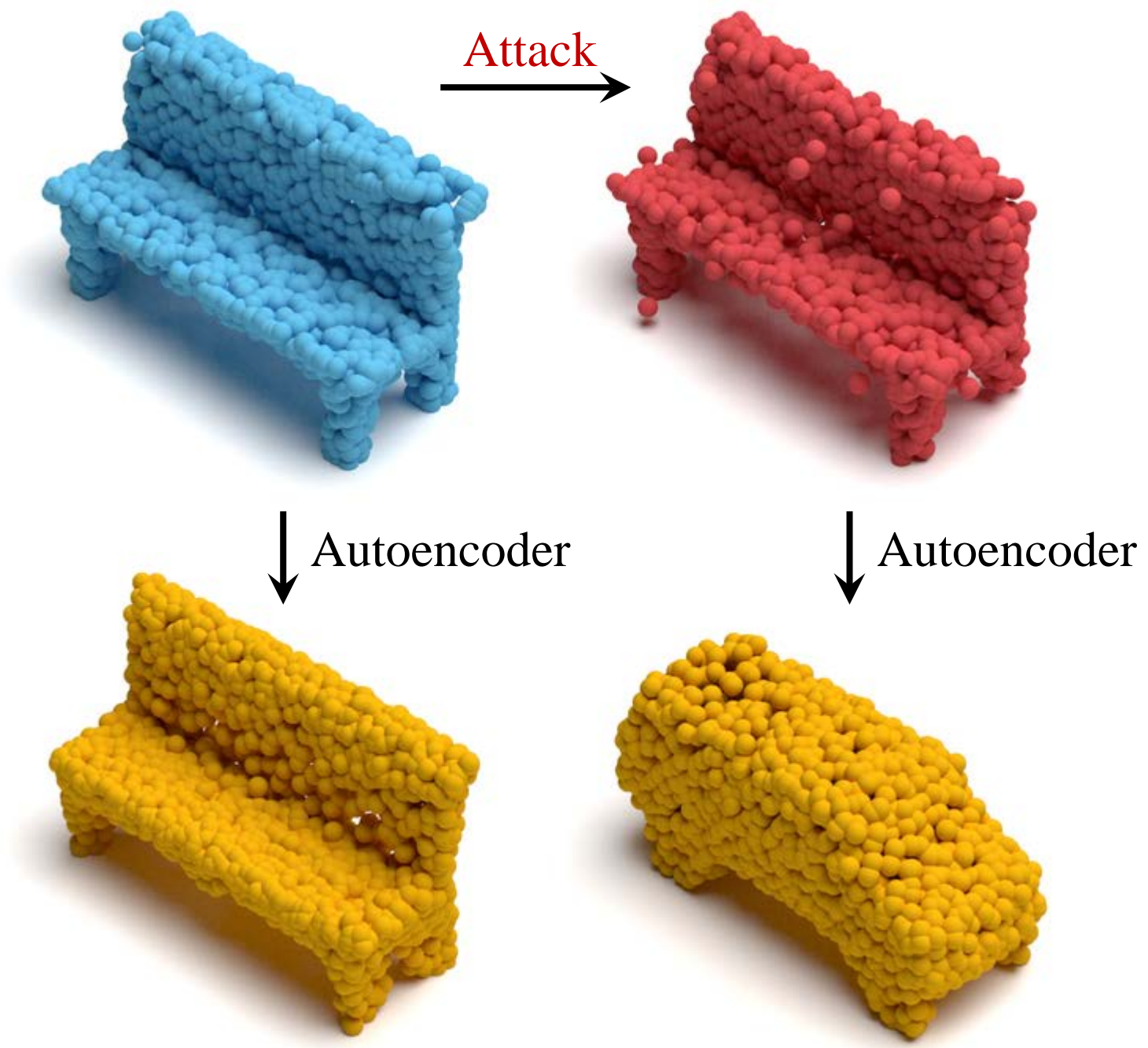


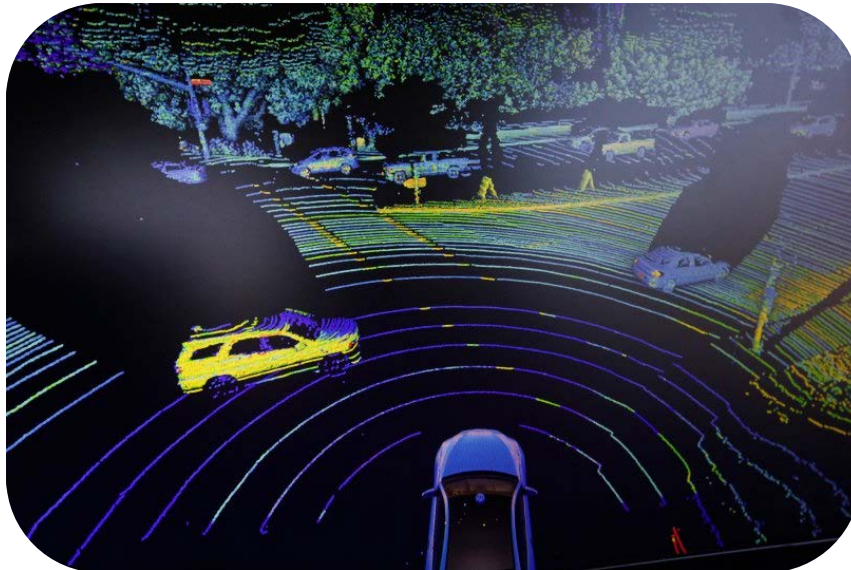
Geometric Adversarial Attacks and Defenses on 3D Point Clouds

