

RODS: Robust Optimization Inspired Diffusion Sampling for Detecting and Reducing Hallucination in Generative Models

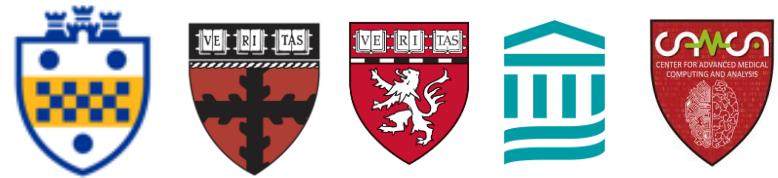
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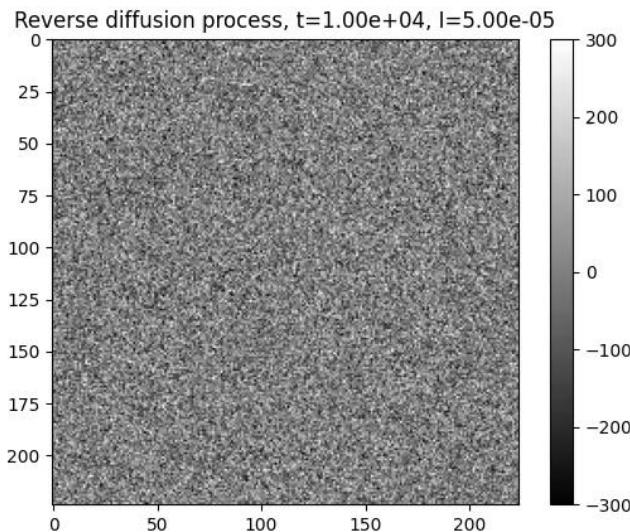
Background and Motivation



- Hallucination in Diffusion Model



Handiffuser,
Narasimhaswamy, S et al.,
CVPR 2024



New AI tool generates realistic satellite images of future flooding

The method could help communities visualize and prepare for approaching storms.

