

JUNE 6 SAT AM - 597



Generative Adversarial Perturbations with Cross-paradigm Transferability on Localized Crowd Counting

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Why Evasion matters for Crowd Models?

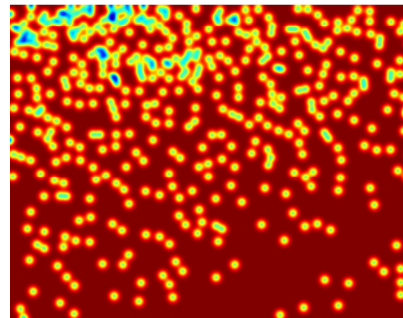
Crowd Models are

- More complex than Classification tasks
- Lacks inherent robustness
- Deployed in real life
 - Public safety
 - Retail analytics
- Attacks can cause:
 - Undercounting = Safety Risk
 - Overcounting = False Alarm

Clean Count:428



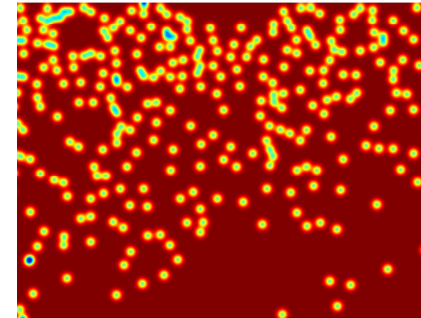
Clean Density Map