Itai Lang, Uriel Kotlicki, Shai Avidan





3D & AI in Safety-critical Systems

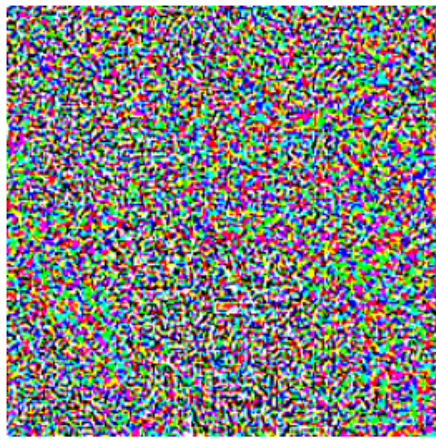


Neural Networks are **Vulnerable!**



Panda

$+ .007 \times$

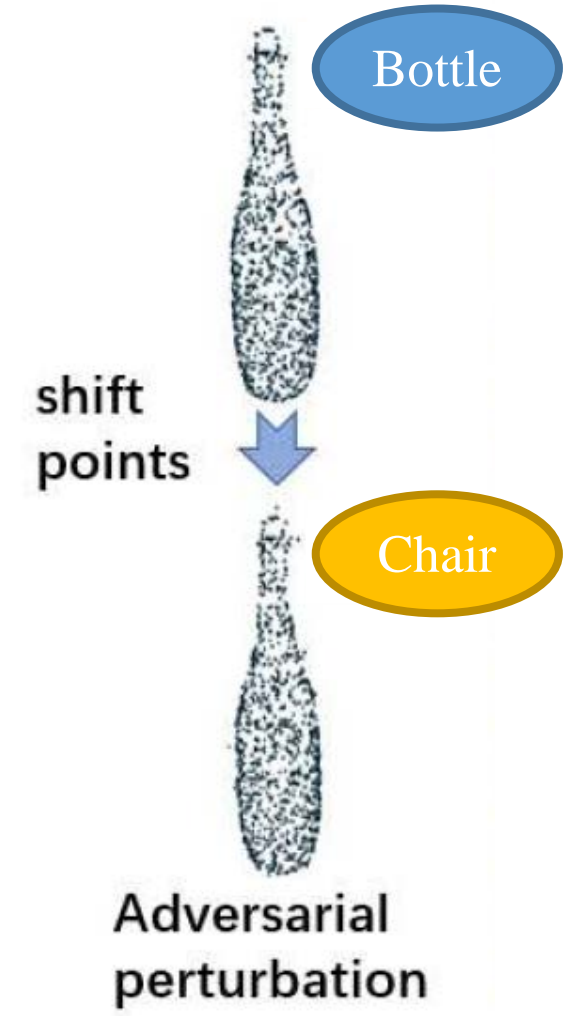


Perturbation

