



Derivative Delay Embedding: Online Modeling of Streaming Time Series

Zhifei Zhang (PhD student),
Yang Song, Wei Wang, and Hairong Qi

Department of Electrical Engineering & Computer Science

Outline

- 1. Challenges of Online Modeling**
- 2. Derivative Delay Embedding (DDE)**
- 3. Markov Geographic Model (MGM)**
- 4. Experimental Results**

Challenges of Online Modeling

Most modeling methods require pre-processing or assumptions:

- Segmentation
- Alignment
- Normalization

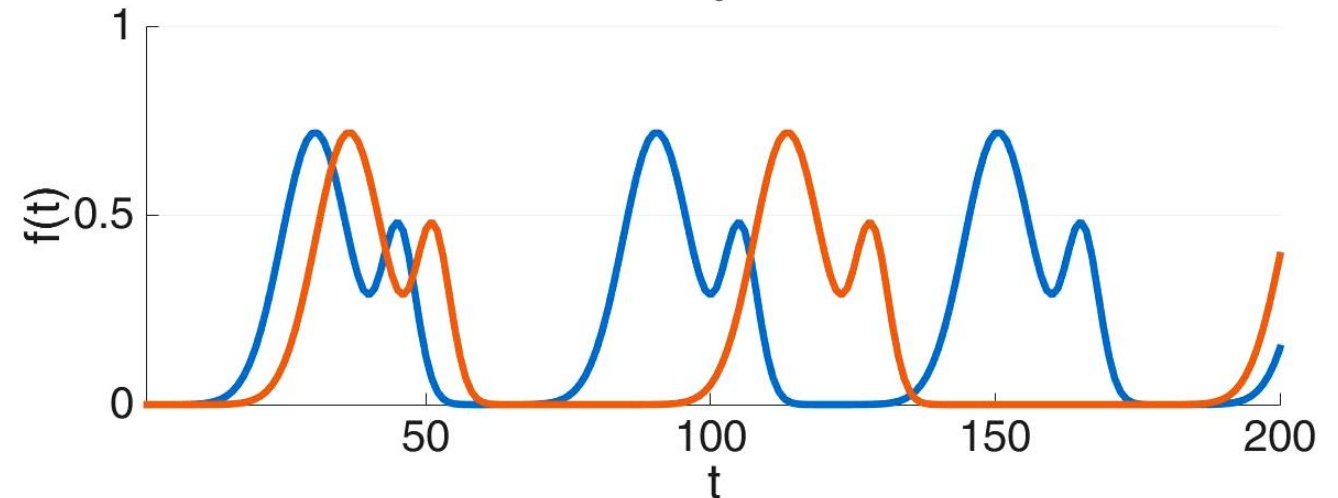
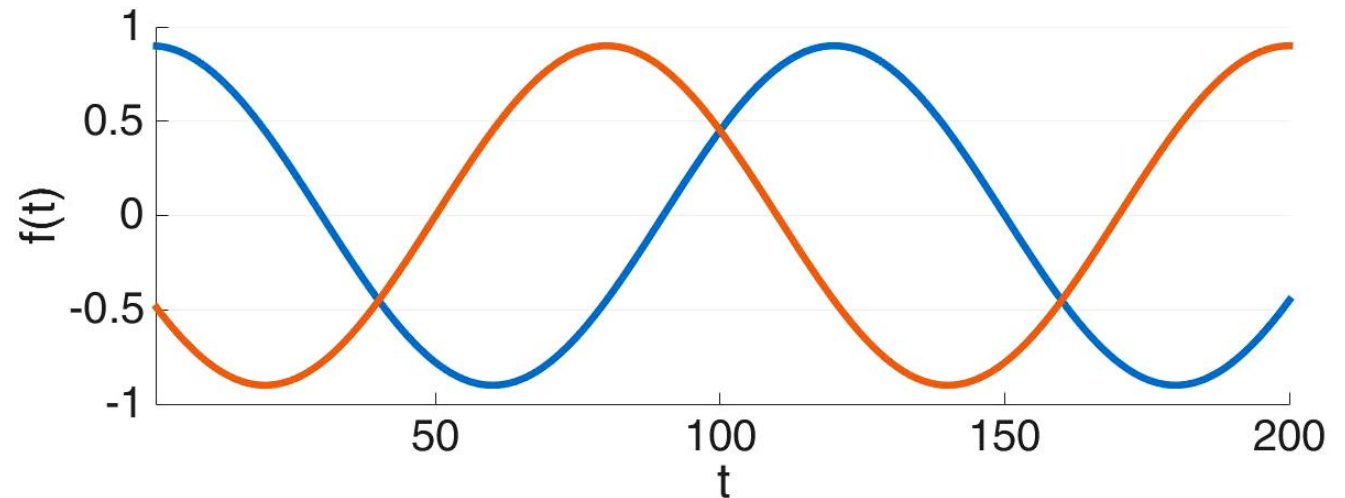
However, for the online scenario:

- Infinite time series
- Real-time
- Misalignment

Pre-processing and unrealistic assumptions are not allowed, thus misalignment challenges the online modeling

Misalignment in Online Modeling

Misalignment mainly refers to the variation in **phase** and **repeat rate** of streaming time series.



The Proposed Approach



- **Misaligned**
- **Non-periodic**

- **Invariant to misalignment**
- **Real-time**
- **Incremental manner**

- **Online modeling**
- **Online testing**

Delay Embedding (DE)

Reconstruct a latent dynamical system which generates the time series regardless of misalignment.