Jennifer Chu | MIT News
November 25, 2024



Related Work

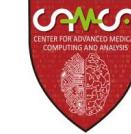
Fine-tuning

Filtering

heuristic modification

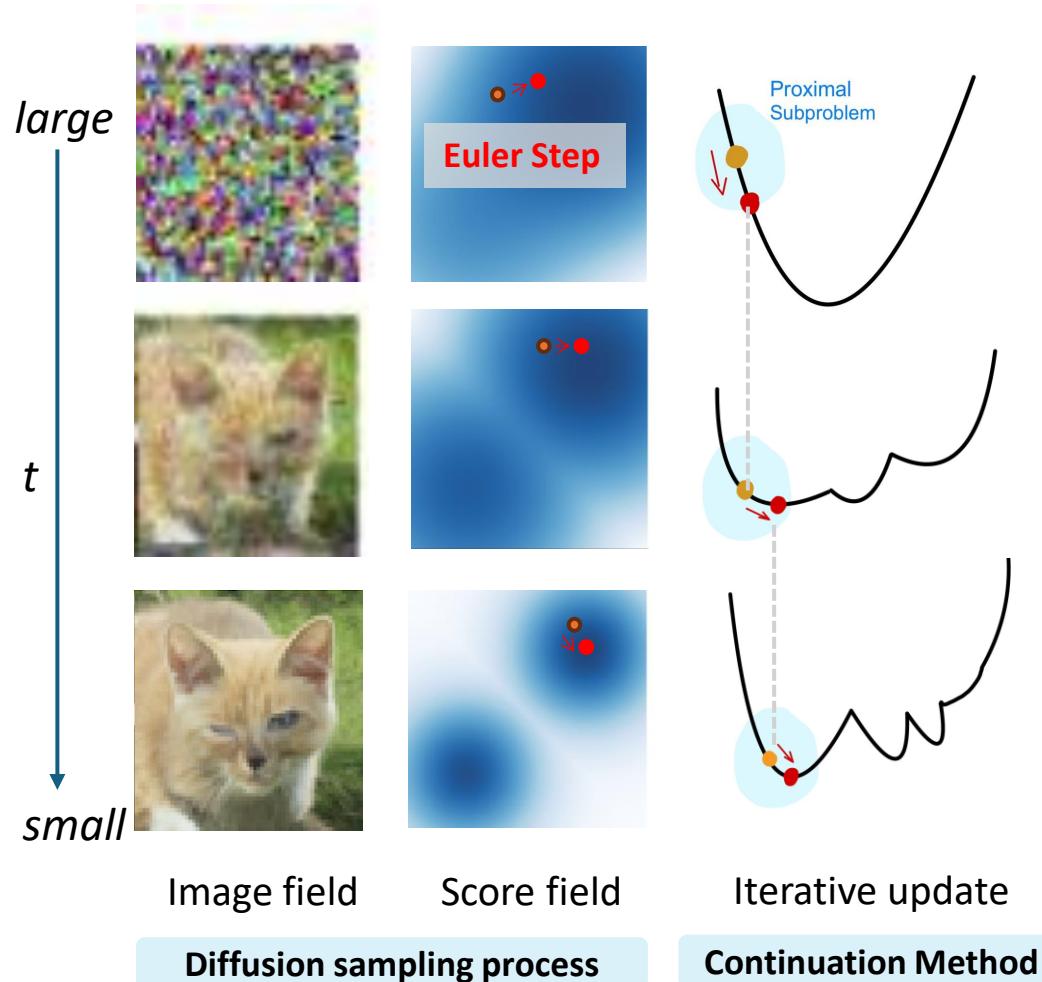
- Limitation

- Lack of theoretical guidance
- Lack of benchmark



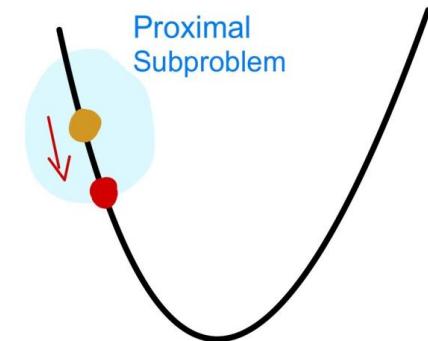
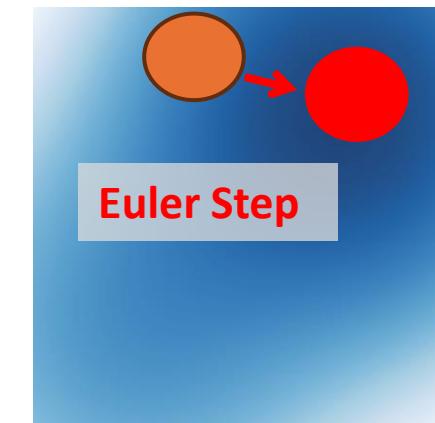
Theory & Methods