Adversarial Count:285



Adversarial Density Map



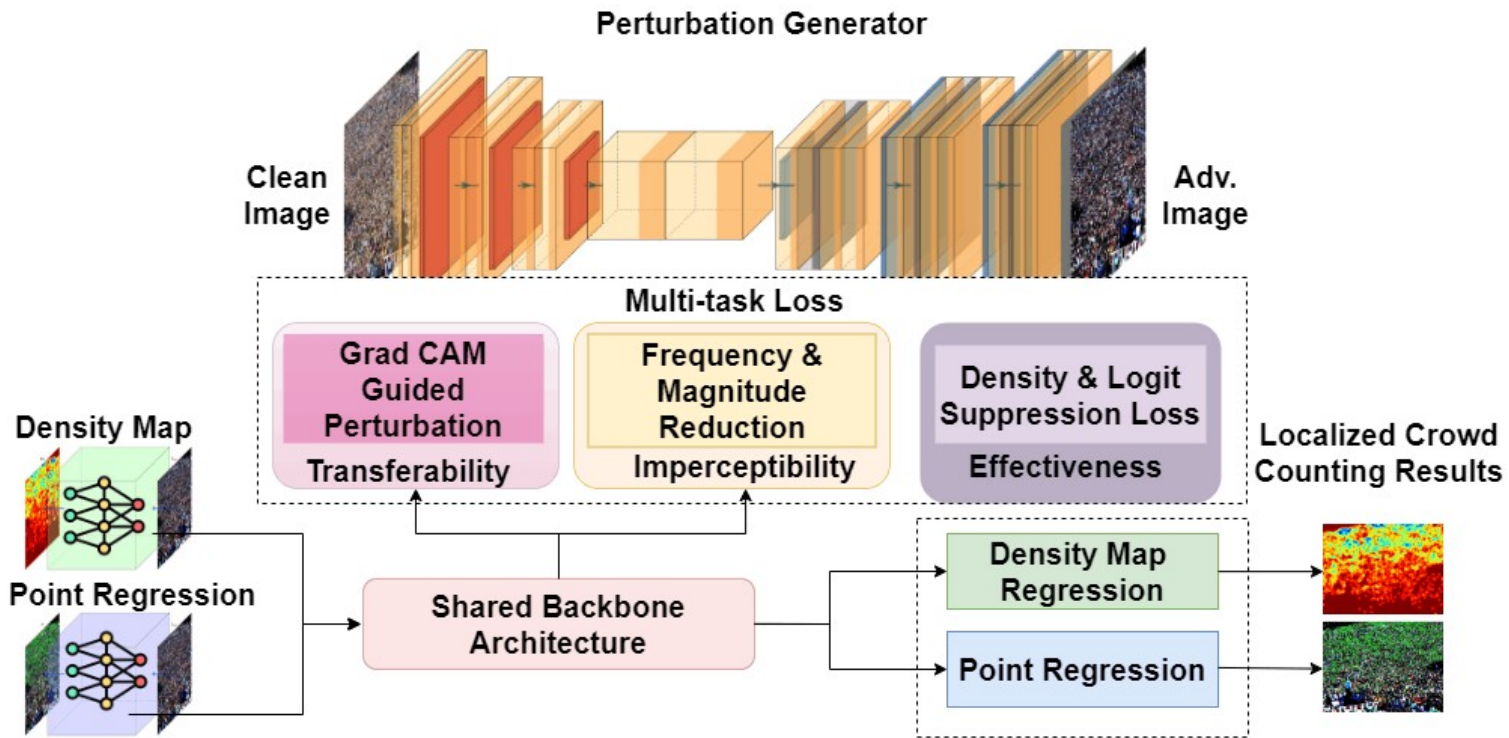
Narrow Scope for Pre-existing Attacks



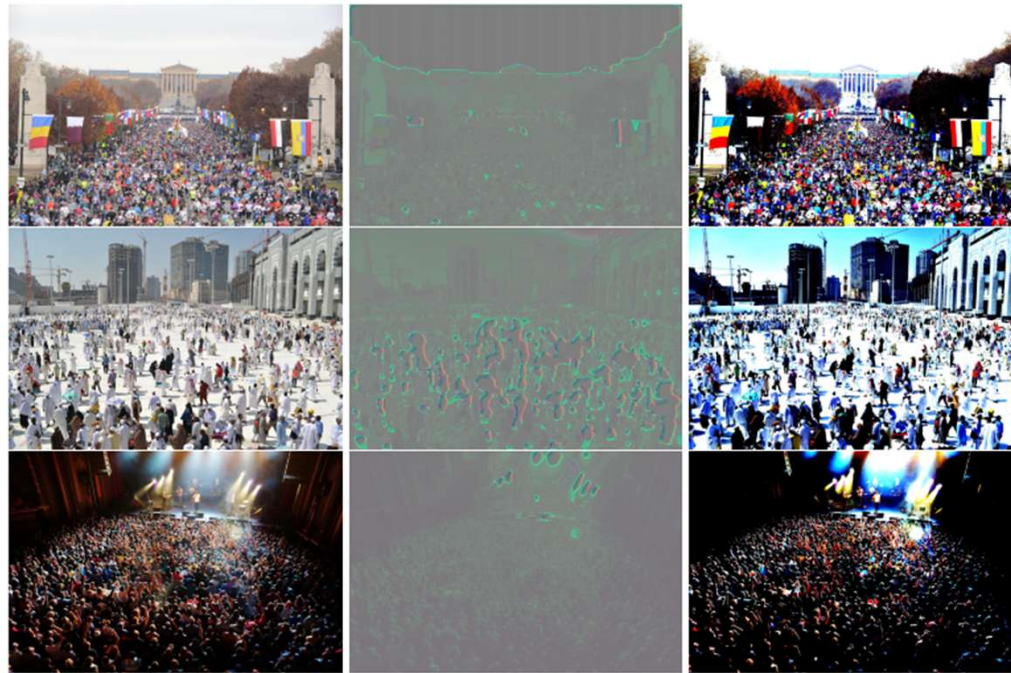
Prior Work	Attack Type	Limitation
PAP (CCS-2022)	Patch-Based	Visible perturbations
APAM (MM- 2021)	Patch-Based	Density-map only
Others	White-box	No transferability study

Q: Can adversarial perturbations transfer across heterogeneous localized crowd counting paradigms – density map and point-regression?

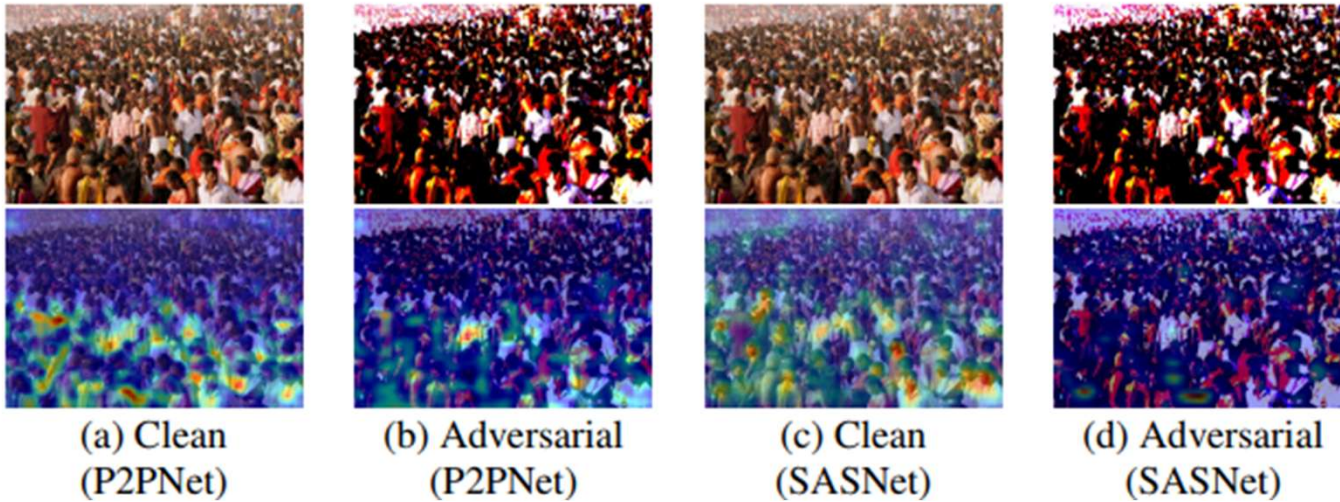
Generator-Based Framework for Cross-Paradigm Transferable Perturbations