$$\Phi(x_t; s, d) = (y_t, y_{t+s}, \dots, y_{t+(d-1)s})$$

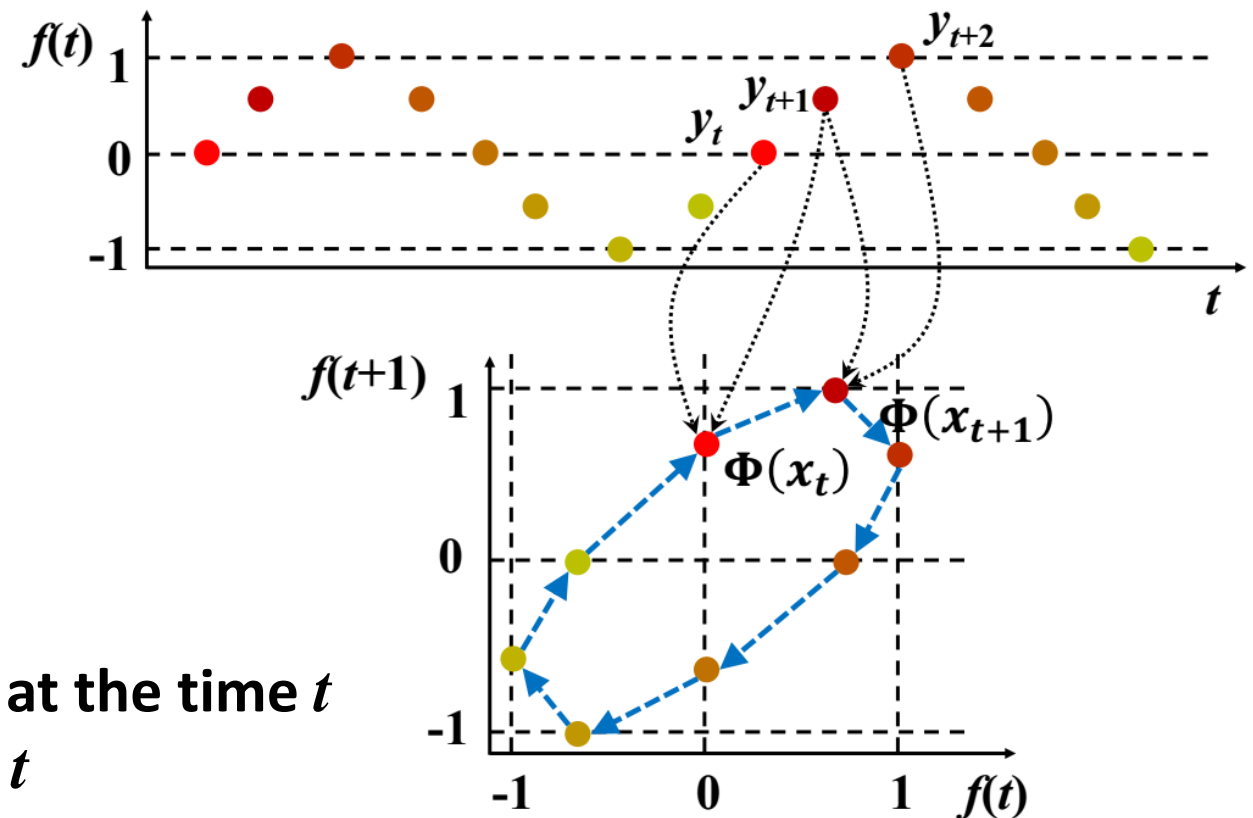
Φ --- estimate of x

s --- delay step

d --- embedding dimension

x_t --- state of the latent dynamical system at the time t

y_t --- observation (time series) at the time t



A Toy Examples of Delay Embedding

$$\Phi(x_t; s, d = 2) = (y_t, y_{t+s}) = (f(t), f(t+s))$$

Φ --- estimate of x