An Optimization View of Diffusion Sampling Process



Diffusion sampling via ODE = continuation method

Euler = One-Step Gradient = Proximal Update

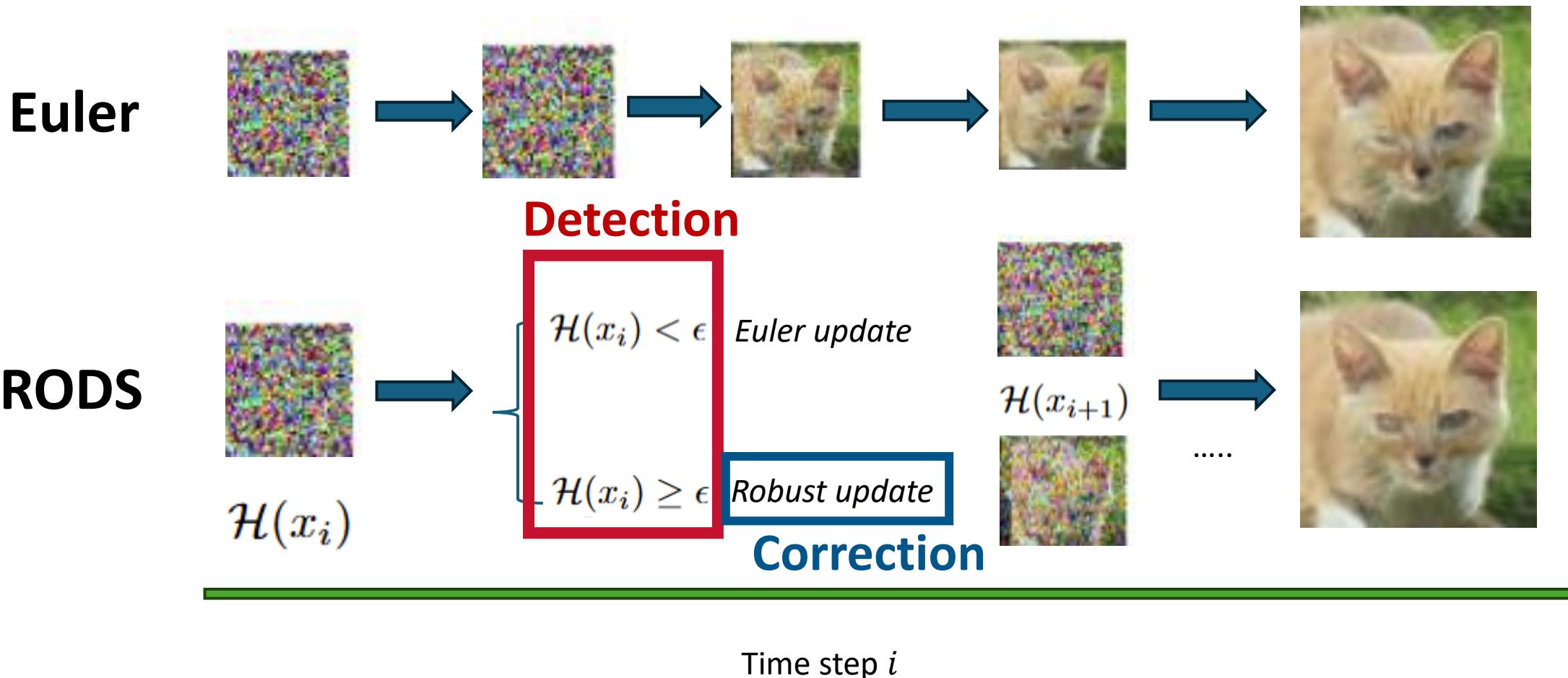


Inspiration: Leveraging various optimization tools

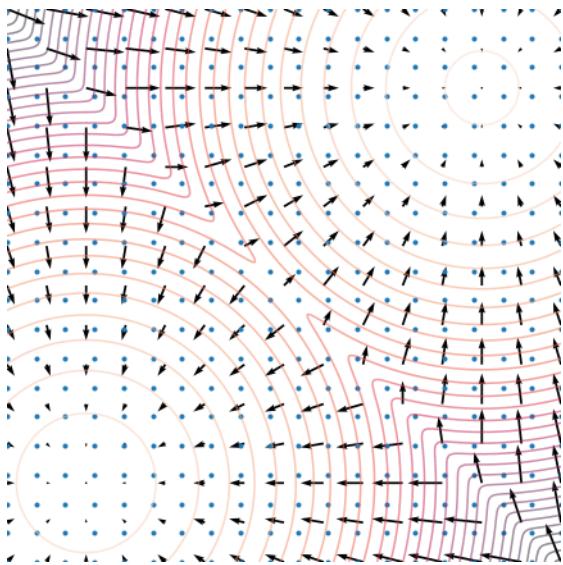
RODS: Robust Optimization inspired Diffusion Sampler

