Performance on OOD Settings Between 80% to 90% No straightforward way of comparing them

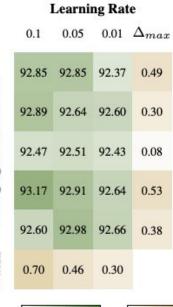
Minimum Perturbation Required Intervention: Accuracy Between to Adversarially Flip the Label Under Adversarial Attack 60% to 65% Between δ =0.01 to δ =0.08 Changing Intervention: Accuracy of Equal Opportunity Gap Between **Batch Size** Between 5% to 20% the Minority Group 65% to 85% Performance on OOD Settings Between Intervention: None Between 80% to 90% 80% to 90%

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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

Learning Rate

0.1 0.05 0.01 Δ_{max}



94 0

Augmentation Rand; Optimizer SGD; Architecture R18 2

Default Config: Learning Rate 0.1; Batch Size 128; Metric: Accuracy Dataset: UTKFace

Changing Random Seeds Δ_{max}

92

2

92

94 0

	Learning Rate					Batch Size				Augmentation			ptimiz	er	Architecture				
	0.1	0.05	0.01	Δ_{max}	128	256	640	Δ_{max}	Rand	Trivial	Δ_{max}	SGD	Adam	Δ_{max}	R18	R50	WR50	Δ_{max}	
Changing Random Seeds	92.85	92.85	92.37	0.49	92.85	92.81	92.18	0.68	92.85	92.51	0.34	92.85	92.60	0.25	92.85	92.22	92.13	0.72	
	92.89	92.64	92.60	0.30	92.89	92.98	92.62	0.36	92.89	92.89	0.00	92.89	92.55	0.34	92.89	92.49	92.05	0.84	
	92.47	92.51	92.43	0.08	92.47	92.66	92.60	0.19	92.47	92.87	0.40	92.47	92.68	0.21	92.47	92.24	92.45	0.23	
	93.17	92.91	92.64	0.53	93.17	93.08	92.68	0.49	93.17	92.79	0.38	93.17	92.87	0.30	93.17	92.05	92.20	1.12	
	92.60	92.98	92.66	0.38	92.60	92.32	92.07	0.53	92.60	92.87	0.27	92.60	92.45	0.15	92.60	92.30	92.18	0.42	
Δ_{max}	0.70	0.46	0.30		0.70	0.76	0.61		0.70	0.38		0.70	0.42		0.70	0.44	0.40		

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

Metric: Accuracy Dataset: UTKFace

2

92

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	92.89	92.64	92.60	0.30	92.89	92.98	92.62	0.36	92.89	92.89	0.00	92.89	92.55	0.34	92.89	92.49	92.05	0.84	
	92.47	92.51	92.43	0.08	92.47	92.66	92.60	0.19	92.47	92.87	0.40	92.47	92.68	0.21	92.47	92.24	92.45	0.23	
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Default Config: Learning Rate 0.1; Batch Size 128;

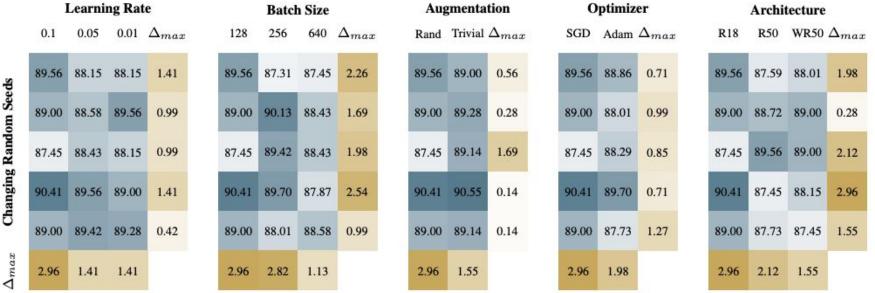
Augmentation Rand; Optimizer SGD; Architecture R18