=



Gibbon



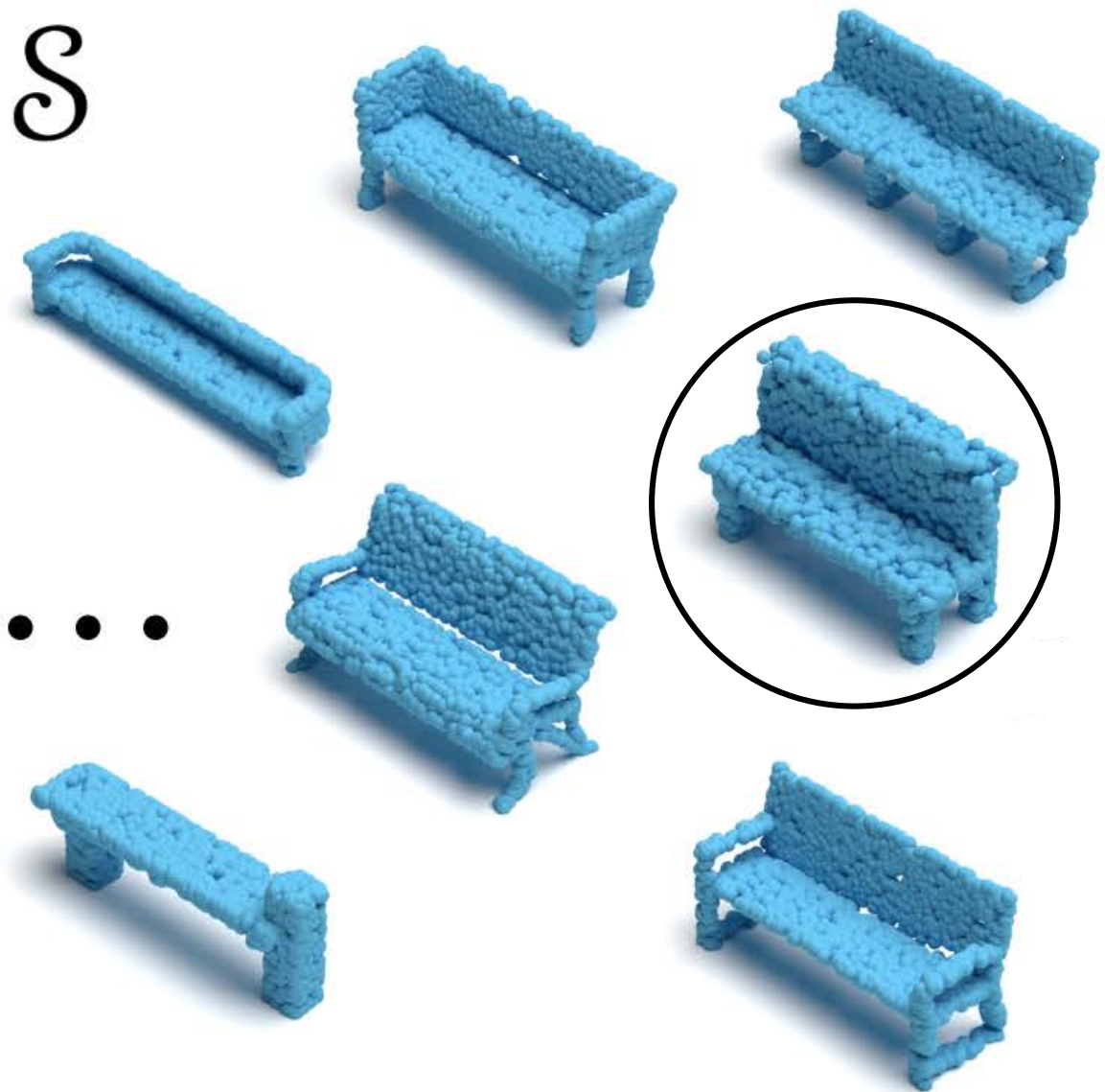
Goodfellow *et al.*, 2015

Xiang *et al.*, 2019

Can we make a small **perturbation** to
an input **point cloud** to change the
reconstructed **geometry** by an
autoencoder model?

