

Fast Inference and Update of Probabilistic Density Estimation on Trajectory Prediction

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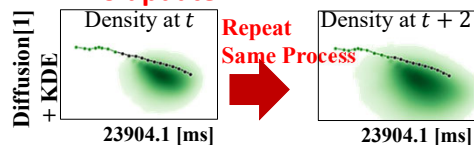
Overview

Trajectory Prediction needs

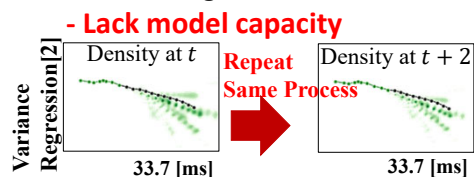
- Probability Density Estimation
- Fast Inference

Previous Methods

- Generative models (e.g. diffusions)
 - **Slow** due to KDE and sampling
 - **No update**

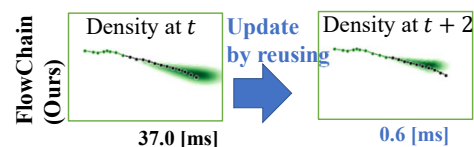


- Variance Regression



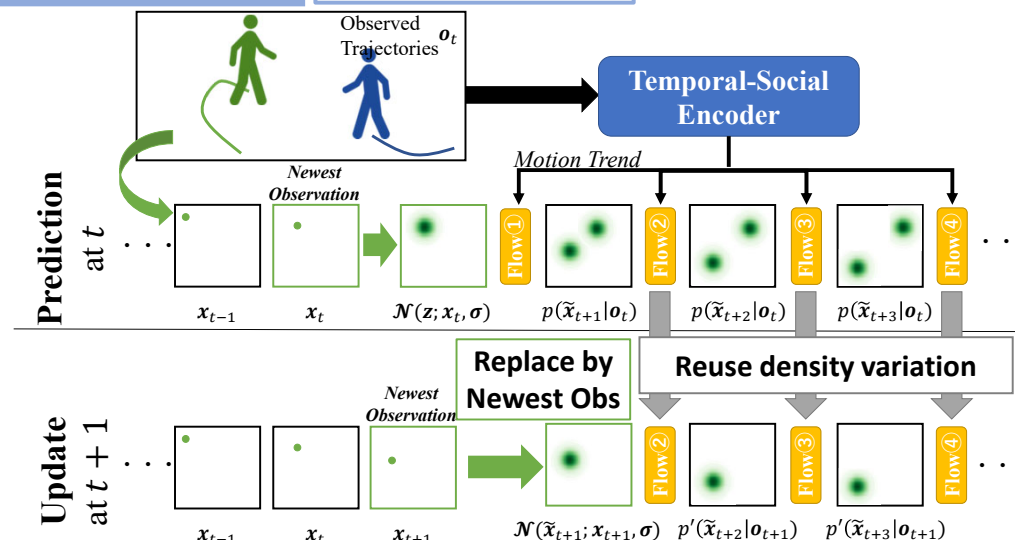
Our FlowChain

- **Fast** thanks to analytical comp.
- **Reusing and Ultra-fast update**

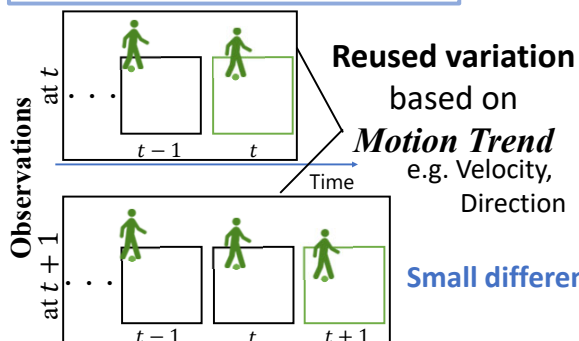


Methodology

Model Overview



Why reusing is possible?



Why reusing matters?

99% of comp. cost is due to density variation

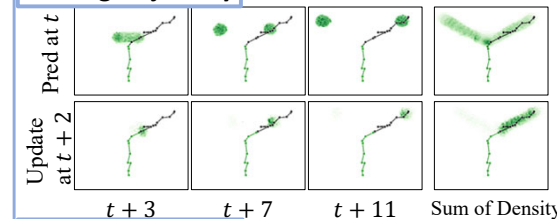
- Trajectory Encoder
- Log-det-jacobian of flow

By reusing density variation, we achieve **Ultra-fast update**

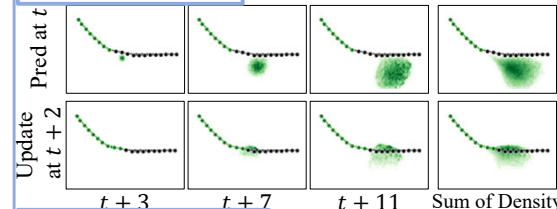
Results

Observation: Estimated dist.: GT future:

Forking trajectory



Turning trajectory



Quantitative Results

Method	Accumulated Comp Time[s]			Log-probability \uparrow		
	Initial	Update@1	Update@5	Initial	Update@1	Update@5
MID[1]	23.9	47.8	119.5	-0.90	-0.90	-0.90
Trajectron[2]	0.47	0.94	2.35	-281.86	-281.86	-281.86
Ours	0.63	0.63	0.63	-0.26	-0.45	-0.53

Limitation

- Performance degrades via our ultra-fast update.
- FlowChain only accepts traj. with fixed interval

Reference & Link

- [1] Motion Indeterminacy diffusion. CVPR2022
- [2] Trajectron++. ECCV2020

Code is available on Github!!!