Metric: Accuracy Dataset: UTKFace Δ_m^{al}

 Δ^{all}_{max} : 1.12

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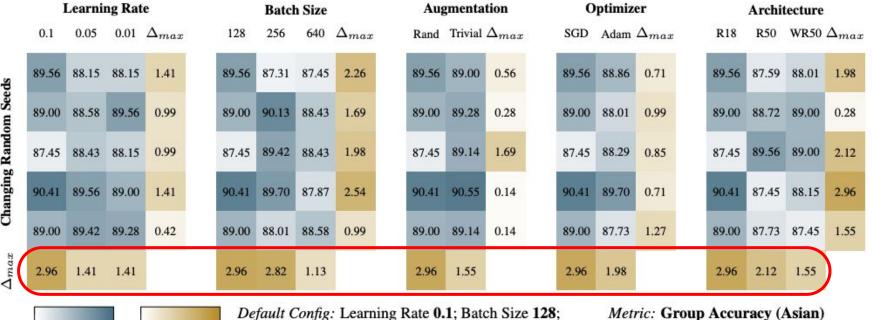
Multiplicity Sheet: Fairness

4

87 91 0 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

Changing Random Seeds



Multiplicity Sheet: Fairness

Changing Random Seeds

87

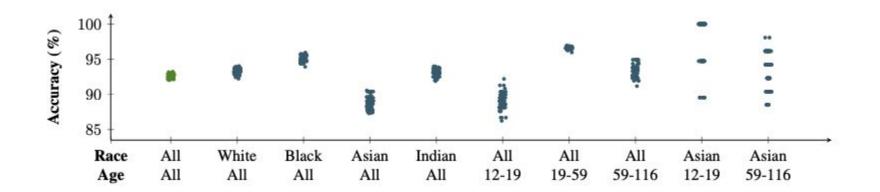
91

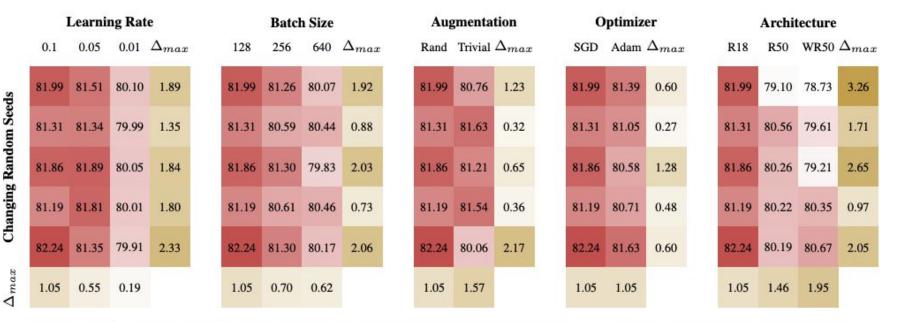
0

Augmentation Rand; Optimizer SGD; Architecture R18 4

Dataset: UTKFace Δ_{max}^{all} : 3.24

Intersectionality and Multiplicity



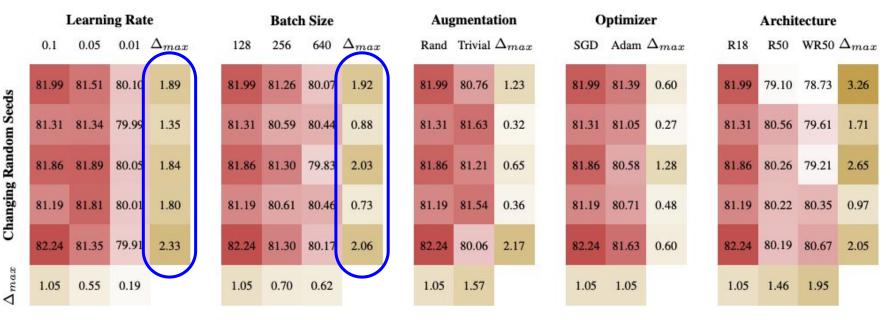


4

79 83 0

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

Metric:OOD Accuracy (FairFace)Dataset:UTKFace Δ_{max}^{all} :3.51



79

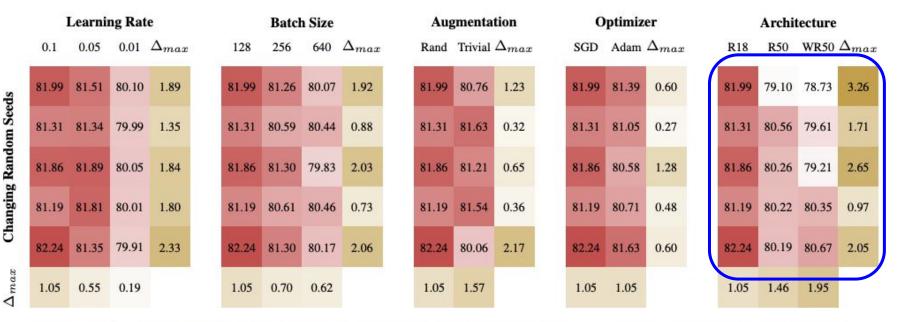
83

0

4

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