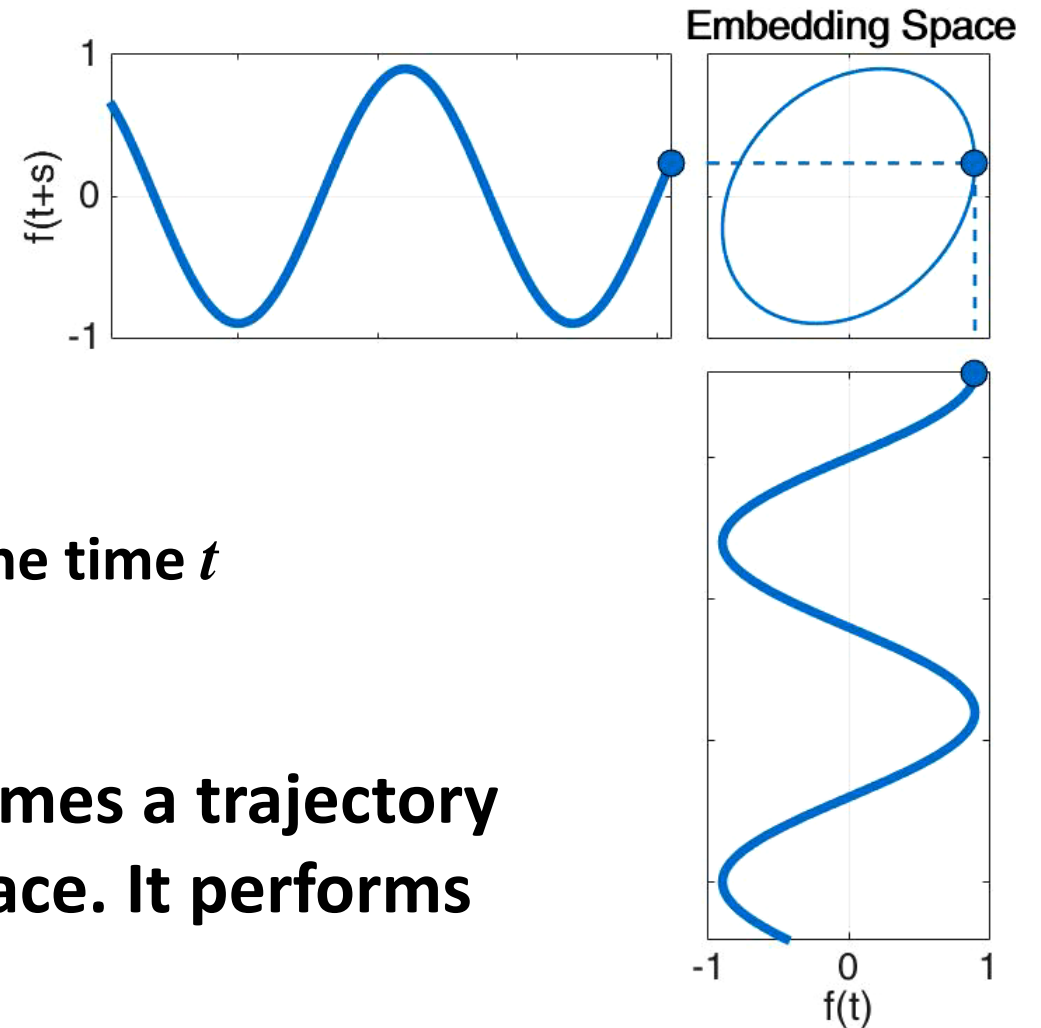
s --- delay step

d --- embedding dimension

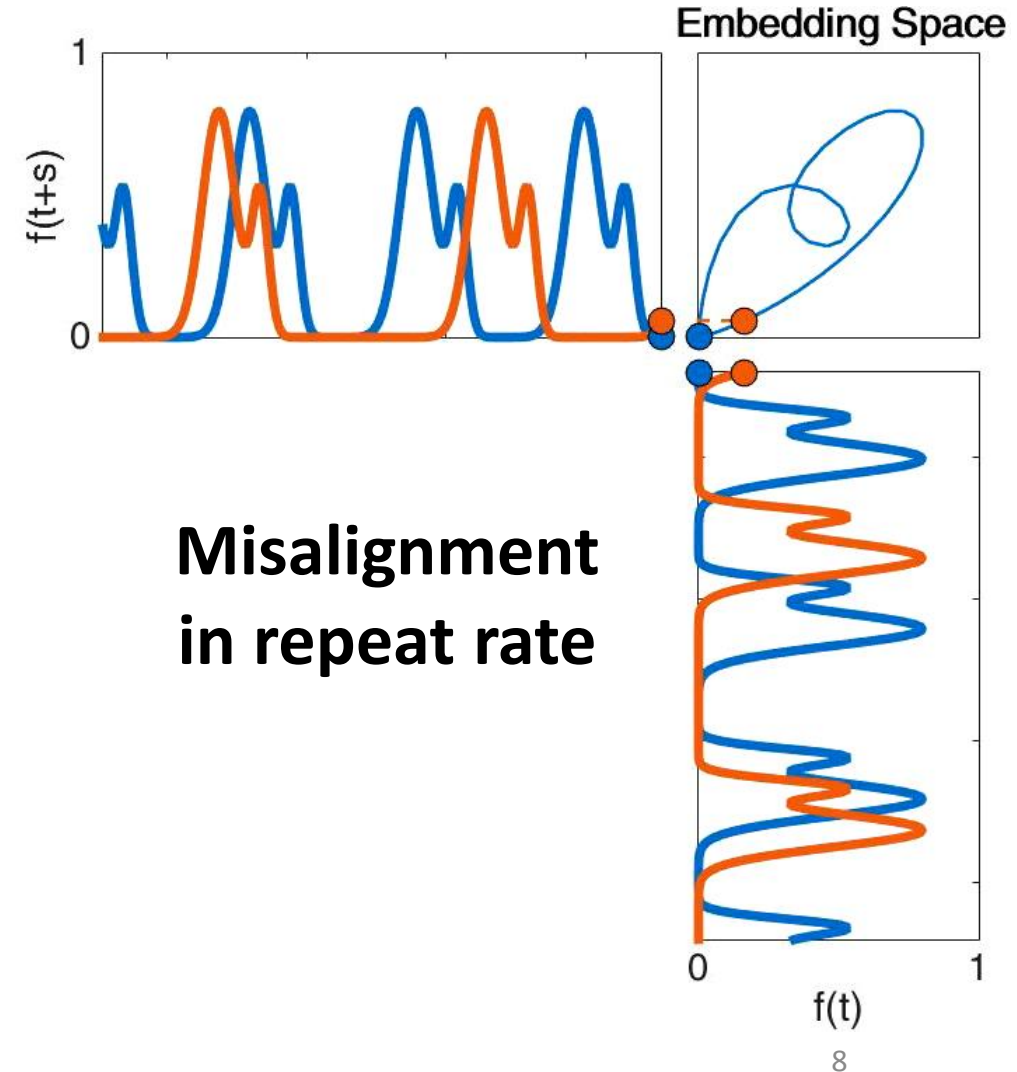
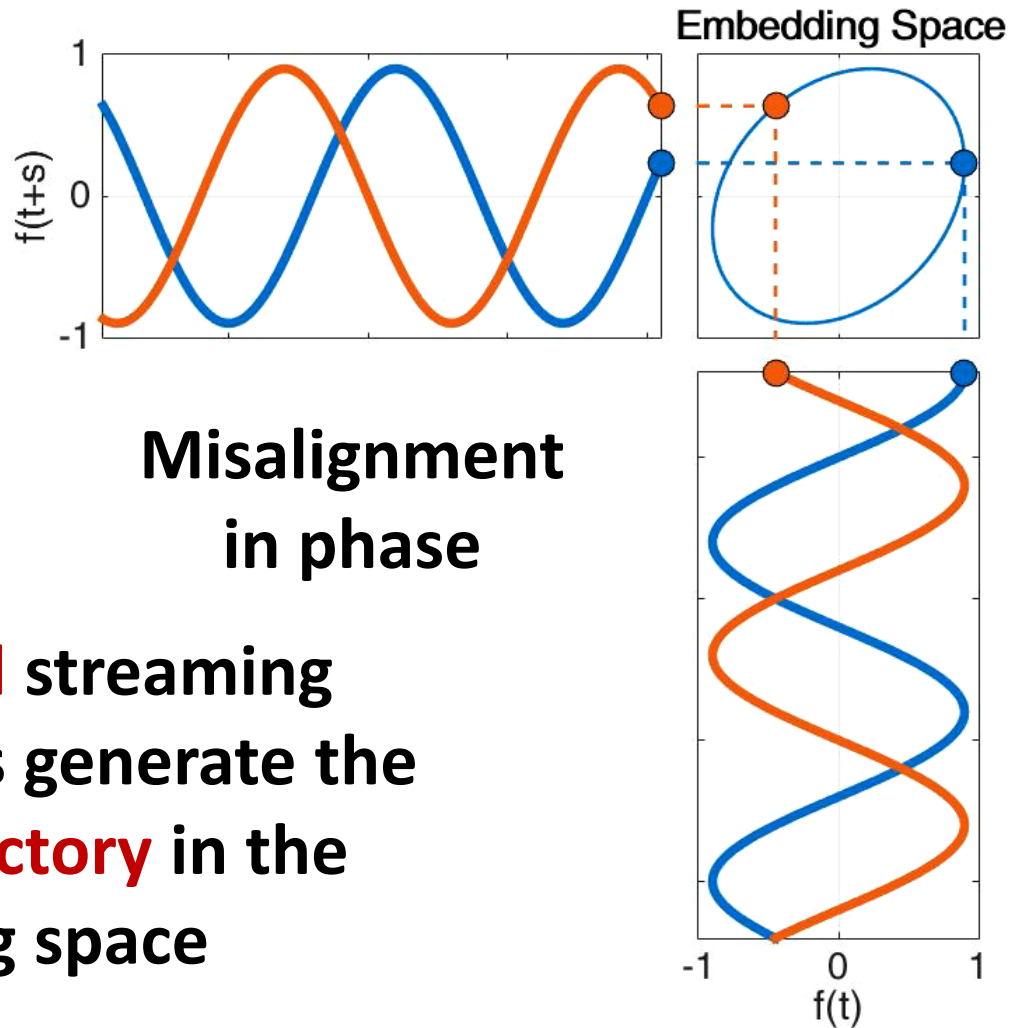
x_t --- state of the latent dynamical system at the time t

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The **infinite** time series becomes a trajectory in a **bounded** embedding space. It performs in **real time**.



