

Data-Driven Tracking with Event Cameras

Code Progress

- Downloaded dataset
- Created tracks from MultiFlow dataset using generate_tracks.py and generate_representations.py
- Finished training a model on synthetic data

```
GPU available: True (cuda), used: True
TPU available: False, using: 0 TPU cores
IPU available: False, using: 0 IPUs
HPU available: False, using: 0 HPUs
Initialized recurrent dataset with 256 tracks.
Initialized recurrent dataset with 64 tracks.
LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES: [0]
```

	Name	Type	Params
0	l2	L2Distance	0
1	reference_encoder	FPMEncoder	14.9 M
2	target_encoder	FPMEncoder	14.9 M
3	reference_redir	Conv2d	442 K
4	target_redir	Conv2d	442 K
5	softmax	Softmax	0
6	joint_encoder	JointEncoder	4.1 M
7	predictor	Linear	1.0 K
8	flatten	Flatten	0
9	loss	LITruncated	0

```
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34.9 M Trainable params
0 Non-trainable params
34.9 M Total params
139.574 Total estimated model params size (MB)
/home/aircraft-lab/Documents/Deep_Learning_Project/dlenv/lib/python3.9/site-packages/pytorch_lightning/loggers/tensorboard.py:250: UserWarning: Could not log computational graph since the `model.example_input_array` attribute is not set or `input_array` was not given
rank zero warn(
/home/aircraft-lab/Documents/Deep_Learning_Project/dlenv/lib/python3.9/site-packages/pytorch_lightning/trainer/trainer.py:1892: PossibleUserWarning: The number of training batches (8) is smaller than the logging interval Trainer(log_every_n_steps=10). Set a lower value for log_every_n_steps if you want to see logs for the training epoch.
rank zero warn(
Epoch 2: 80%|███████████| | 8/10 [00:10<00:02, 1.30s/it, loss=7.4, v_num=, loss/train_step=7.080, loss/val=5.570, loss/train_epoch=7.510]
Validation: 0it [00:00, ?it/s]
```

Research

Paper	Key Findings	Limitations	Feature Extraction	Results
Low-Latency Automotive Vision with Event Cameras	Reduces latency in automotive vision; enhances detection	Data fusion challenges with existing systems	Asynchronous data tracking	Faster reaction times, improved object detection
Temporal Feature Markers for Event Cameras	Uses strobe LEDs for accurate marker recognition	Issues with varying lighting conditions	Detects periodic LED events	High speed and accuracy in marker tracking
BlinkTrack: Feature Tracking over 100 FPS via Events and Images	Combines event and RGB data for high-frequency tracking	Relies on RGB quality and sync	Differentiable Kalman filter	Over 100 FPS tracking with events and images
Enhancing Robustness in Asynchronous Feature Tracking	Fuses event and frame data for robust tracking	Increased computational complexity	Patch-based fusion	Improved accuracy in dynamic environments
Data-driven Feature Tracking for Event Cameras	First data-driven event tracker; frame attention module	Performance in highly dynamic scenes uncertain	Template matching with correlation volume	130% improvement in feature age; faster inference