Uncertainty

Hallucination



RODS: Robust Optimization inspired Diffusion Sampler



Low-density region
Unstable score field



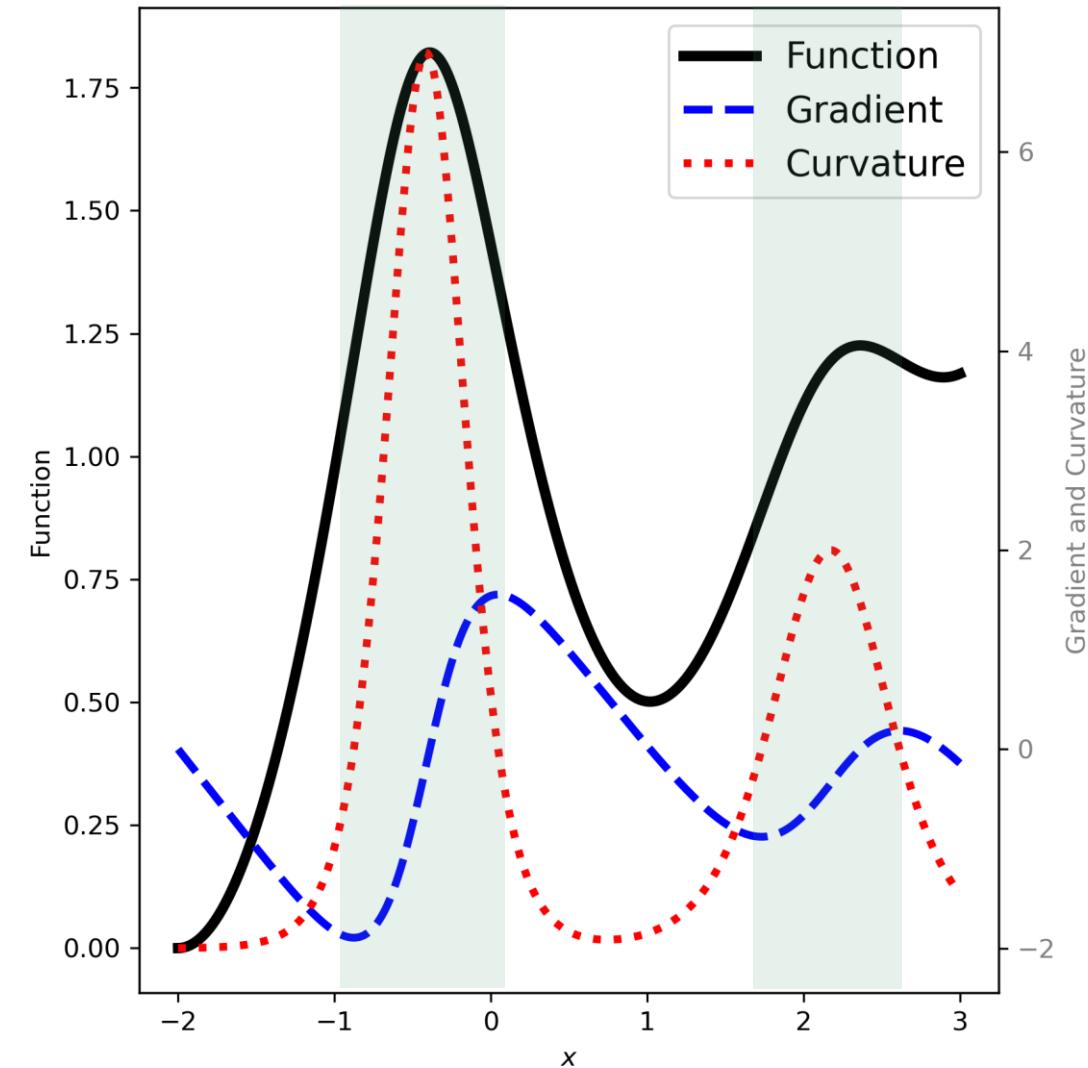
Rapid gradient change
High curvature

- **Curvature Change Detection**

- Hallucination Index:

$$\mathcal{H}(x) = \|\nabla_x \|v(x + \delta)\| - \nabla_x \|v(x)\|\|,$$

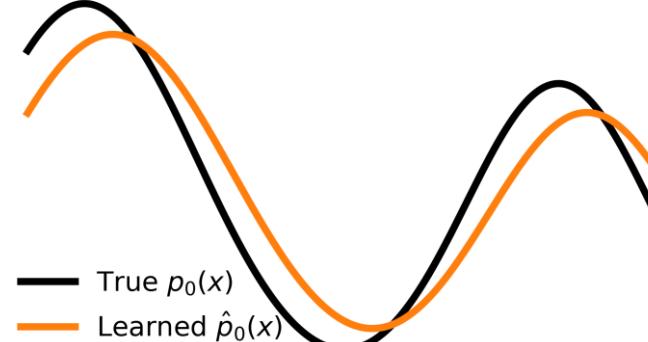
$$\delta = \arg \max_{\|\delta\|=\rho} \|v(x + \delta)\|,$$



