SINGAPORE UNIVERSITY OF TECHNOLOGY AND DESIGN

NEURAL INFORMATION PROCESSING SYSTEMS

Label-Only Model Inversion Attacks via Knowledge Transfer

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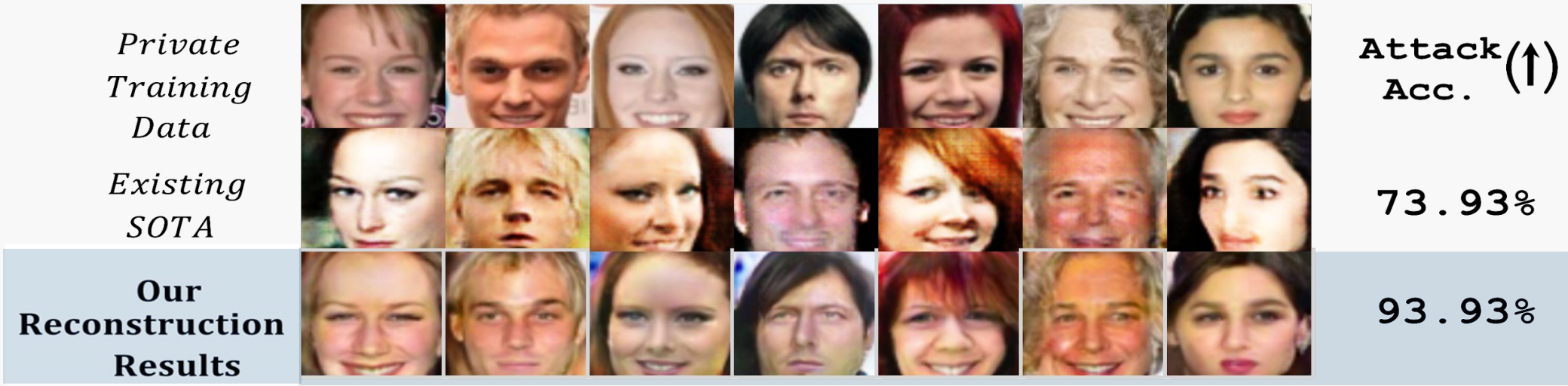
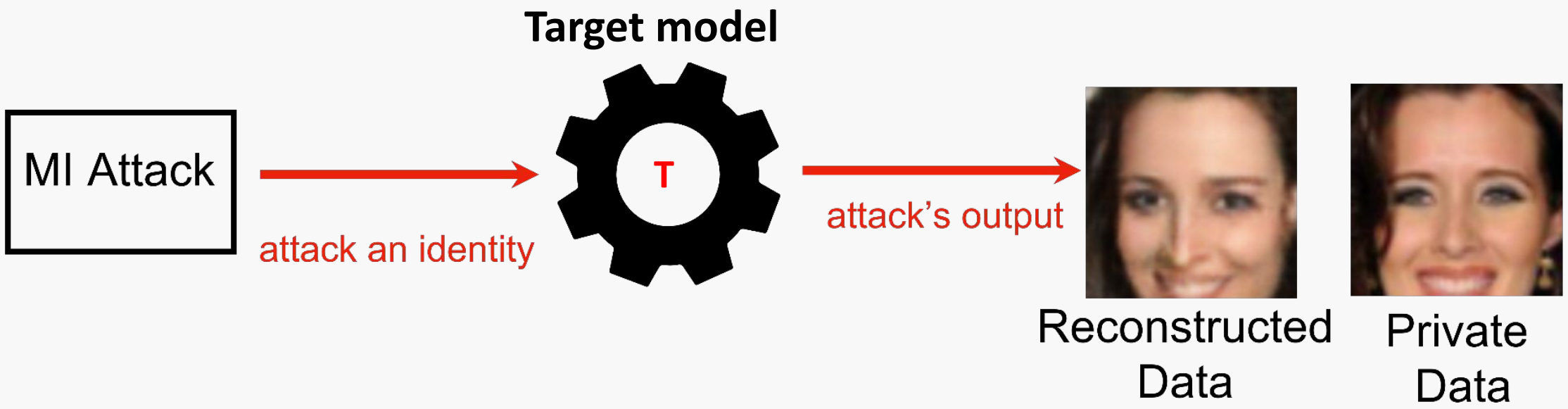
Our contributions

- We propose Label-only Model inversion via Knowledge Transfer (LOKT), a new label-only MI by transferring decision knowledge from the target model to surrogate models and performing white-box attacks on the surrogate models.
- We propose a new T-ACGAN to leverage generative modeling and the target model for effective knowledge transfer.
- We perform analysis to support that our surrogate models are effective proxies for the target model for MI.