Sample Instances of Adversarial Examples

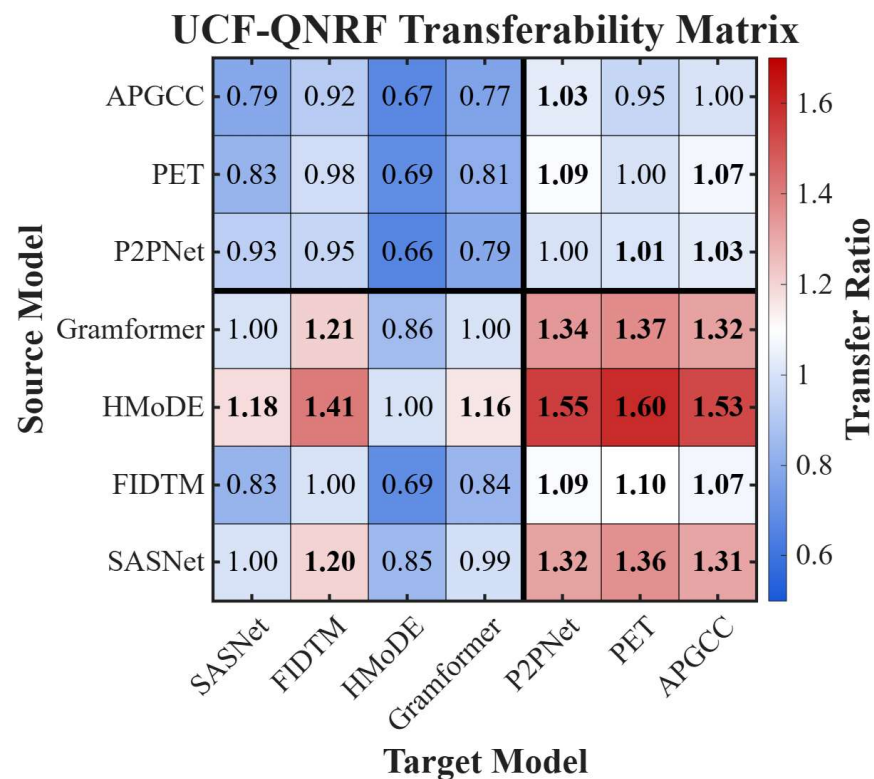
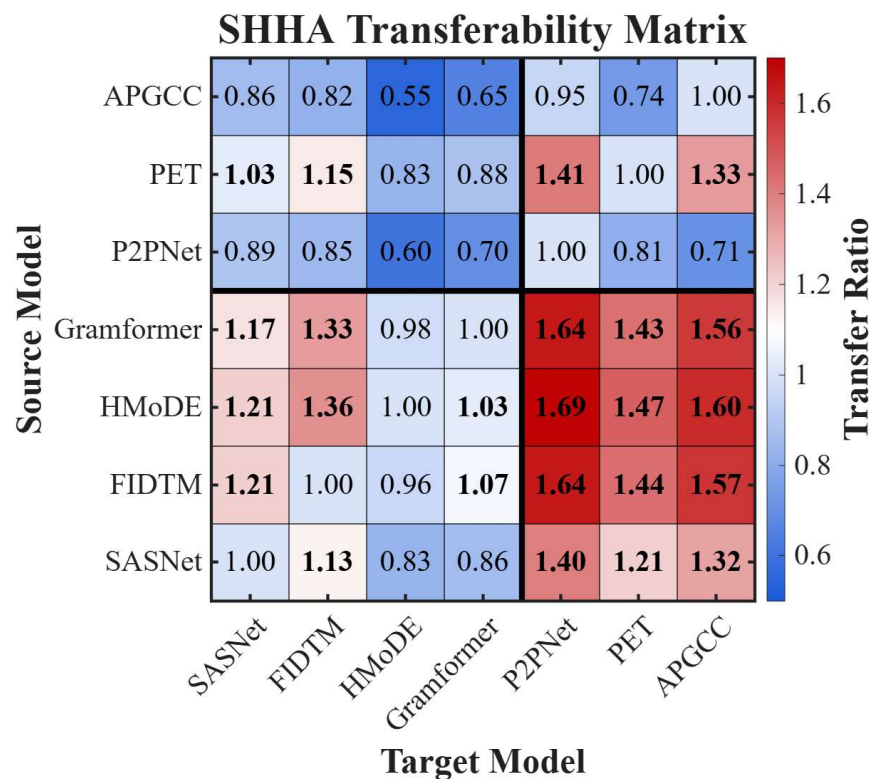


