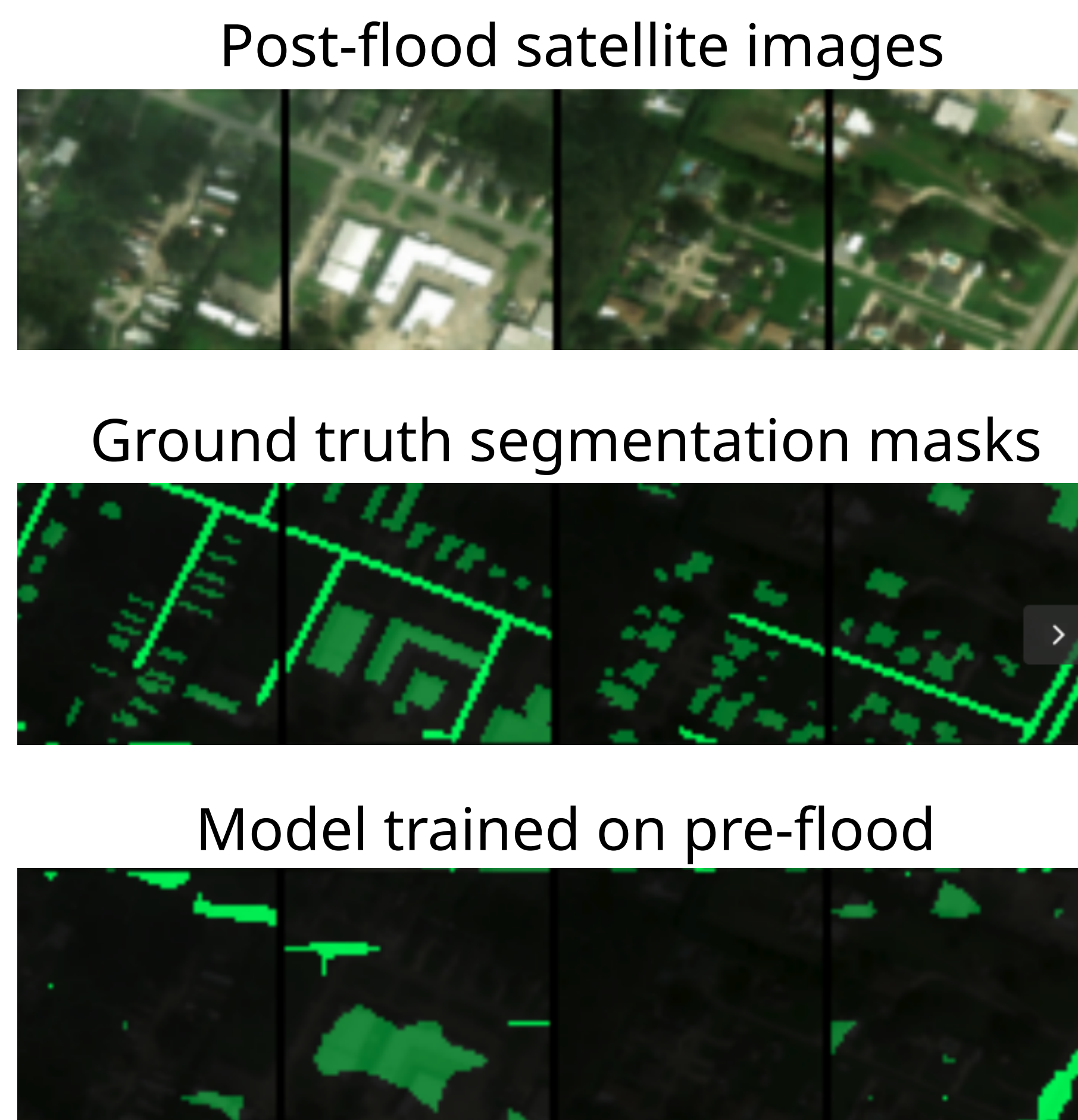


Earth Observation Context

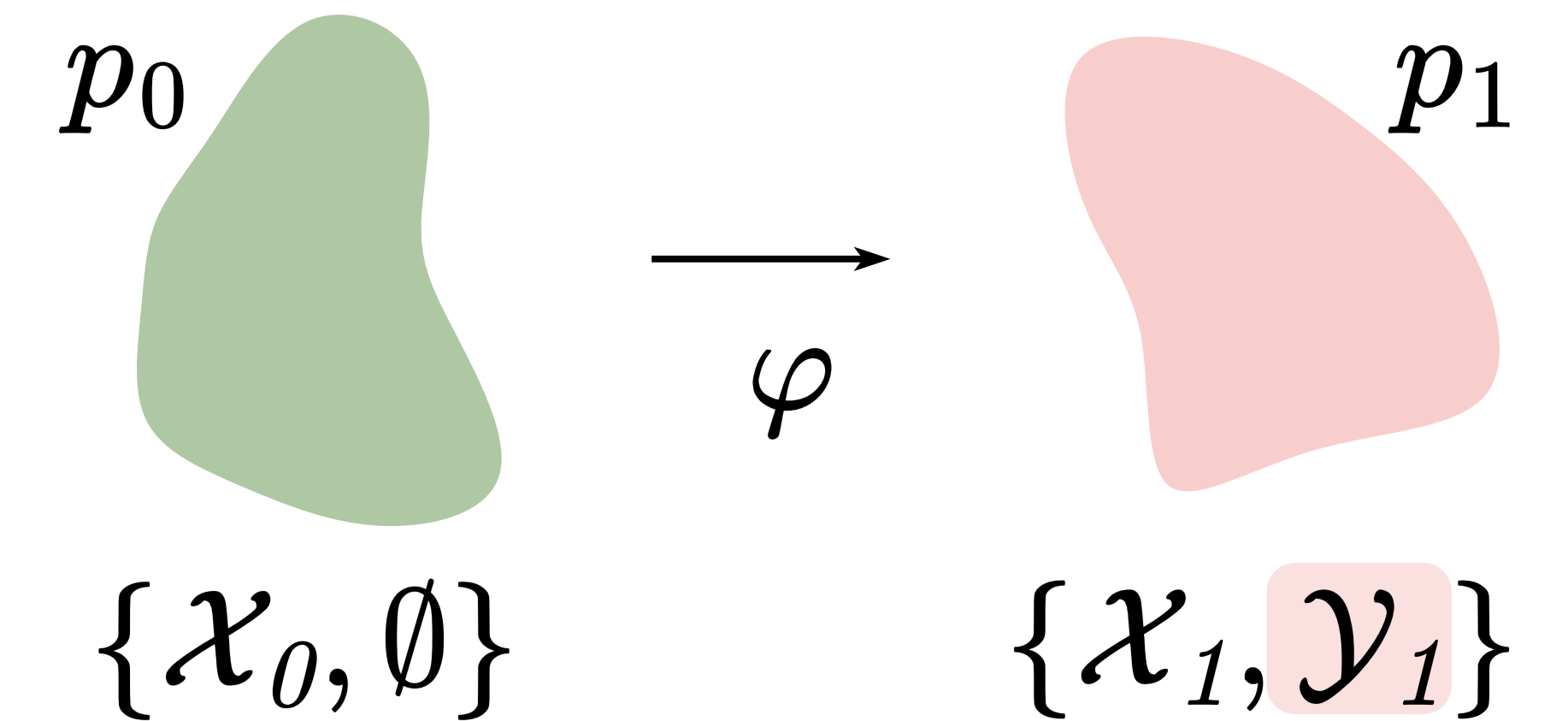
EO models often face distribution shift at inference time:

- different geographical areas
- seasonal changes
- different sensors
- extreme events (flooding, fire)
- ...

→ Their predictive capabilities are severely reduced.