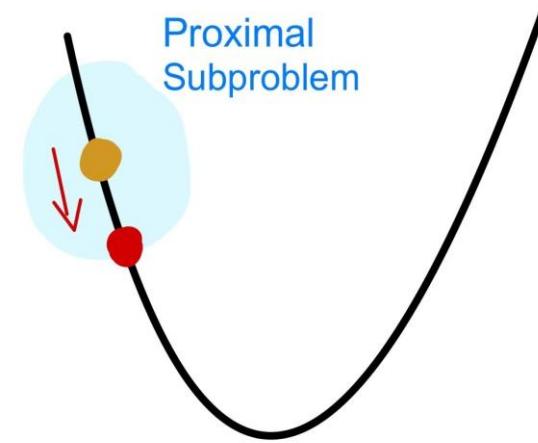
RODS: Robust Optimization inspired Diffusion Sampler

- Robust Sampling Process

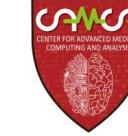
$$\min_x f_t(x) \iff \min_x [-\log p_t(x)].$$



$$\min_x \max_{\hat{f}_t \in \mathcal{F}_t} \hat{f}_t(x) \iff \min_x \max_{\hat{p}_t \in \mathcal{P}_t} [-\log \hat{p}_t(x)]$$