Model Inversion (MI)

Model inversion (MI) attacks aim to infer and reconstruct private training data by abusing access to a model.

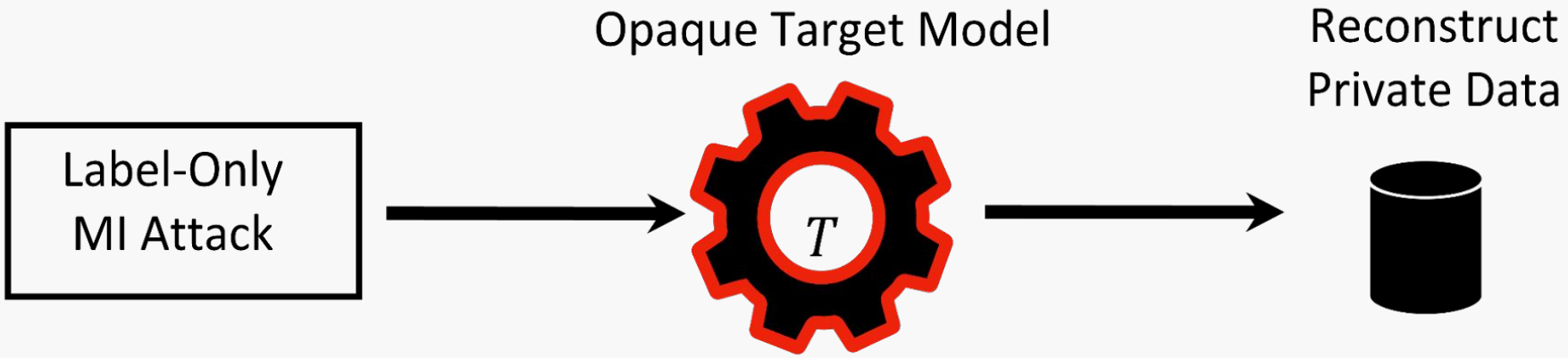


Model Inversion (MI)

We focus on label-only model inversion attack which is the most challenging MI Attack.

Criteria	Architecture / Parameters	Soft-labels	Hard-labels	Concern reg. Queries
White-box	✓	✓	✓	Low
Black-box	✗	✓	✓	High
Label-only	✗	✗	✓	High