Domain Adaptation

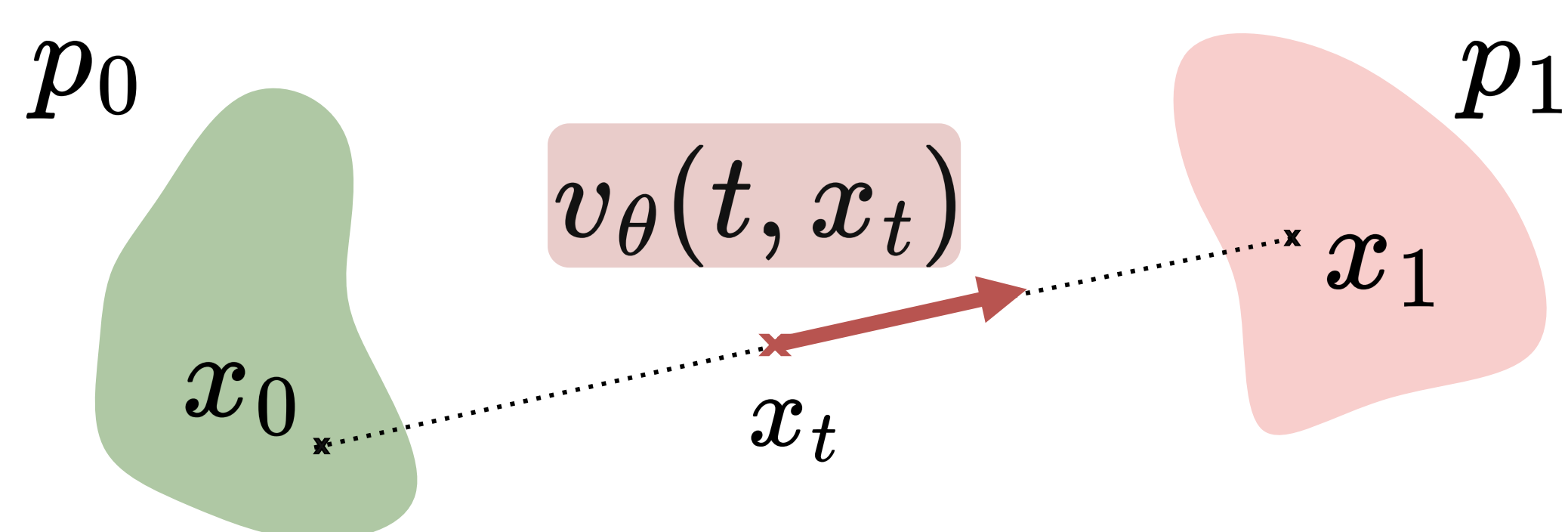


1. Train a segmentation model f_ϕ on labelled dataset $\{X_1, Y_1\}$
2. Learn a mapping φ between p_0 and p_1
3. Inference time: apply segmentation model to the transferred data