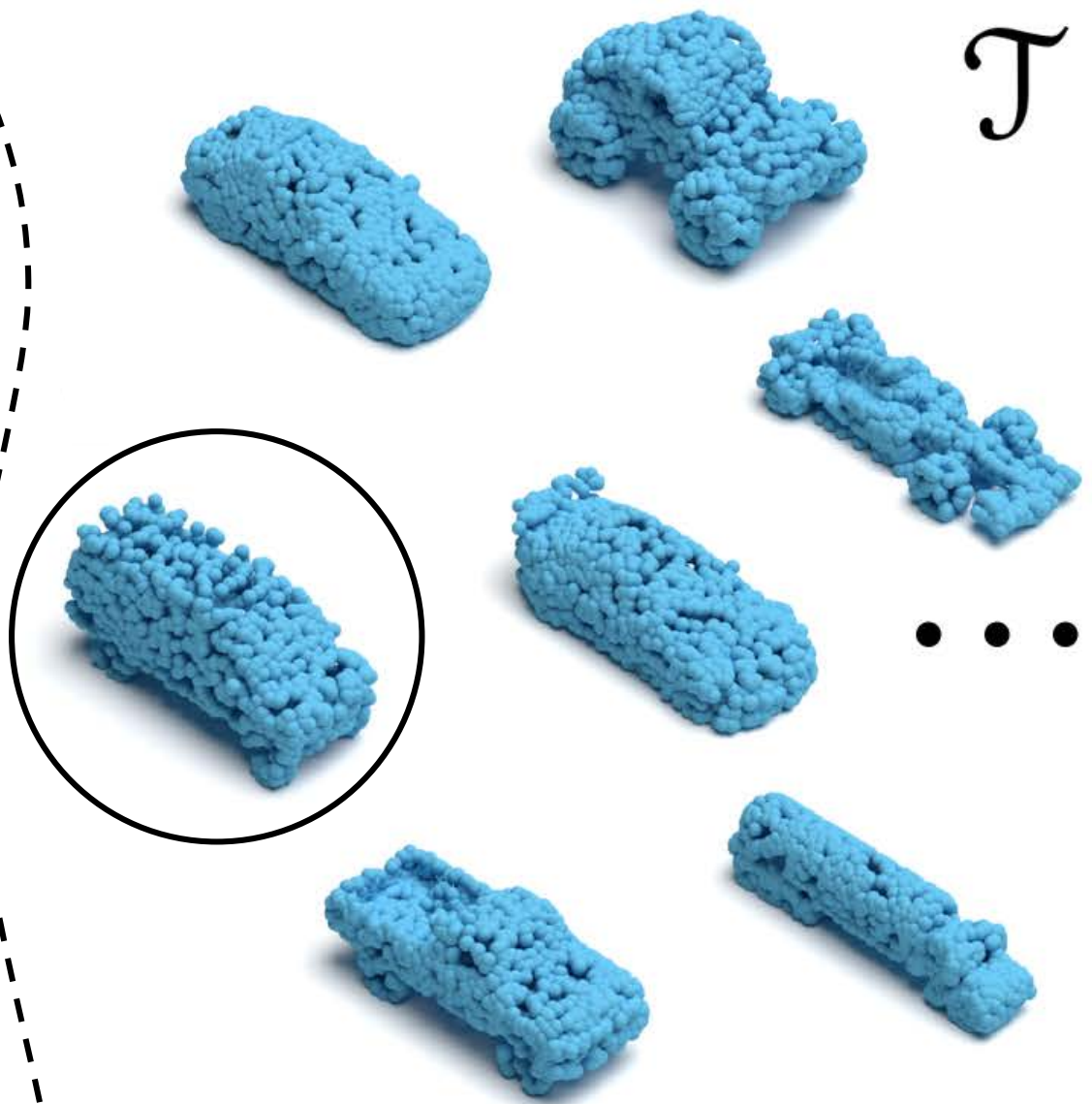
Problem Statement

\mathcal{S}



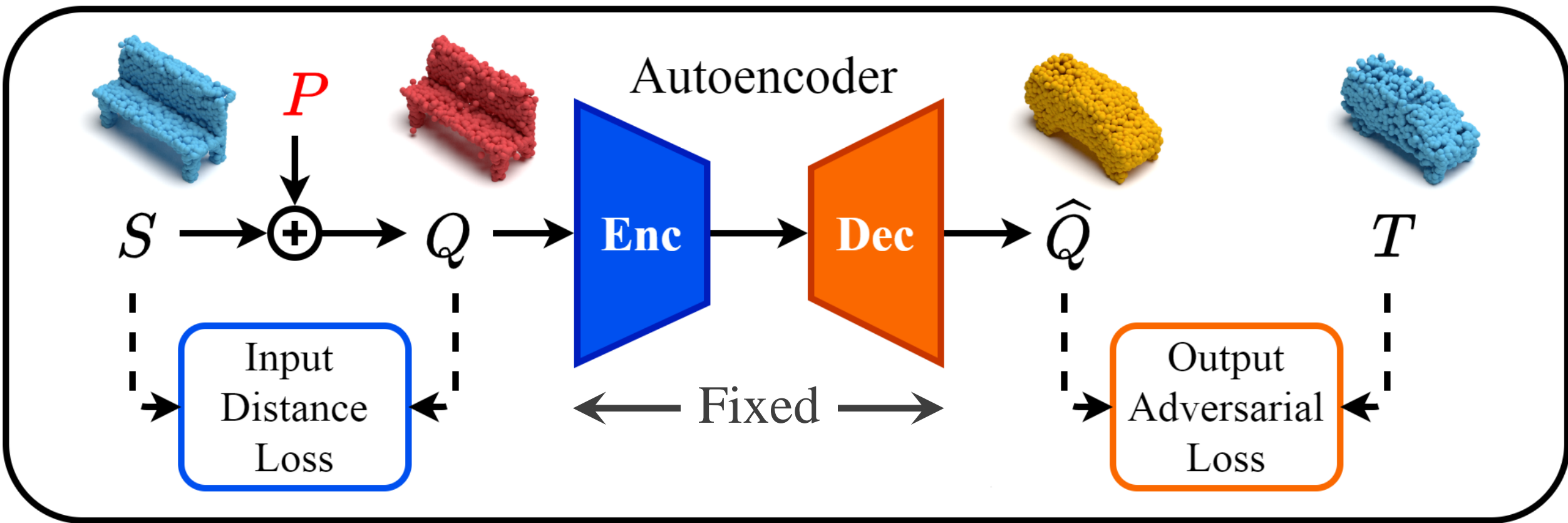
...

