



# Label-Only Model Inversion Attacks via Knowledge Transfer

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#### **Our contributions**

- We propose Label-only Model inversion via Knowledge Transfer (LOKT) 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.







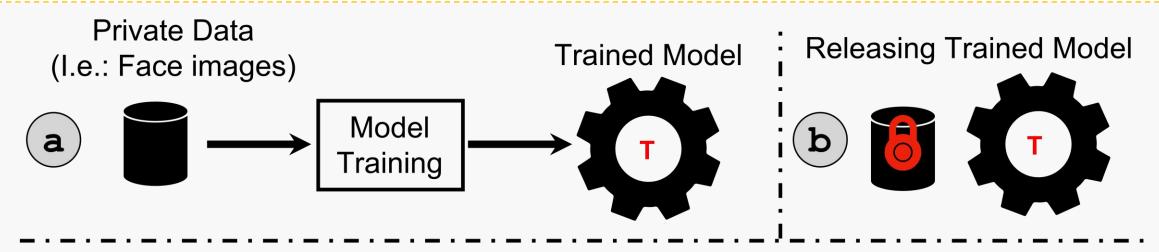
Attack (1)

73.93%

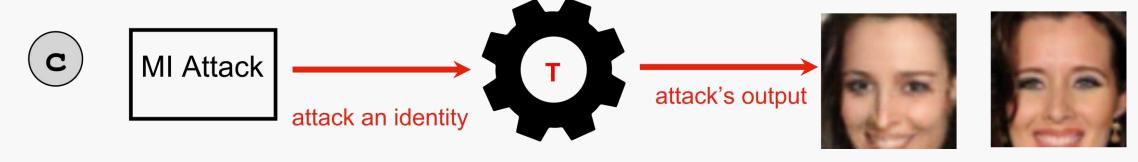
93.93%

#### **Model Inversion (MI)**

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



Model Inversion (MI) attack on Target Model to recover Private Training Data

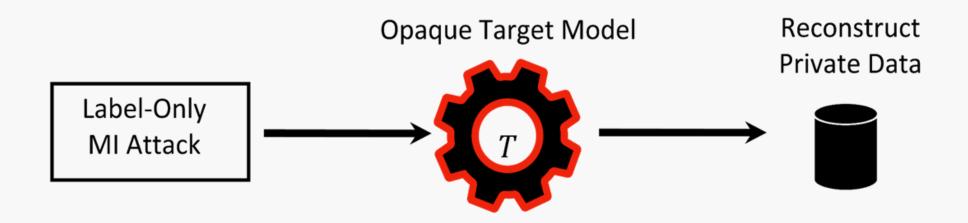


#### **Model Inversion (MI)**

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

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

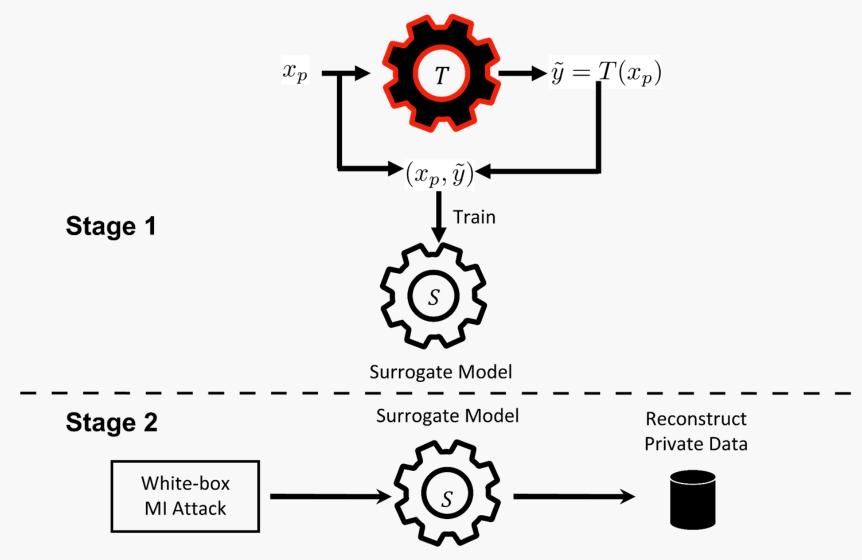
## **Existing work on Label-only Model Inversion Attack**