Transferability through Backbone Dependency



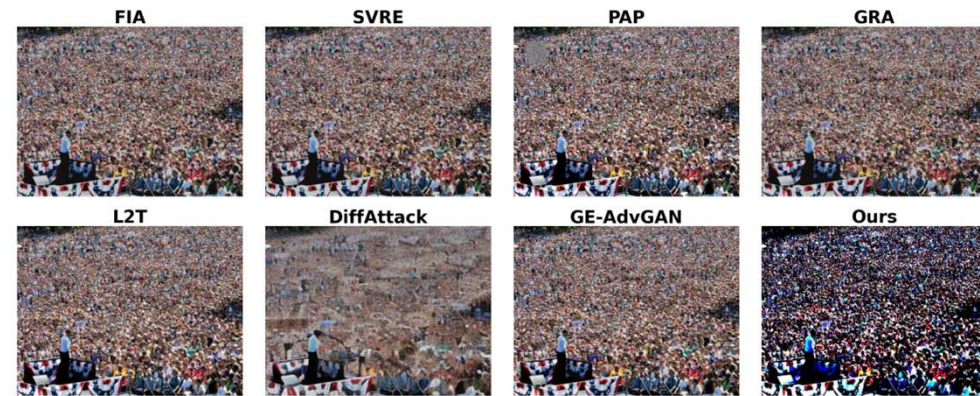
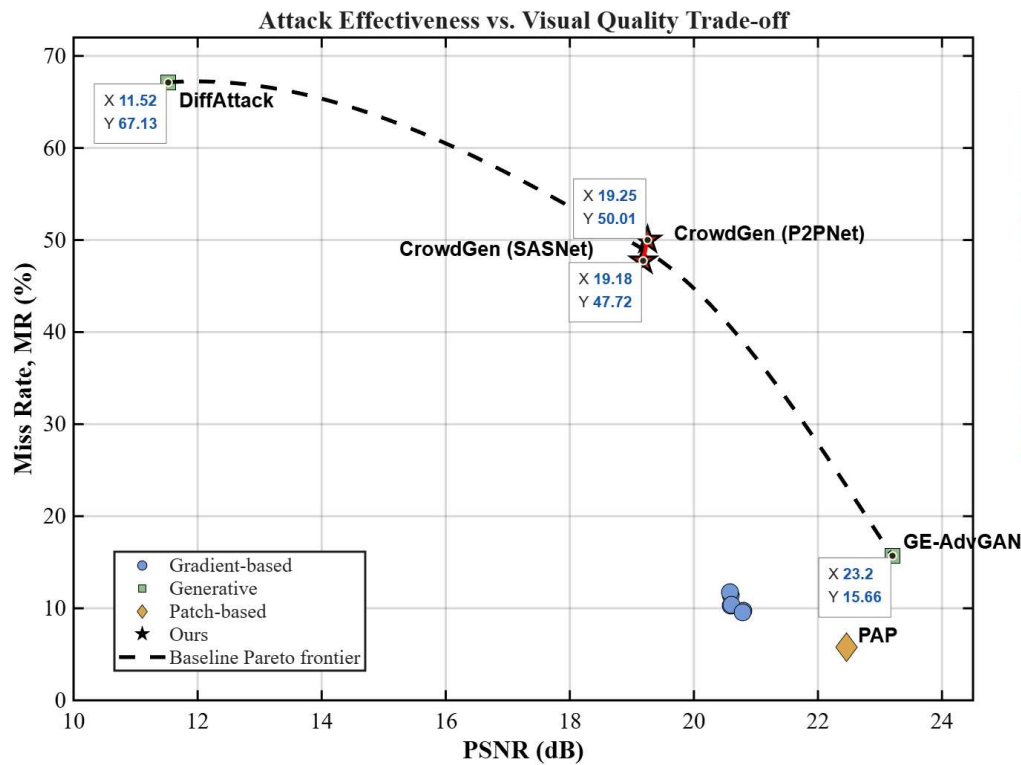
Both P2PNet (Point Regression) and SASNet (Density Map) model has same VGG-16 backbone

Transferability Analysis



Super Transferability!

Baseline Comparison



- DiffAttack: Very Strong but Visible
- Ge-AdvGAN: Very Stealthy but Weak
- Ours (CrowdGen): Strong + Stealthy



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Paper: <https://arxiv.org/abs/2603.24821>

Code: <https://github.com/simurgh7/CrowdGen>