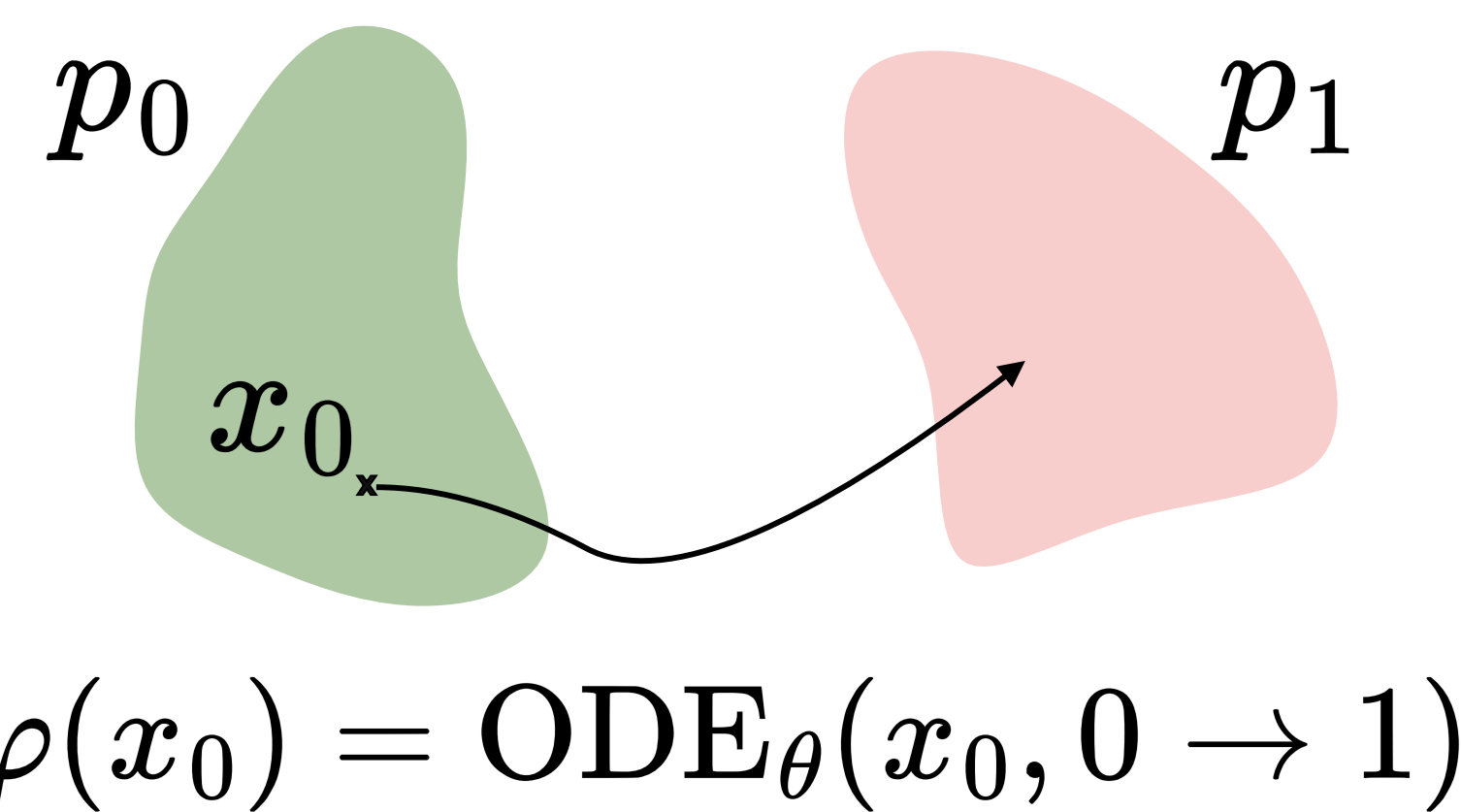
$$\hat{y} = f_\phi(\varphi(x_0))$$

Flow Matching

Training



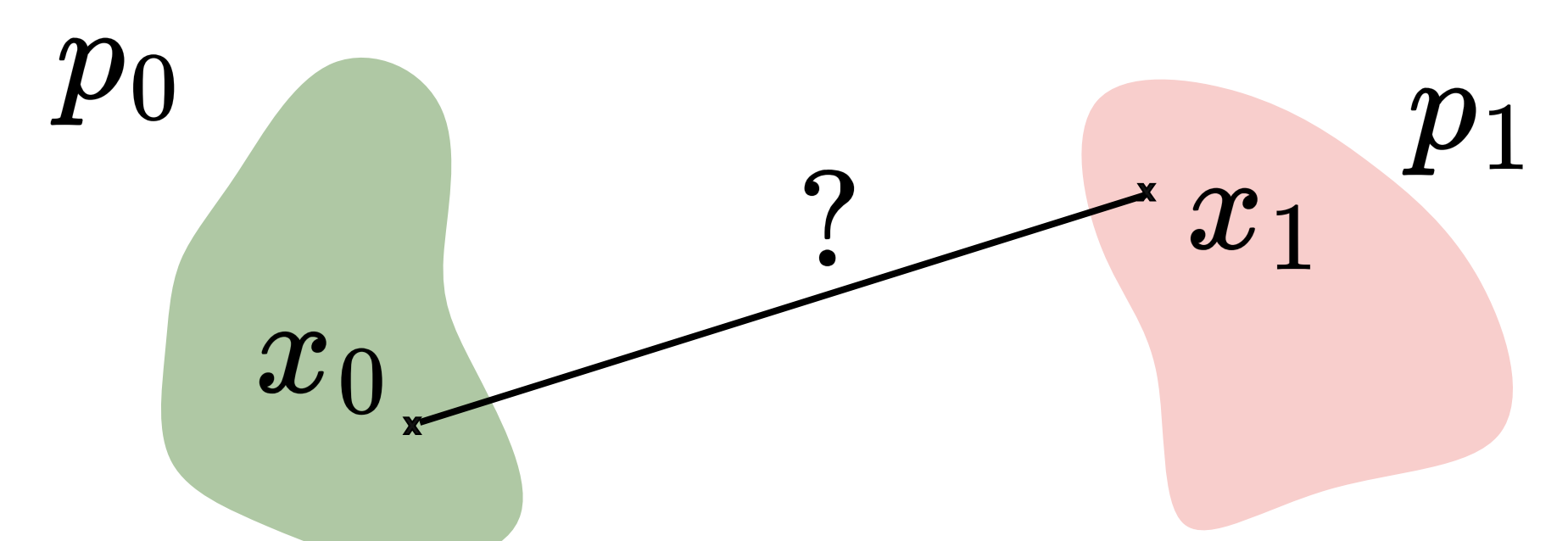
Inference



- Flow Matching generalizes diffusion models
- Simple regression loss

$$\mathcal{L}_{CFM} = \mathbb{E}_{x_0, x_1 \sim p(x_0, x_1)} \left[\left\| v_\theta(t, x_t) - \underbrace{(x_1 - x_0)}_{\text{True conditional velocity}} \right\|^2 \right]$$

Coupling



How to select (x_0, x_1) pairs during training?