Euler =
One-Step Gradient =
Proximal Update

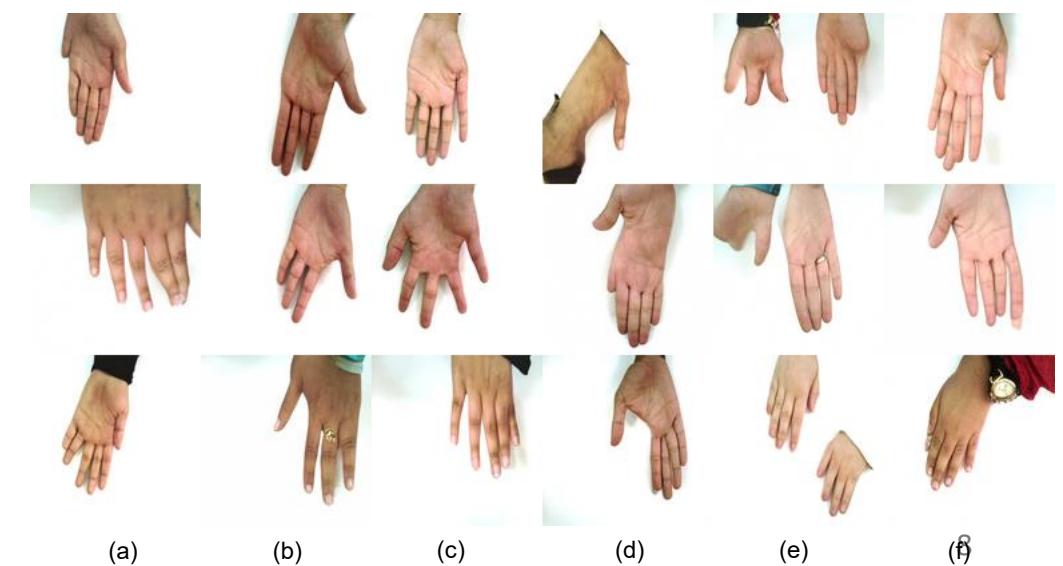
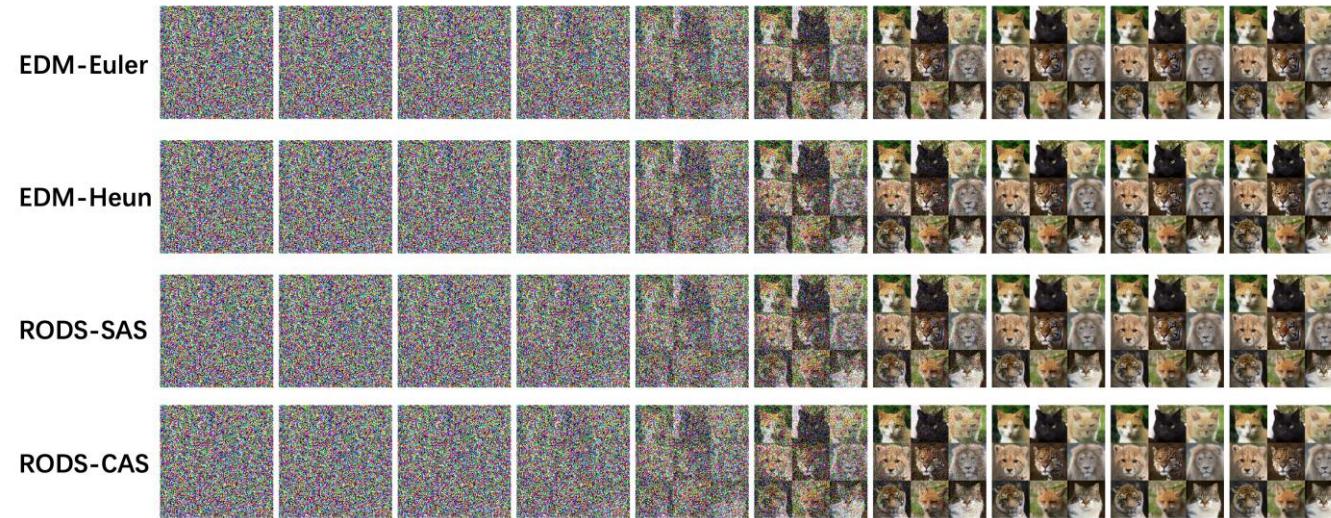
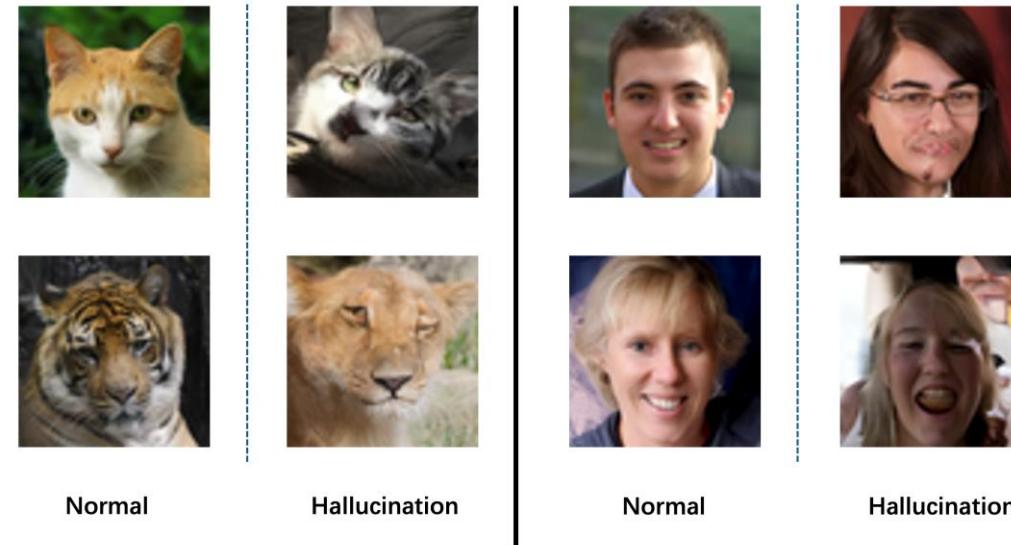