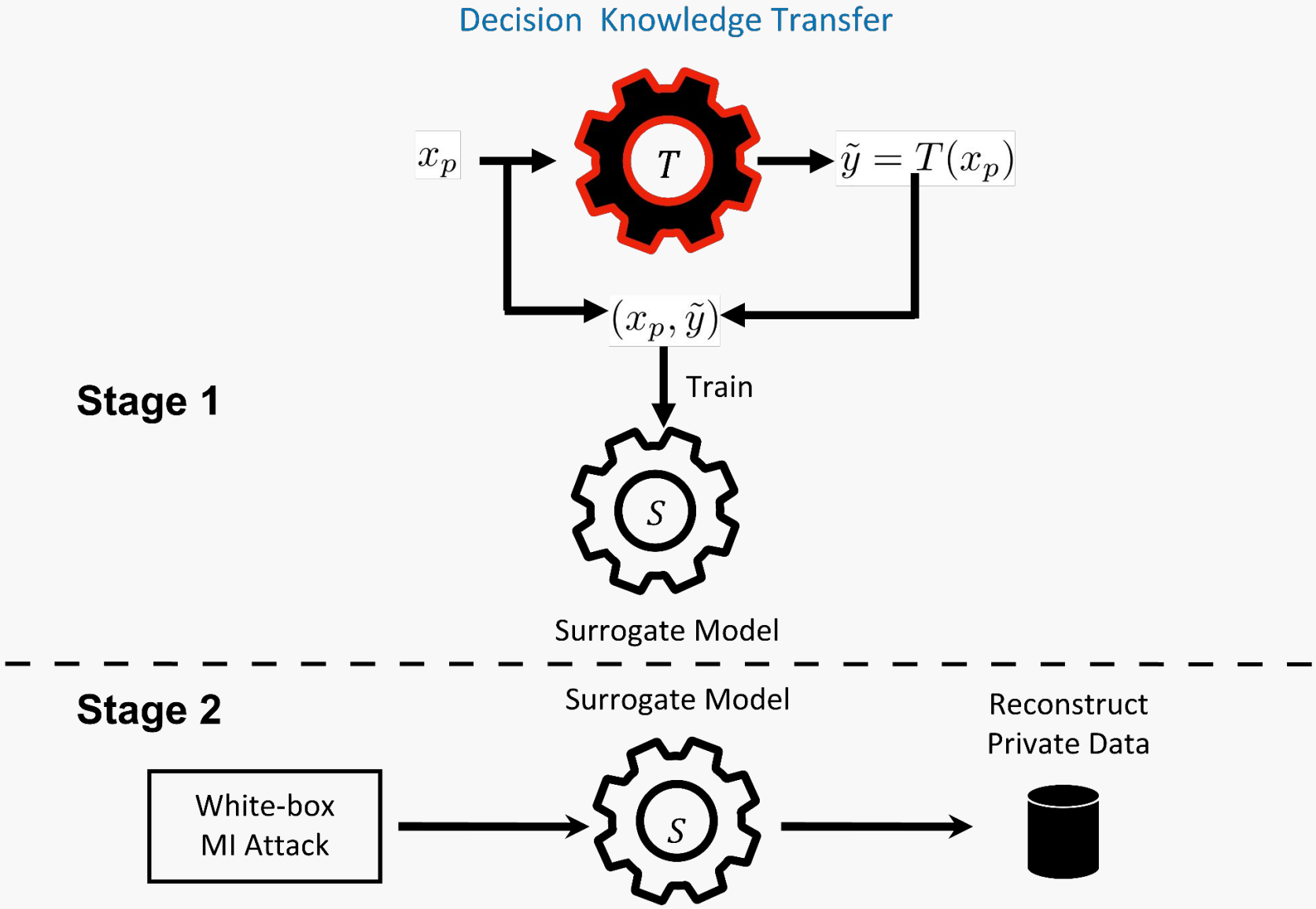
Existing work on Label-only Model Inversion Attack



SOTA Label-only Model Inversion attacks employ **black-box search on the target model T** to reconstruct training data.

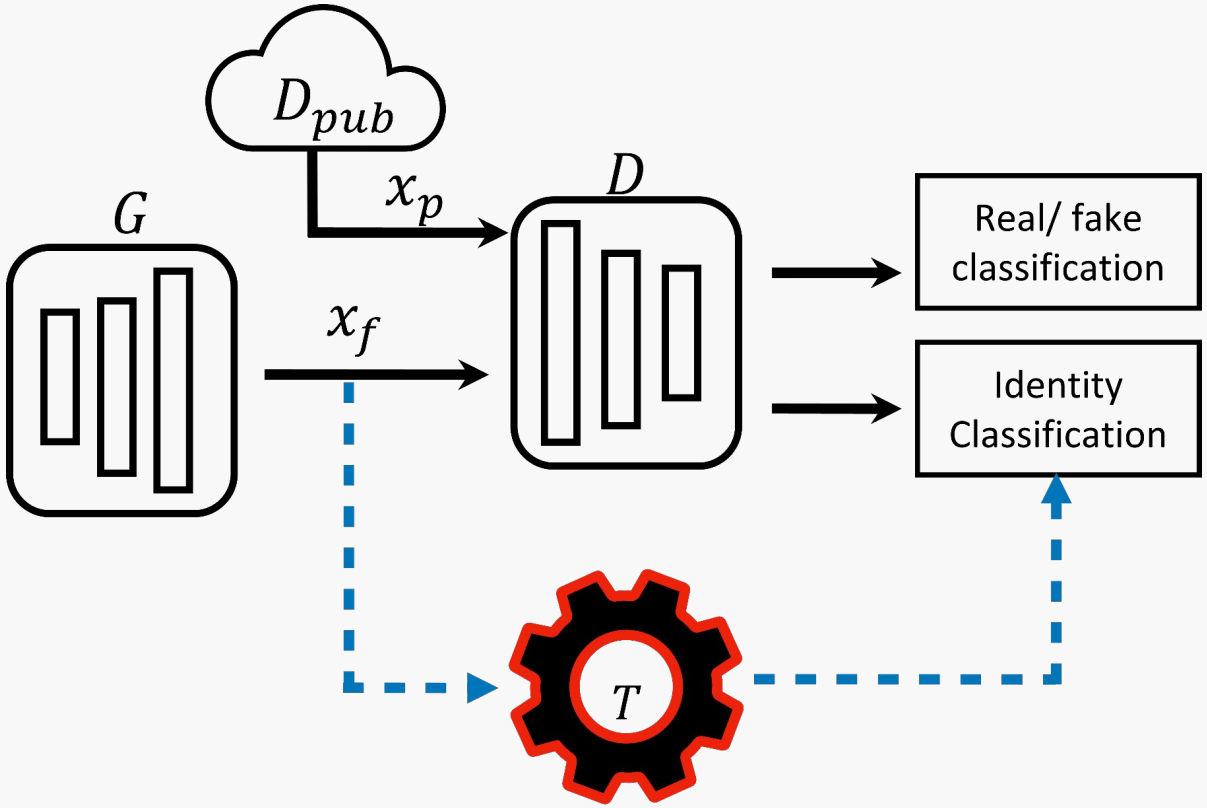
Mostafa et. al. Label-only model inversion attacks via boundary repulsion. In CVPR 2022.