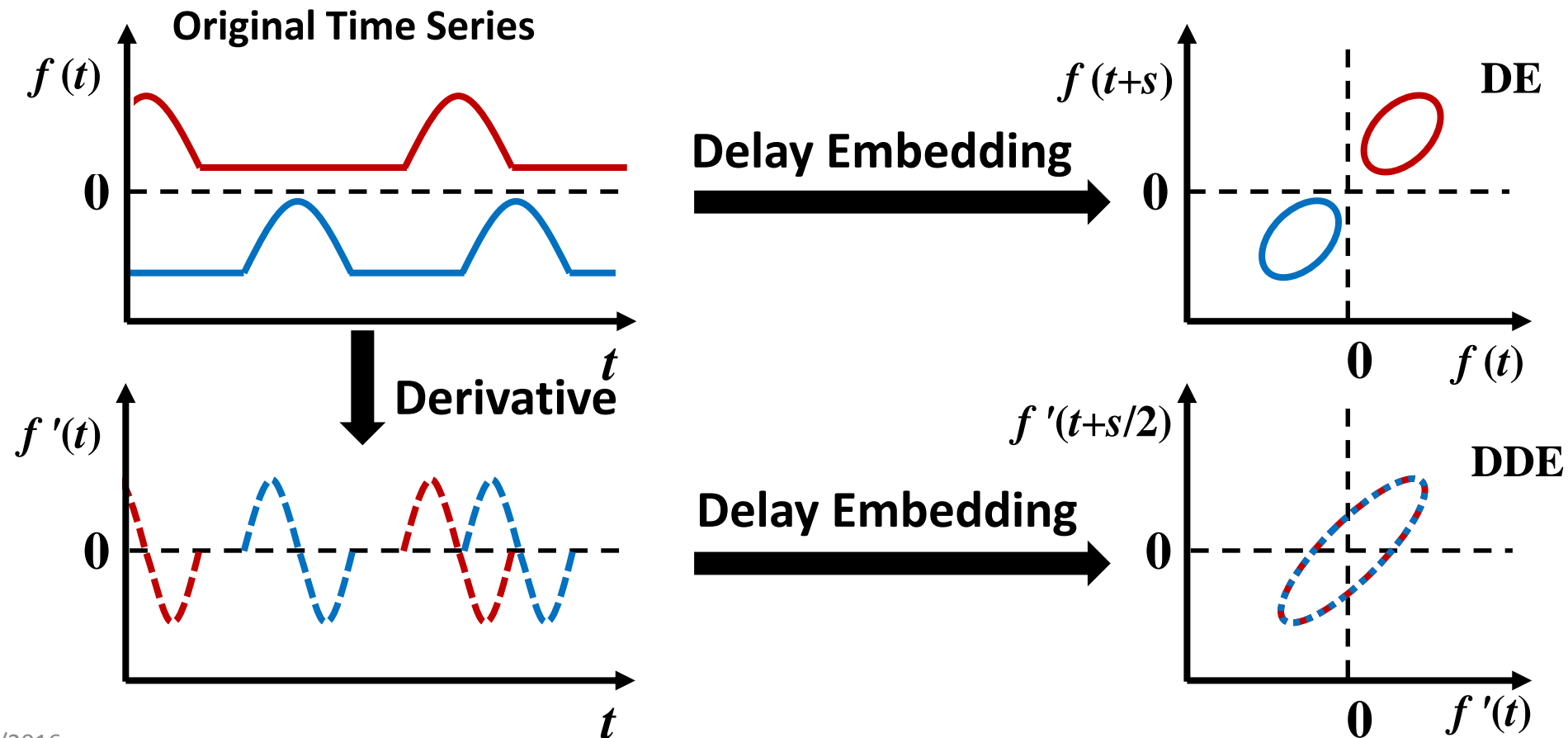
Invariance to Misalignment



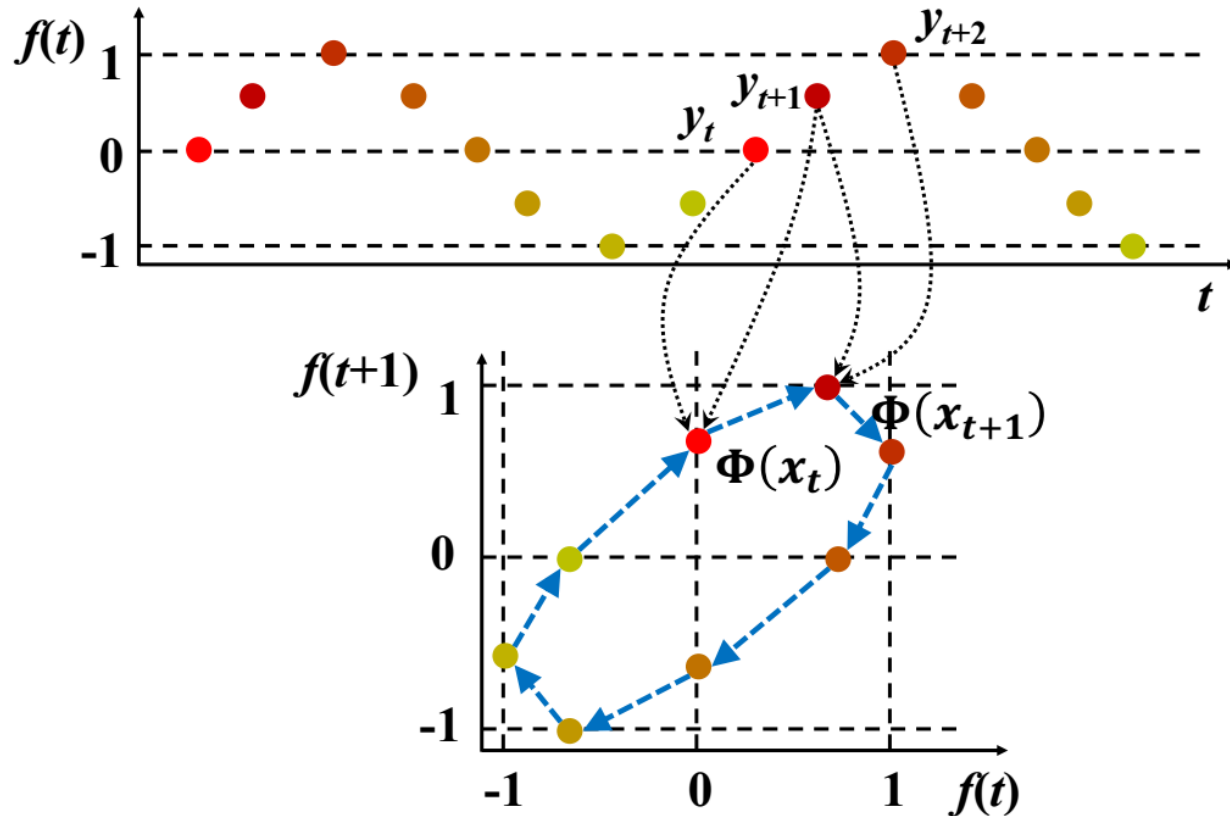
Misaligned streaming
time series generate the
same trajectory in the
embedding space

DE \rightarrow Derivative Delay Embedding (DDE)

Invariant to misalignment of baseline



Trajectory Modeling



In the embedding space, **location** of the states, and **transition** from one state to another carry the pattern of a trajectory.