Experiments & Conclusion

Experiment Setting

- Datasets:
 - AFHQ-v2
 - FFHQ
 - 11k-hands

- Labeling by human



Experiment Results

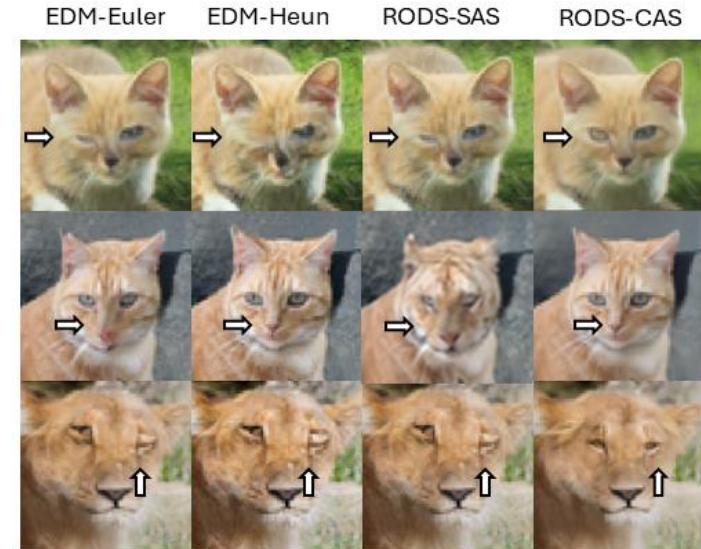
- **Effective Detection**

- Animal Face: 87.5%
- Human Face: 72.5%
- Hands: 96.6%

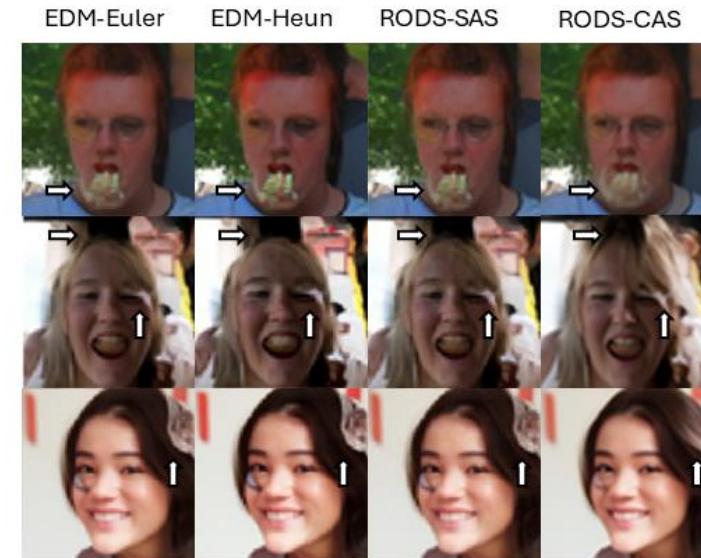
- **Reduces hallucination**

- Animal Face: 43%
- Human Face: 25%
- Hands: 27%

- **No new hallucinations introduced**



Labeled Ground Truth		Prediction	
		Predicted Normal	Predicted Hallucination
Labeled Normal	896	176	better: 7 same: 155 worse: 0 unclear: 14
	1	7	better: 3 same: 4 worse: 0 unclear: 0