Label-only Model inversion via Knowledge Transfer (LOKT)



Casting Label-only MI Attack as a White-box MI Attack

Decision Knowledge Transfer using our T-ACGAN



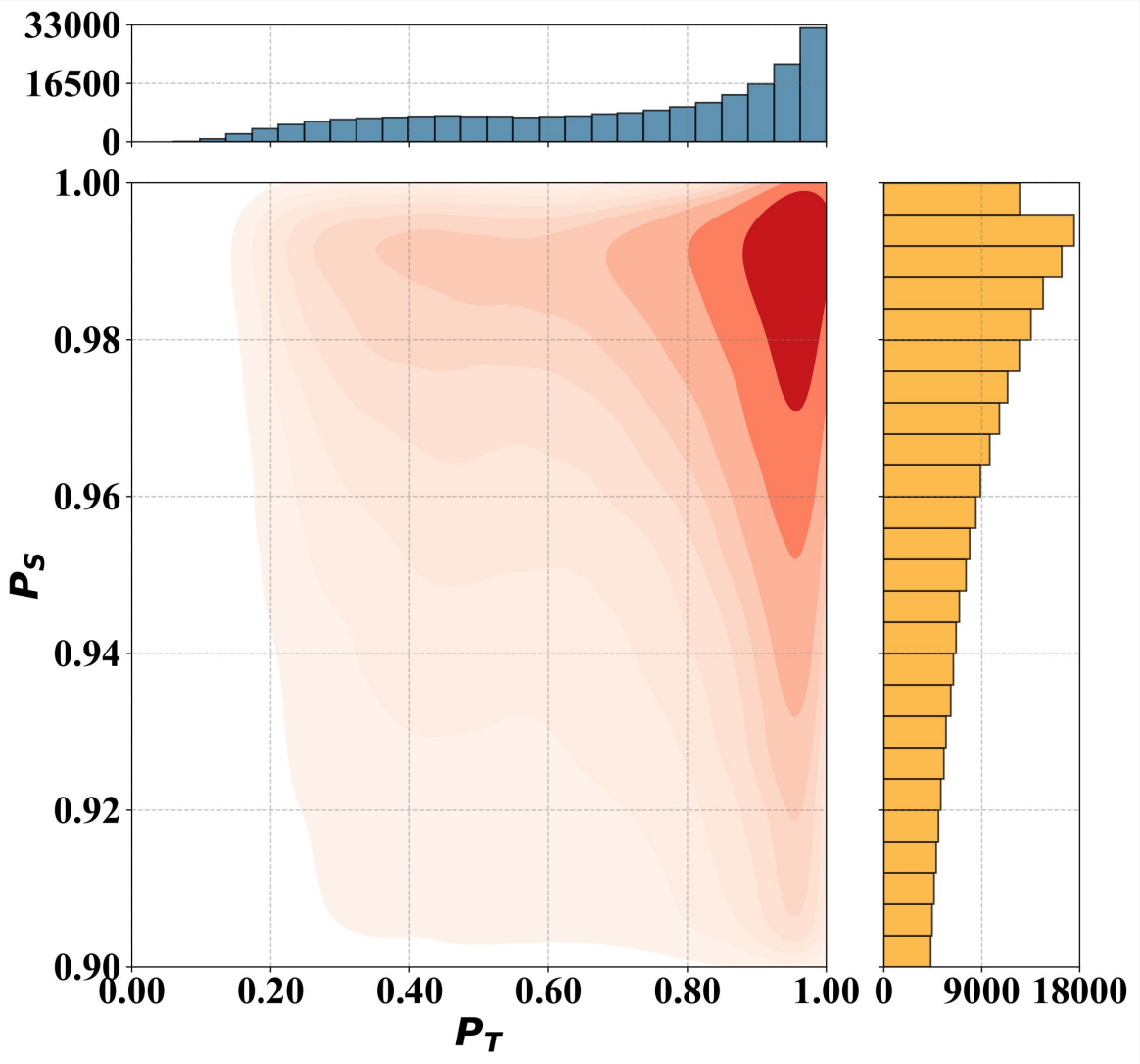