\mathcal{T}



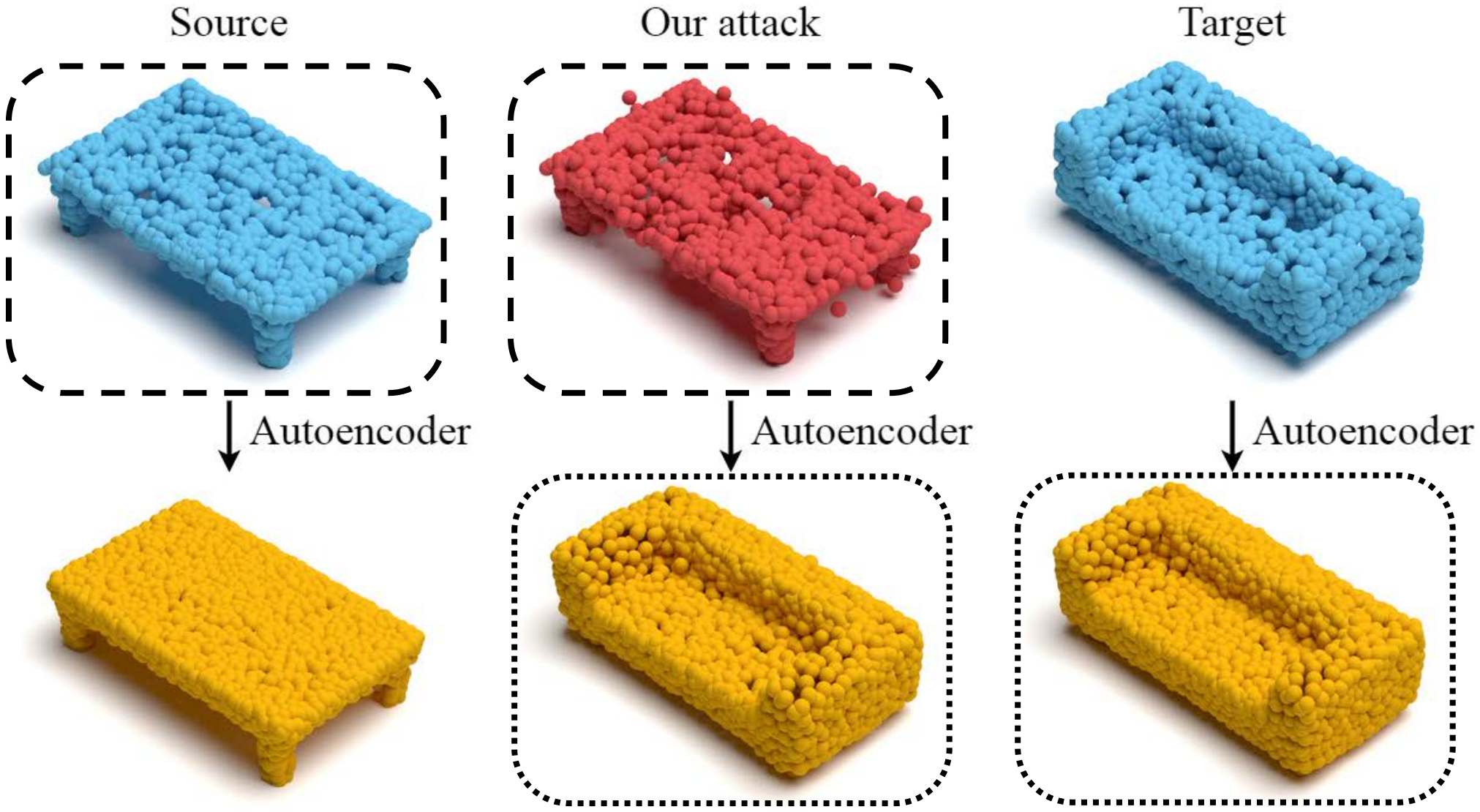
...