Available pairs → pair as a natural coupling
 $x_1 = T(x_0)$

No prior information → independent coupling
 $(x_0, x_1) \sim p(x_0)p(x_1)$

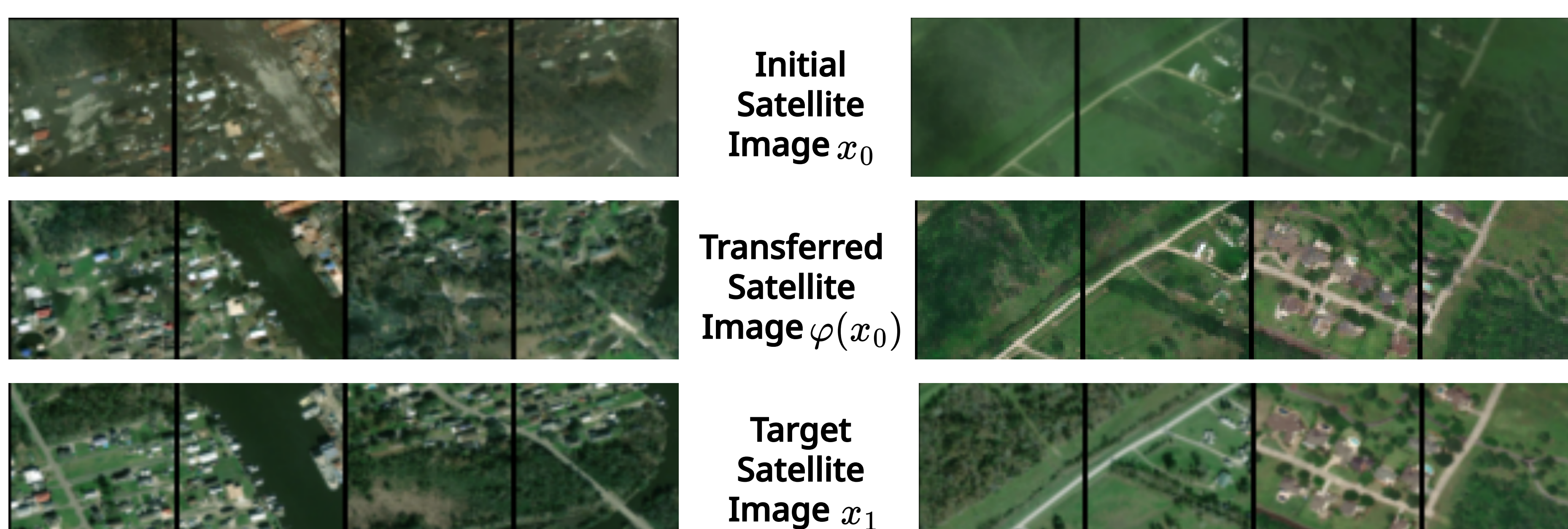
Efficient sampling objective → OT coupling
 $(x_0, x_1) \sim \pi(x_0, x_1)$

Preliminary Results

Setup:

- SpaceNet 8 dataset
- Two locations: Germany and Louisiana
- 256x256 RGB images
- Segmentation model: DeepLabV3+

Initial distribution p_0 → Pre-Flood
Target distribution p_1 → Post-Flood



Quantitative results

	Pre/Post-Flood	
	F1 ↑	mIoU ↑
No transfer	04.63	05.21
On target (upper bound)	35.67	35.94
Independent coupling	01.89	02.12
OT coupling	05.63	06.15
Aligned coupling	25.45	26.56

Takeaways:

1. Flow Matching training is similar to diffusion models
2. Flow Matching provides a gain with respect to the segmentation metrics.
3. Independent and OT couplings fail to preserve semantic information during transfer