Metric: OOD Accuracy (FairFace)Dataset: UTKFace Δ_{max}^{all} : 3.51

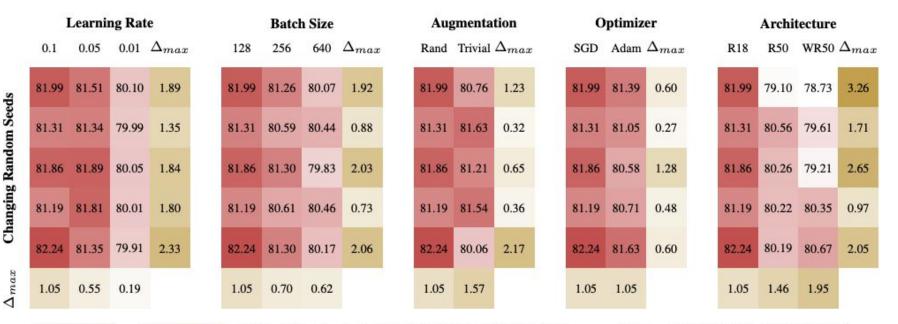


4

79 83 0

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

Metric:OOD Accuracy (FairFace)Dataset:UTKFace Δ_{max}^{all} :3.51



79

83

0

4

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

Metric: **OOD Accuracy (FairFace)** *Dataset:* **UTKFace** Δ^{all}_{max} : **3.51**

Multiplicity Sheet: Privacy



84 89 0 6

Default Config: Learning Rate 0.1; Batch Size 128; Augmentation Rand; Optimizer SGD; Architecture R18 Metric: Pert. Accuracy ($\lambda = 5$) Dataset: UTKFace Δ_{max}^{all} : 5.30

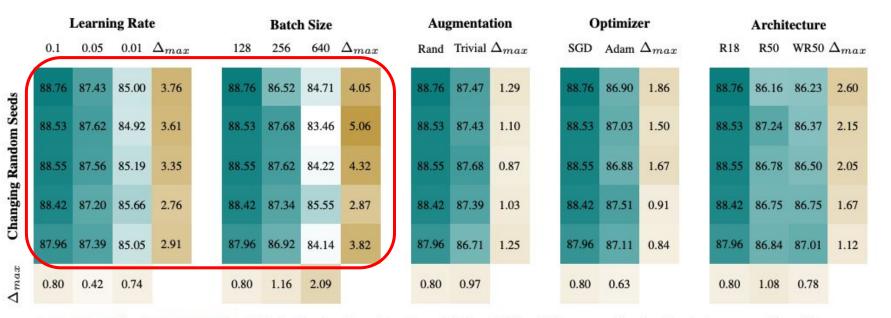
Multiplicity Sheet: Privacy

6

84

89

0



Default Config: Learning Rate 0.1; Batch Size 128; Augmentation Rand; Optimizer SGD; Architecture R18 *Metric:* Pert. Accuracy ($\lambda = 5$) *Dataset:* UTKFace Δ_{max}^{all} : 5.30