The Proposed Attack

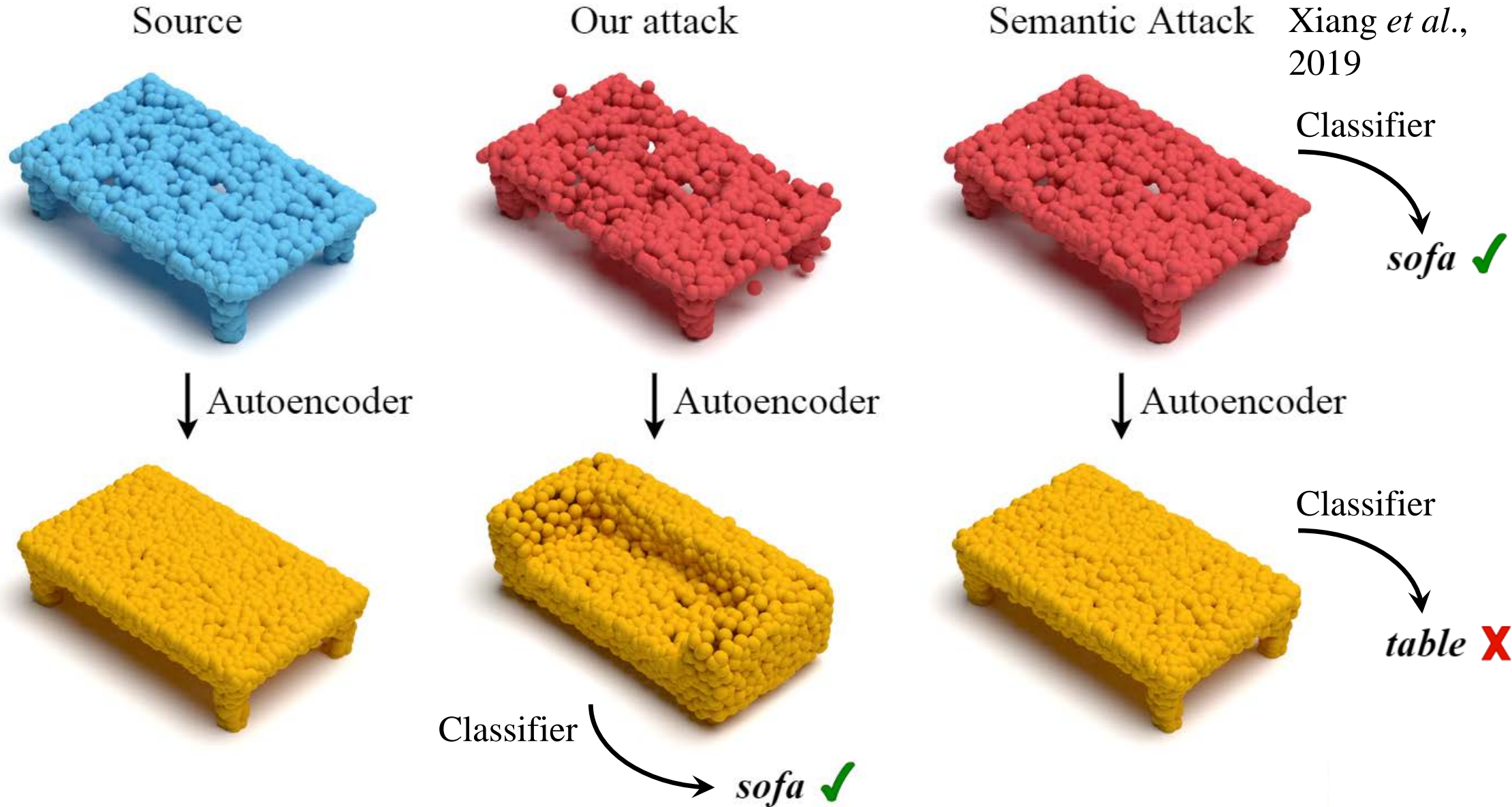


The attacked autoencoder is from the work of Achlioptas *et al.*, 2018

Attack Results