Decision Knowledge Transfer

$$\mathcal{L}_{D,C} = -E[\log P(s = Fake|x_f)] - E[\log P(s = Real|x_p)] - E[\log P(c = \tilde{y}|x_f)]$$

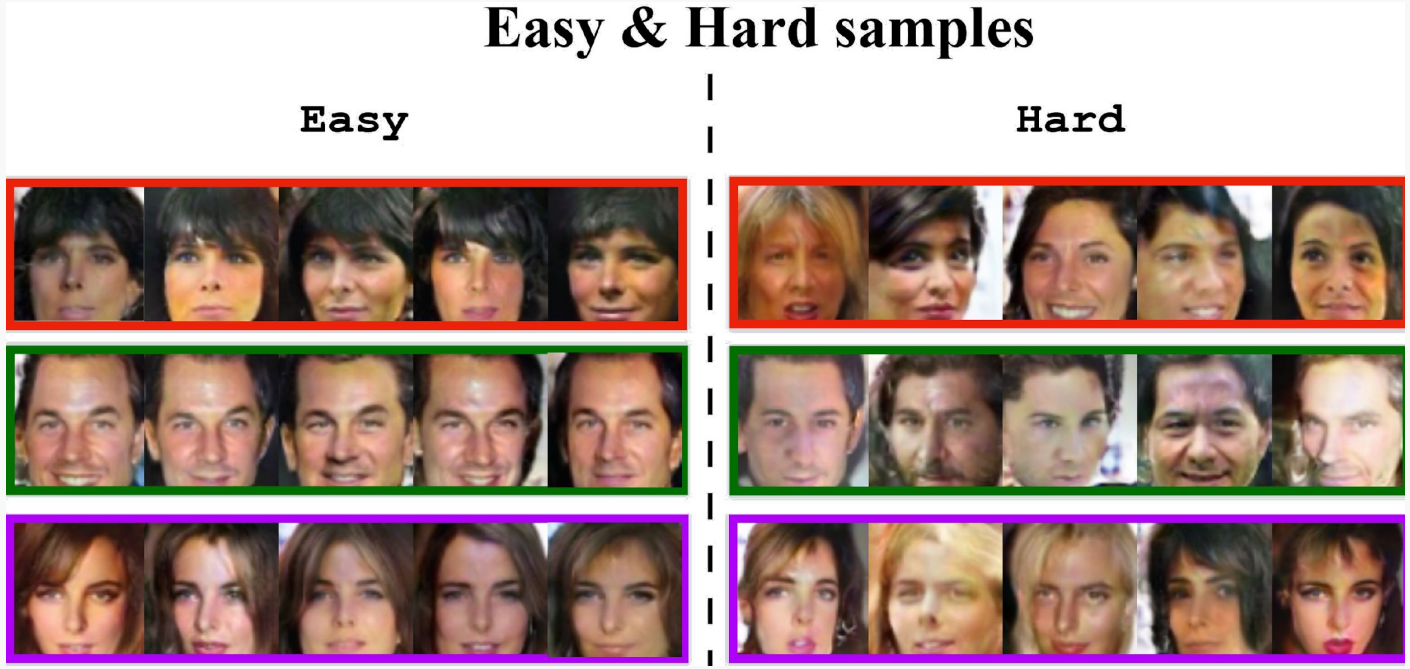
Analysis for justification of surrogate models