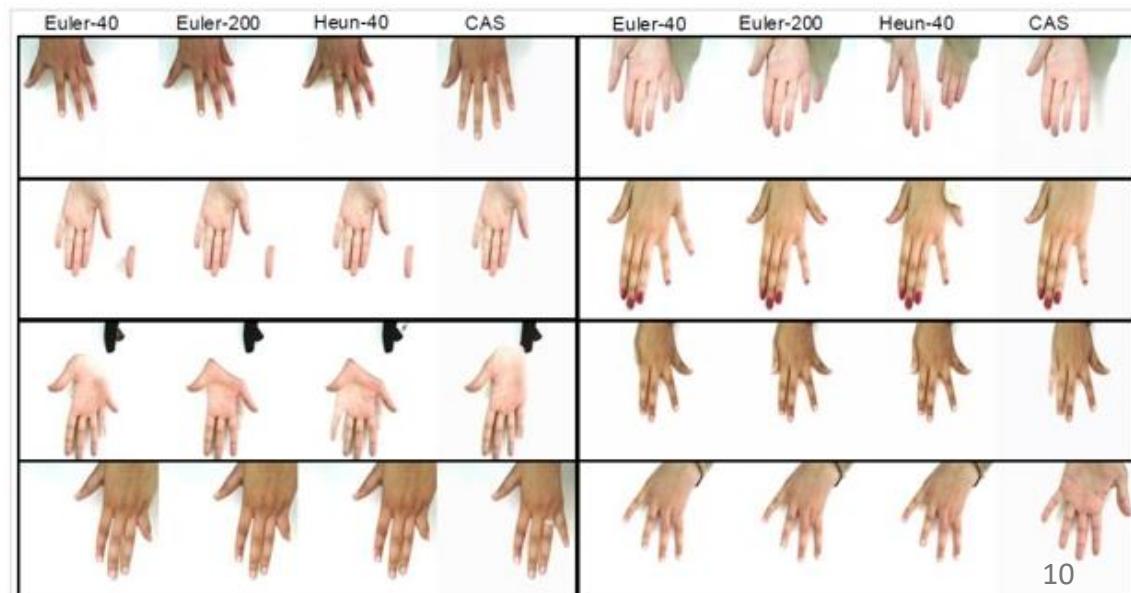
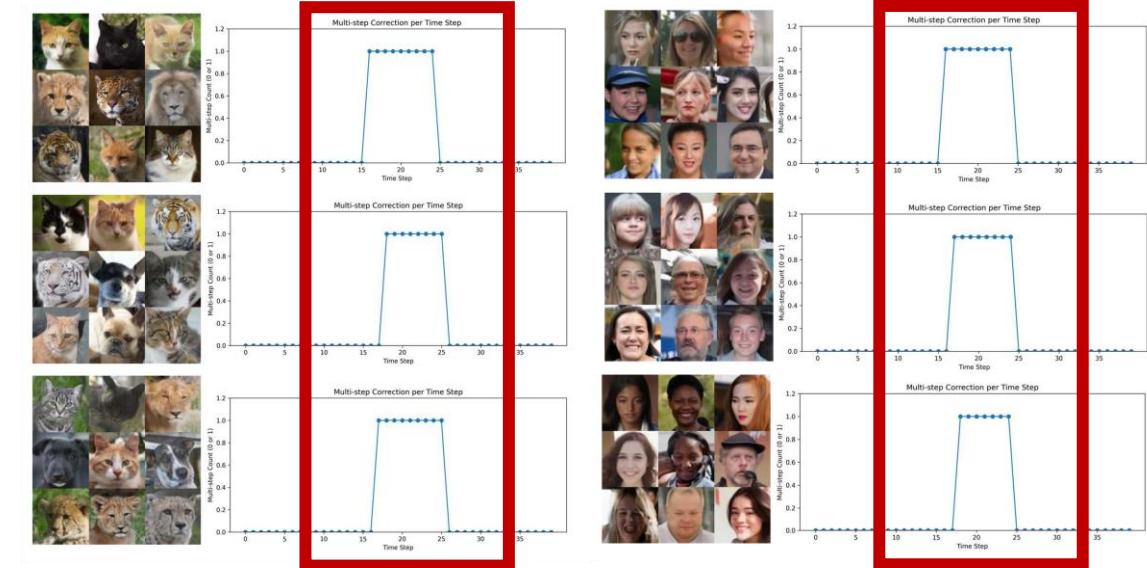


Labeled Ground Truth		Prediction	
		Predicted Normal	Predicted Hallucination
Labeled Normal	859	181	better: 4 same: 170 worse: 0 unclear: 7
	11	29	better: 5 same: 24 worse: 0 unclear: 0

Experiment Results

- Hallucination occur in the middle
- Keep Comparable Time

Metric	EDM-Euler	EDM-Heun	RODS-CAS
Hall.% ↓	19.4%	18.0%	14.3%
Correction % ↑	N/A	17.7%	26.3%
New Hall. % ↓	N/A	2.5%	0.0%
Time (s)	2.71	5.38	4.82



Conclusion