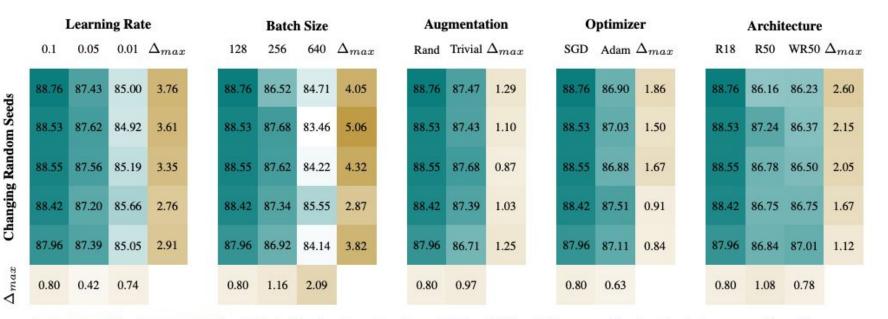
Multiplicity Sheet: Privacy

6

84

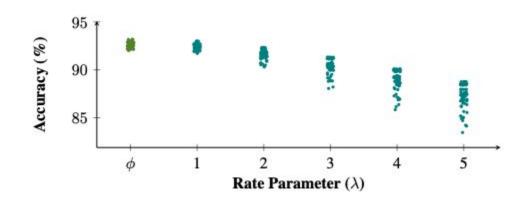
89

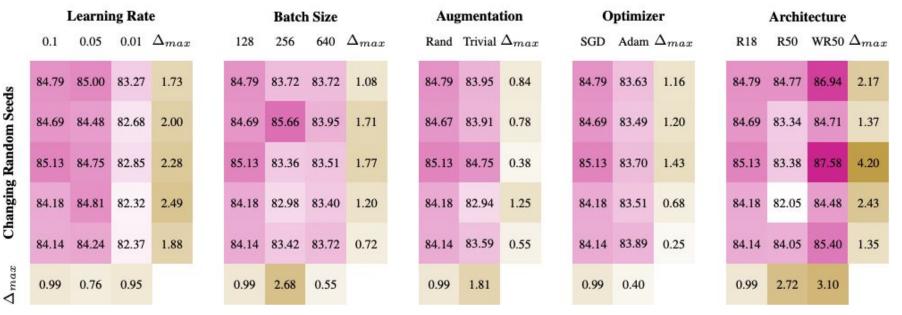
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Default Config: Learning Rate 0.1; Batch Size 128; Augmentation Rand; Optimizer SGD; Architecture R18 *Metric:* Pert. Accuracy ($\lambda = 5$) *Dataset:* UTKFace Δ_{max}^{all} : 5.30