Property P1:

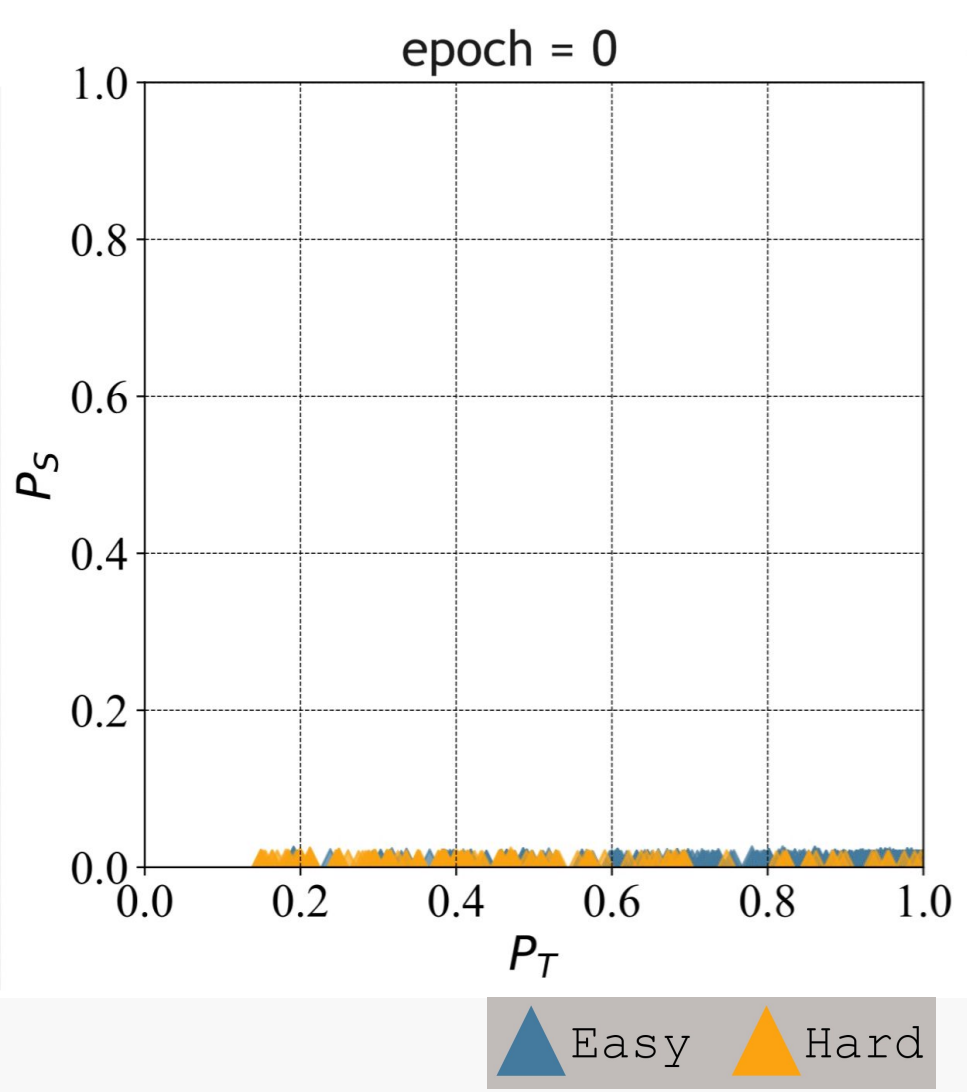
For high-likelihood samples under S, it is likely that they also have high likelihood under T.