Non-parametric model

- Probability the a state appear at certain location $P(x_t)$
- Transition probability $P(x_t | x_{t-1})$
- Discretized embedding space

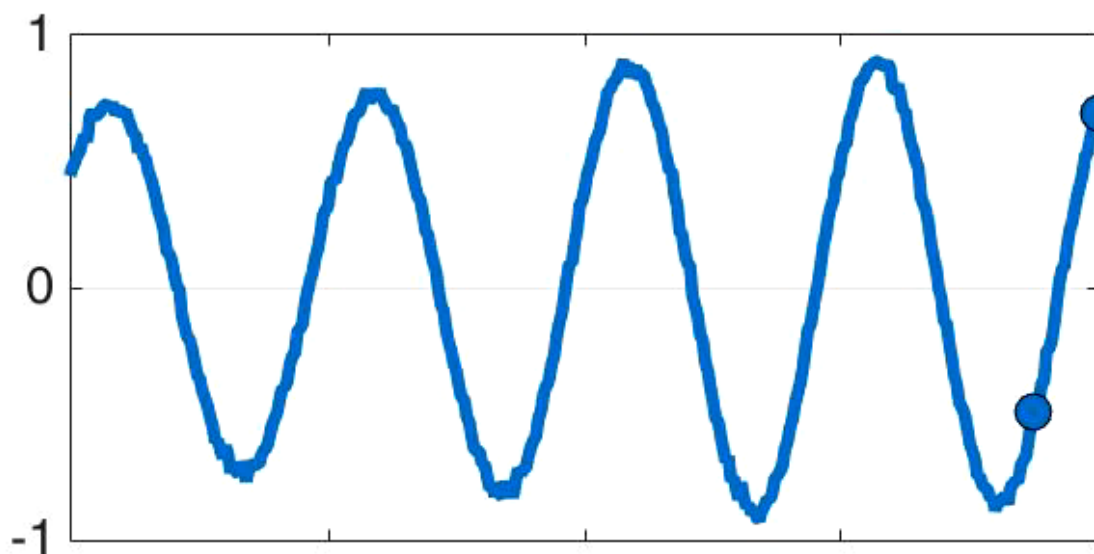
$$\begin{aligned}
 S_{\text{MGM}}(X) &= \sum_{j=1}^t P(x_j) \prod_{i=2}^t P(x_i | x_{i-1}) \\
 &= S_G(X) \times S_M(X)
 \end{aligned}$$

Markov Geographic Model (MGM)

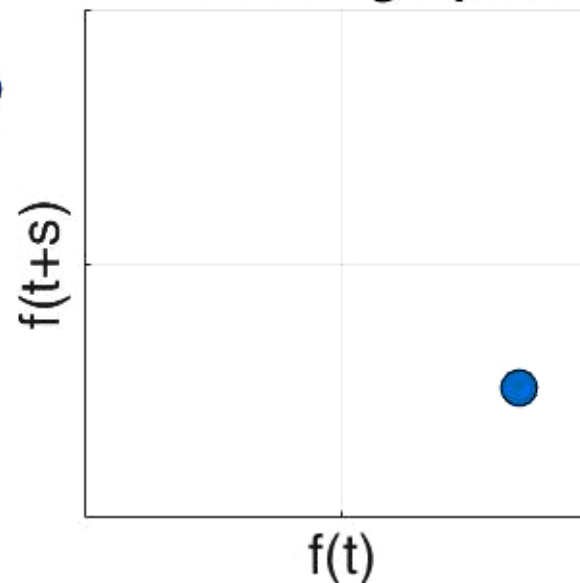
Online update the **transition** and **distribution** of states.