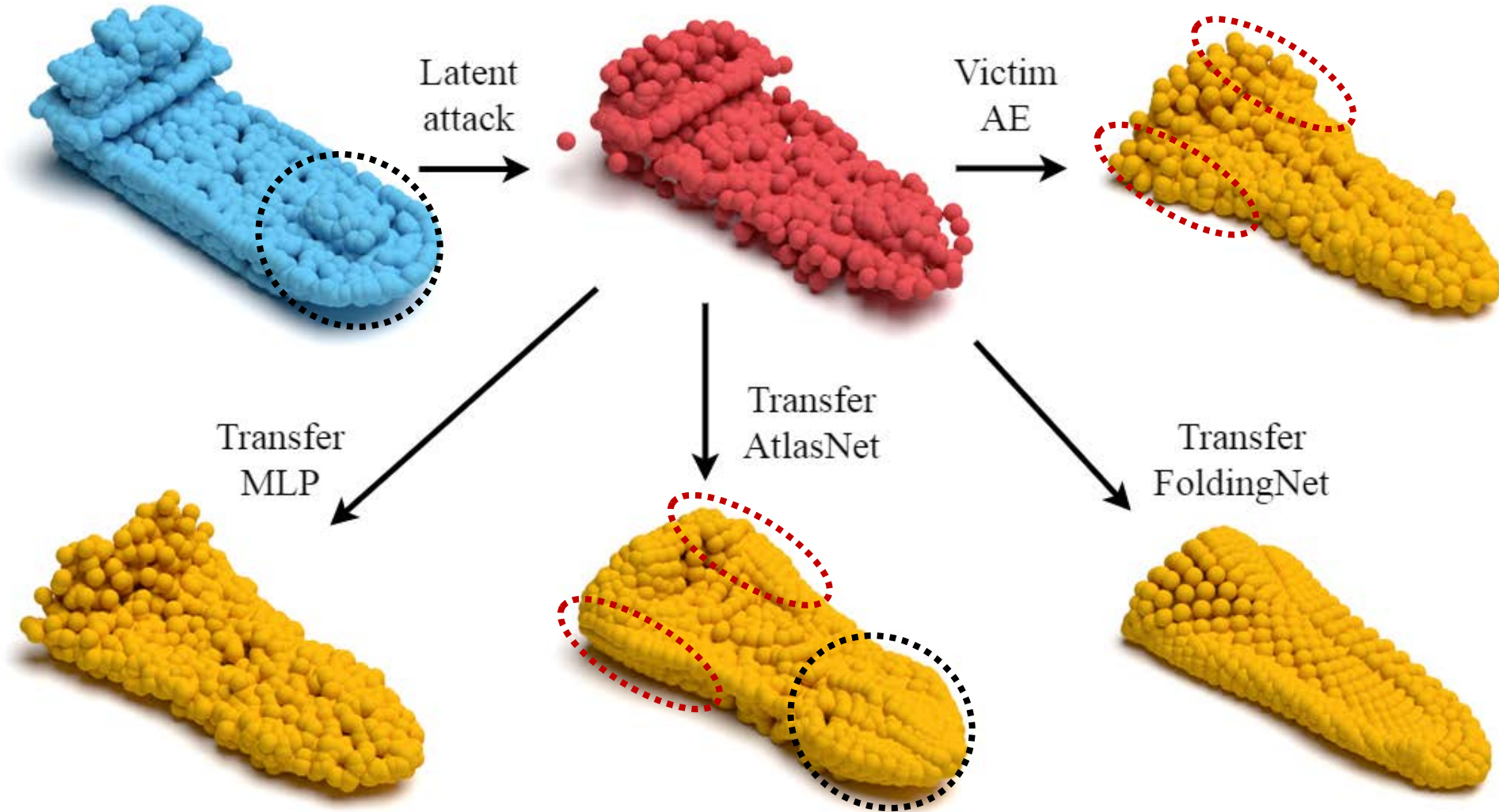


Attack Comparison



Input type	Hit Target	Avoid Source
Our attack	76.0%	94.7%
Semantic attack	1.0%	9.6%