Analysis for justification of surrogate models



DNNs Learn Patterns First

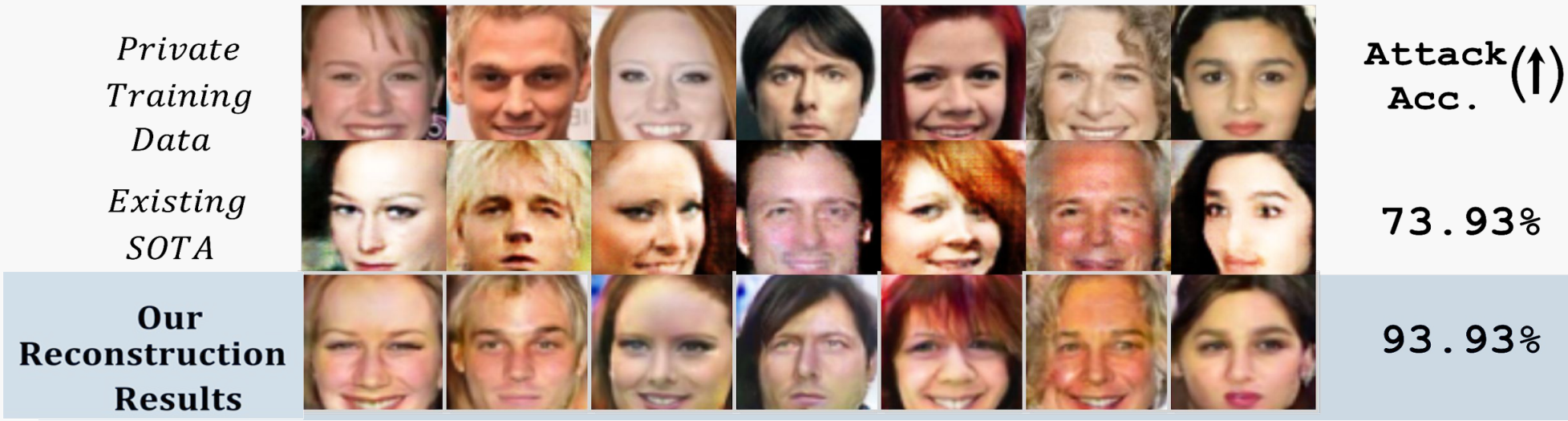


Devansh et.al. A closer look at memorization in deep networks. In ICML, 2017

Model inversion attack results

Setup	Attack		Attack acc. ↑	KNN dt. ↓
T = FaceNet64	BREPMI		73.93 ± 4.98	1284.41
$\mathcal{D}_{priv}$ = CelebA	LOKT	$C \circ D$	81.00 ± 4.79	1298.63
$\mathcal{D}_{pub}$ = CelebA		S	92.80 ± 2.59	1207.25
		S_{en}	$\mathbf{93.93 \pm 2.78}$	**1181.72**
T = IR152	BREPMI		71.47 ± 5.32	1277.23
$\mathcal{D}_{priv}$ = CelebA	LOKT	$C \circ D$	72.07 ± 4.03	1358.94
$\mathcal{D}_{pub}$ = CelebA		S	89.80 ± 2.33	1220.00
		S_{en}	$\mathbf{92.13 \pm 2.06}$	**1206.78**

Setup	Attack		Attack acc. ↑	KNN dt. ↓
T = VGG16	BREPMI		57.40 ± 4.92	1376.94
$\mathcal{D}_{priv}$ = CelebA	LOKT	$C \circ D$	71.33 ± 4.39	1364.47
$\mathcal{D}_{pub}$ = CelebA		S	85.60 ± 3.03	1252.09
		S_{en}	$\mathbf{87.27 \pm 1.97}$	**1246.71**
T = FaceNet64	BREPMI		43.00 ± 5.14	1470.55
$\mathcal{D}_{priv}$ = CelebA	LOKT	$C \circ D$	43.27 ± 3.53	1516.18
$\mathcal{D}_{pub}$ = FFHQ		S	59.13 ± 2.77	1437.86
		S_{en}	$\mathbf{62.07 \pm 3.89}$	**1428.04**



Conclusion

- We propose Label-only Model inversion via Knowledge Transfer (LOKT), a new label-only MI by transferring decision knowledge from the target model to surrogate models and performing white-box attacks on the surrogate models.
- We propose a new T-ACGAN to leverage generative modeling and the target model for effective knowledge transfer.
- We perform analysis to support that our surrogate models are effective proxies for the target model for MI.

Thank you!

Poster Session
Wed 13 Dec 10:45 a.m. CST — 12:45 p.m. CST
Great Hall & Hall B1+B2
#1517

Project page

https://ngoc-nguyen-0.github.io/lokt/