↓
Markov process

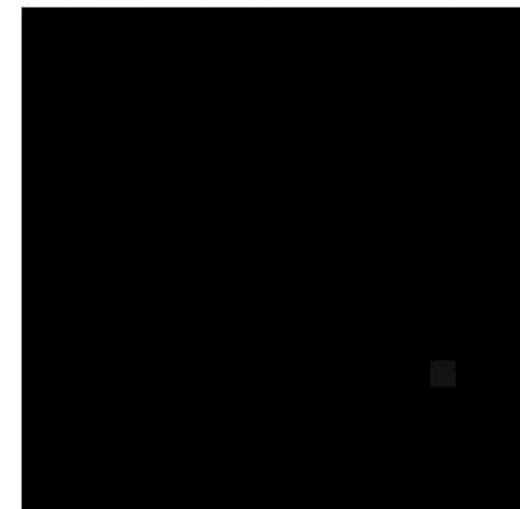
↓
Geographic distribution



Embedding Space



Geographic Distribution



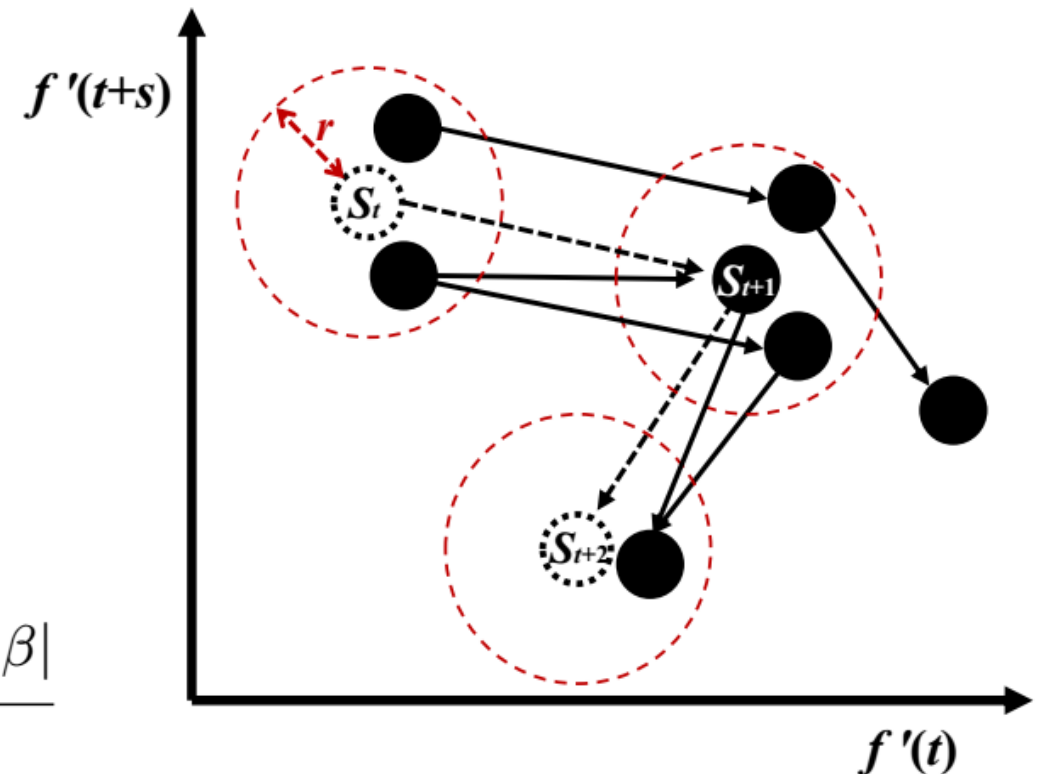
Neighborhood Matching

$$\begin{aligned}
 S_{\text{MGM}}(X) &= \sum_{j=1}^t P(x_j) \prod_{i=2}^t P(x_i | x_{i-1}) \\
 &= S_G(X) \times S_M(X)
 \end{aligned}$$

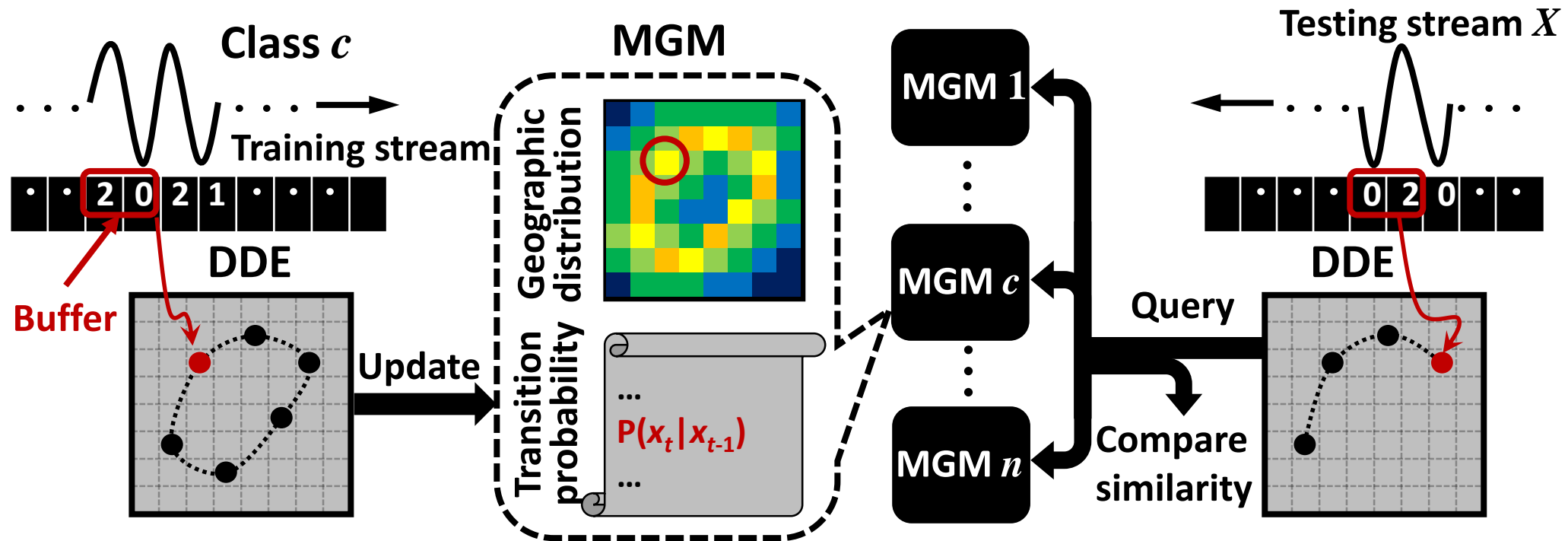
Make the transition probability more robust to **noise** and **unseen** samples in testing.

$$S_M(X) = \prod_{i=2}^t \frac{\sum_{\alpha \in N_r(\Phi'(x_i)), \beta \in N_r(\Phi'(x_{i-1}))} |\alpha; \beta|}{\sum_k \sum_{\gamma \in N_r(\Phi'(x_{i-1}))} |\Phi'(x_k); \gamma|}$$

$N_r(\Phi'(x_i))$ --- the set of neighbors within radius r around $\Phi'(x_i)$



Online Modeling and Classification by DDE-MGM



Experimental Results

Datasets:

- **UCI Character Trajectory --- 2858** character samples of 20 classes, x and y axes were recorded.
- **MSR Action3D --- 567** action samples of 20 classes performed by 10 subjects, human skeleton is recorded.