Attack Transfer



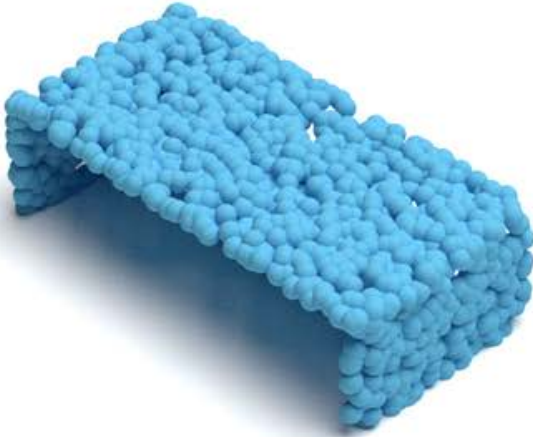
MLP AE, Achlioptas *et al.*, 2018

AtlasNet, Groueix *et al.*, 2018

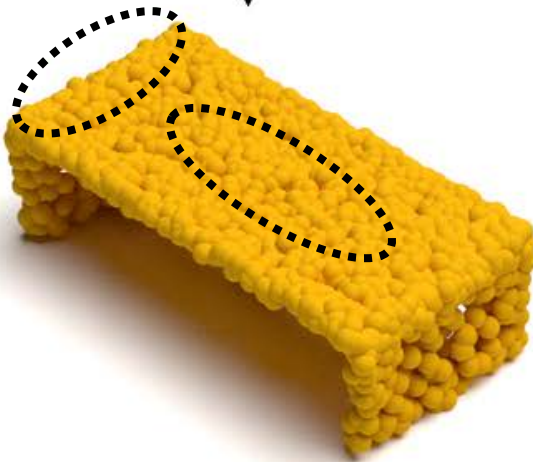
FoldingNet, Yang *et al.*, 2018

Attack Robustness to Defense

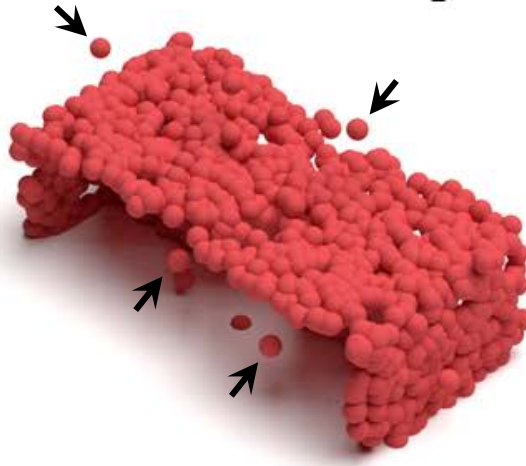
Source



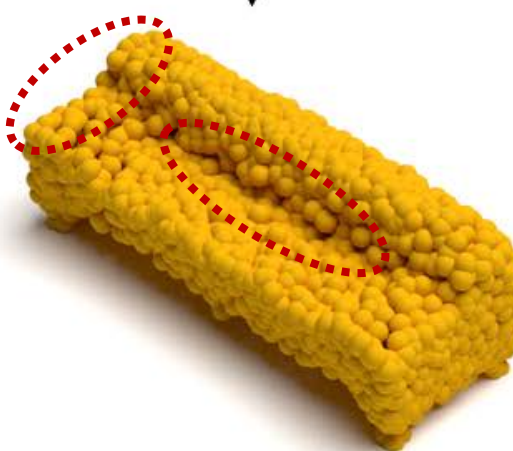
↓ Autoencoder



Adversarial Example



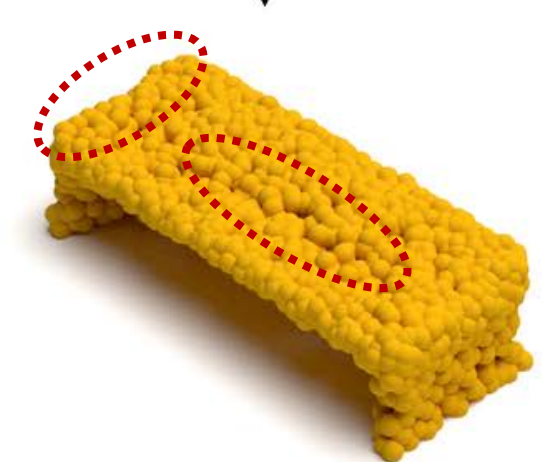
↓ Autoencoder



Defended Input



↓ Autoencoder



Summary



Geometric adversarial attack –
changes the reconstructed shape



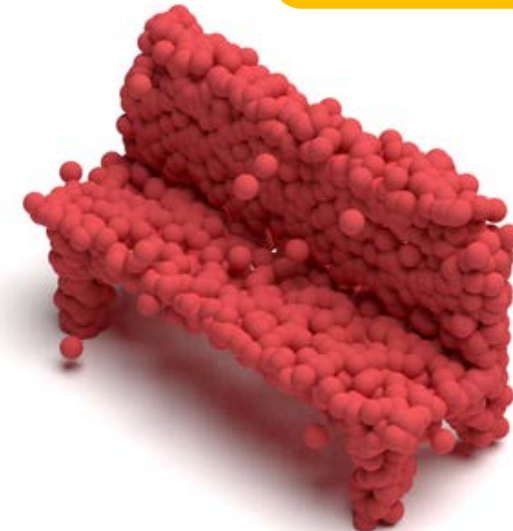
Not entirely defendable –
a residual effect remains



Paper and code are available –
github.com/itailang/geometric_adv



THANK YOU!



Autoencoder ↓