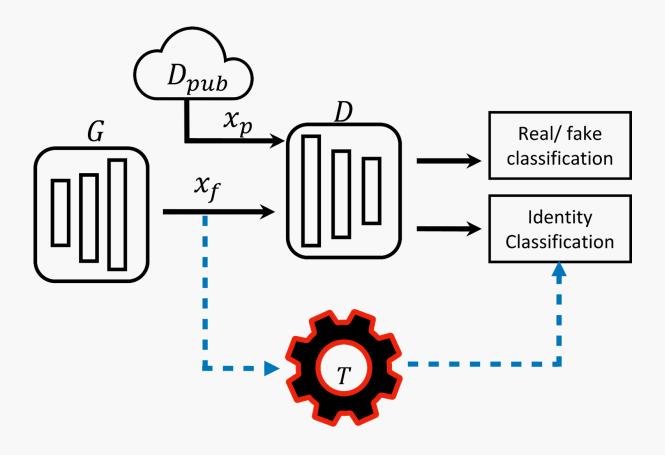
SOTA Label-only Model Inversion attacks employ **black-box search on the target model T** to reconstruct private data.

# Label-only Model inversion via Knowledge Transfer (LOKT)

Decision Knowledge Transfer



## **Decision Knowledge Transfer using our T-ACGAN**



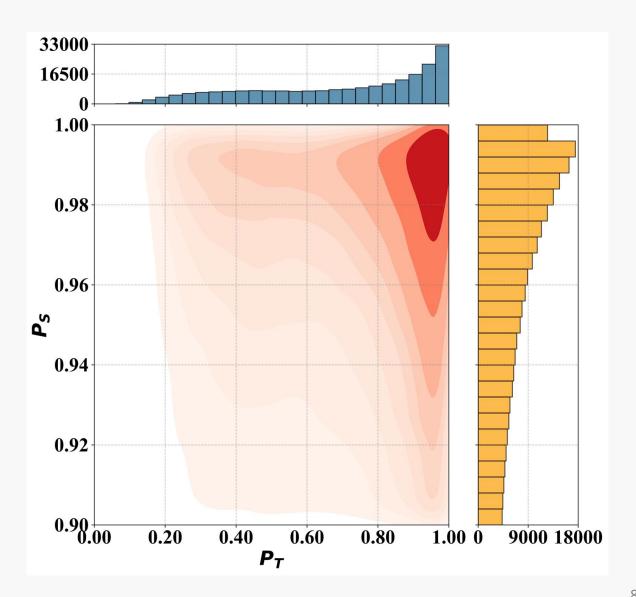
**Decision Knowledge Transfer** 

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

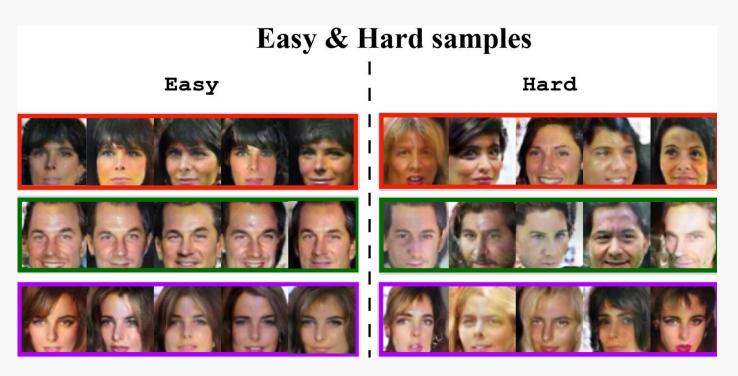
# **Analysis for justification of surrogate models**

#### **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



epoch = 01.0 0.8  $0.6^{-}$ Ps  $0.4^{-1}$ 0.2  $0.0^{+}_{0.0}$ 0.2 0.4 0.6 0.8 1.0 Easy Hard

**DNNs Learn Patterns First** 

#### **Model inversion attack results**

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

Setup	Attack		Attack acc. ↑	KNN dt. ↓
T = VGG16	BREPMI		$57.40 \pm 4.92$	1376.94
$\mathcal{D}_{priv} = \text{CelebA}$	LOKT	$C \circ D$	$71.33 \pm 4.39$	1364.47
$\mathcal{D}_{pub} = \text{CelebA}$		S	$85.60 \pm 3.03$	1252.09
		$S_{en}$	$87.27 \pm 1.97$	1246.71
T = FaceNet64	BREPMI		$43.00\pm5.14$	1470.55
$\mathcal{D}_{priv}$ = CelebA		$C \circ D$	$43.27 \pm 3.53$	1516.18
$\mathcal{D}_{pub}^{Pred} = FFHQ$	LOKT	S	$59.13 \pm 2.77$	1437.86
P		$S_{en}$	$\textbf{62.07} \pm \textbf{3.89}$	1428.04





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73.93%

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#### Conclusion

- We propose Label-only Model inversion via Knowledge Transfer
   (LOKT) 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/