Normalized
Well aligned

Misaligned
Variant length

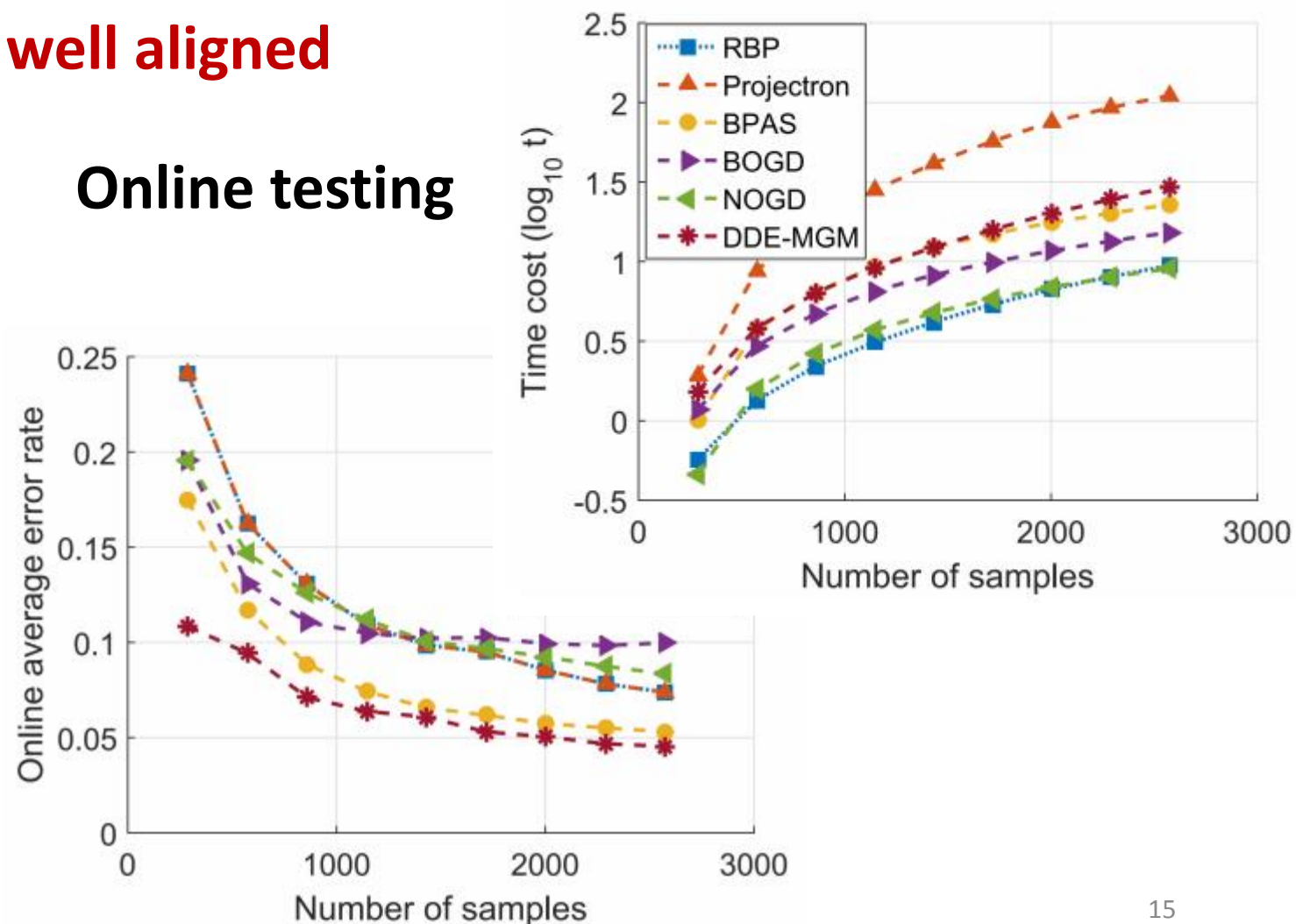
Experimental Results --- UCI Character Trajectory

The data is **normalized** and **well aligned**

Half-vs-half validation

Algorithm	Accu. (%)	Time (sec)
1NN-DTW	91.37	3.9×10^4
SAX	89.96	128.85
HMM	57.89	7.4×10^3
DDE-MGM	92.07	34.21
RBP	92.62	9.44
Projectron	92.62	110.26
BPAS	94.68	22.81
BOGD	90.02	15.24
NOGD	91.65	9.04
DDE-MGM	95.45	63.92