Accuracy Under Intervention: Privacy



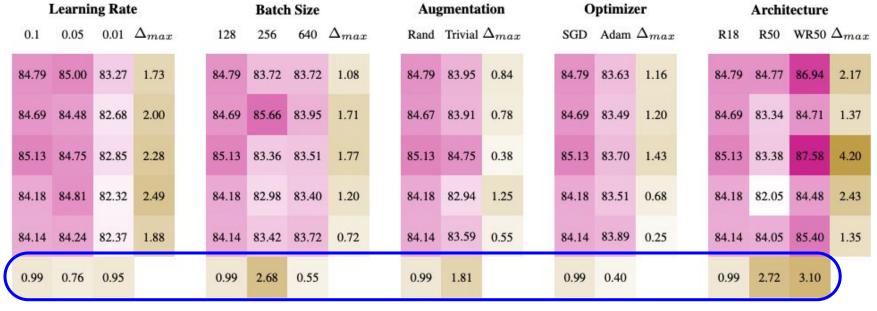


5

82 88 0 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

Changing Random Seeds



5

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

Changing Random Seeds

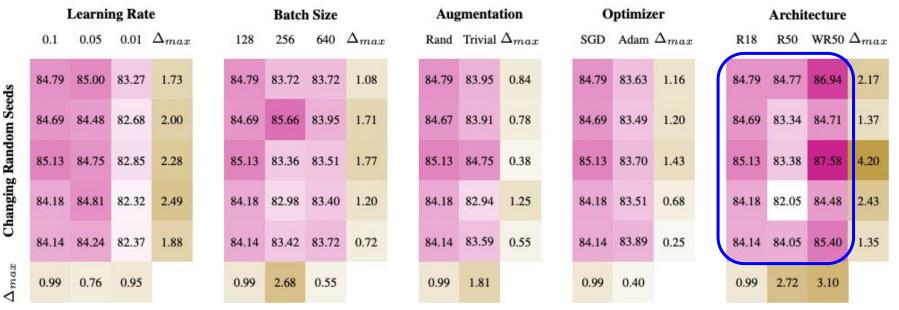
max

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88

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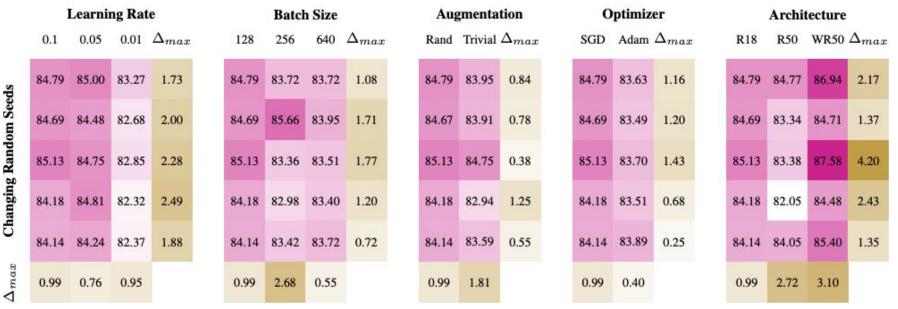


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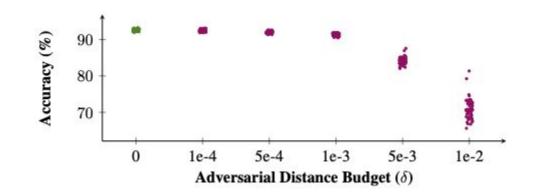
5

82 88 0 Default Config: Learning Rate 0.1; Batch Size 128; Augmentation Rand; Optimizer SGD; Architecture R18

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

Changing Random Seeds

Accuracy Under Intervention: Security



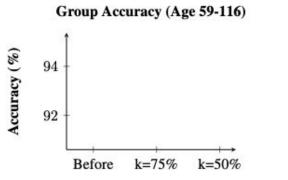
Contributions

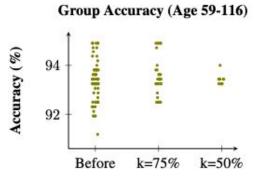
- A standardized framework to measure and study model multiplicity.
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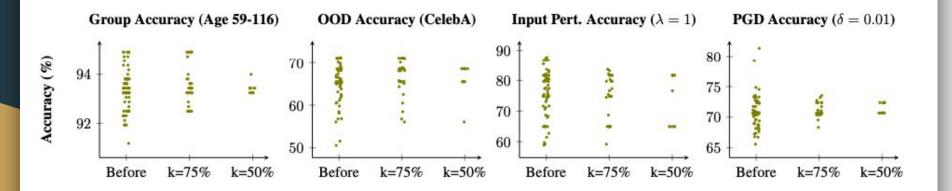
- We collect all 45 models in previous multiplicity sheets.
- We only select models which rank in the top k% of each metric, i.e.,
 - Accuracy for racial group 'Asian'
 - Accuracy on OOD dataset 'FairFace'
 - Accuracy under Output Perturbations for Privacy
 - Accuracy under PGD Adversarial Attacks δ =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 δ=0.01







• We created an empirical language to talk about multiplicity (accuracy under intervention and multiplicity sheets)

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- We performed a detailed case study and benchmarked model multiplicity of various trustworthy ML metrics for image classification.

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- We showed empirically that the concerns of model multiplicity persist even beyond model selection.