Online testing



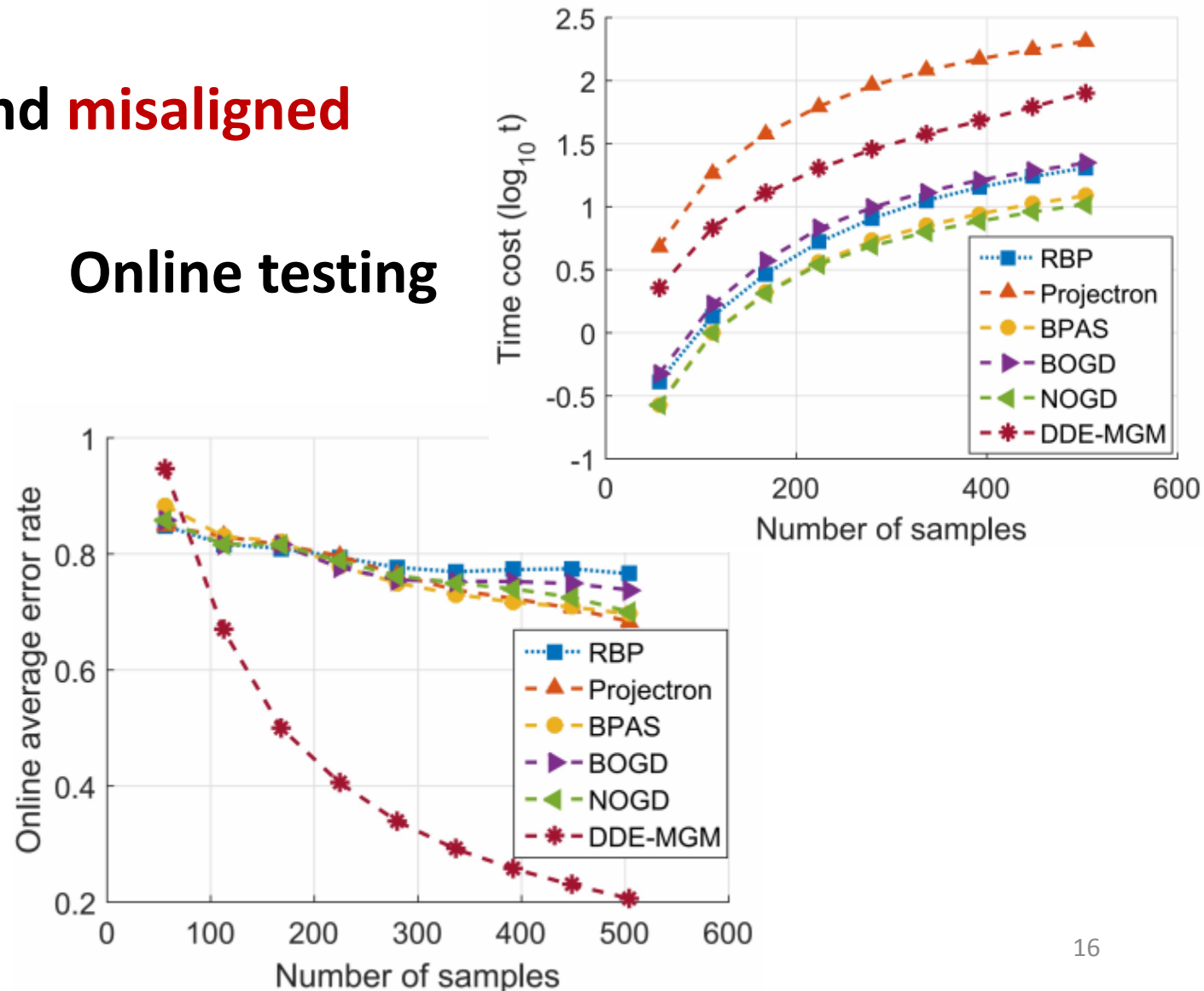
Experimental Results --- MSR Action3D

The data is **not normalized** and **misaligned**

Half-vs-half validation

Algorithm	Accu. (%)	Time (sec)
1NN-DTW	74.73	7.6×10^4
SAX	61.90	54.68
HMM	60.07	2.1×10^3
DDE-MGM	93.04	28.40
RBP	23.41	20.23
Projectron	31.65	205.25
BPAS	30.36	12.25
BOGD	26.19	22.23
NOGD	29.96	10.47
DDE-MGM	79.37	80.38

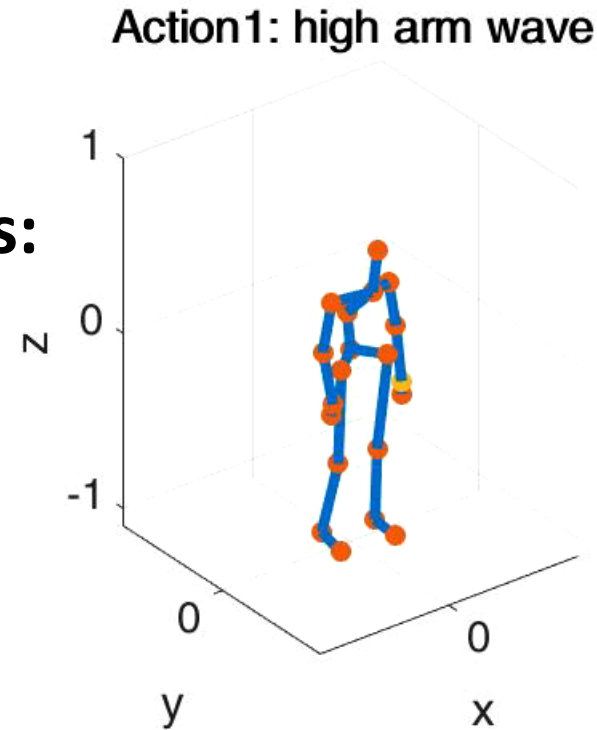
Online testing



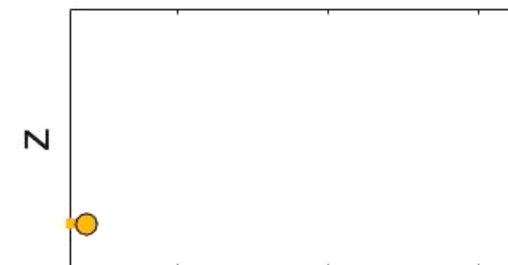
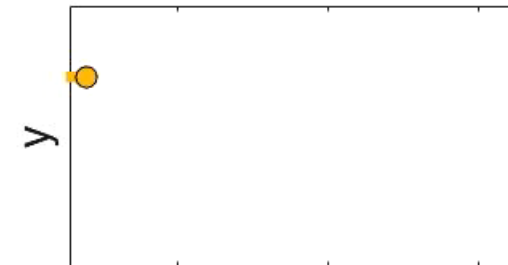
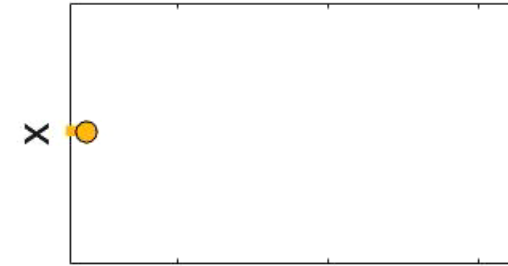
An Example of Action Recognition

The joint of **left wrist** is plotted for three categories of actions shown in different colors:

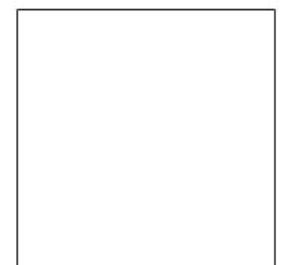
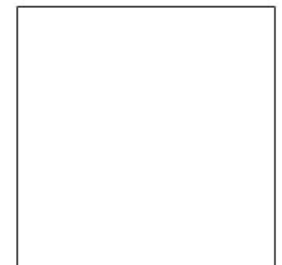
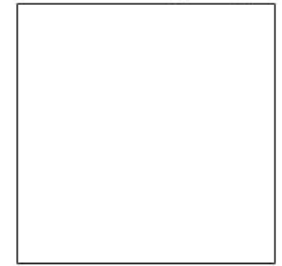
- **high arm wave**
- **horizontal arm wave**
- **hammer**



Time Series



Embedding Space



Conclusion

- **DDE is introduced to solve misalignment in online modeling**
- **The non-parametric model MGM is proposed to model the trajectories in an online manner**
- **Both modeling and classification are achieved in real time.**

T THANKS

SIGIR
Special Interest Group
on Information Retrieval

Student Travel Grant

Appendix: Parameter Setting of DE

$$\Phi(x_t; s, d) = (y_t, y_{t+s}, \dots, y_{t+(d-1)s})$$

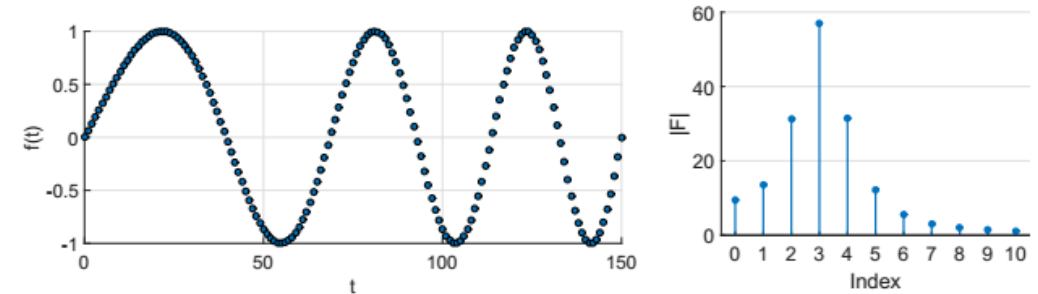
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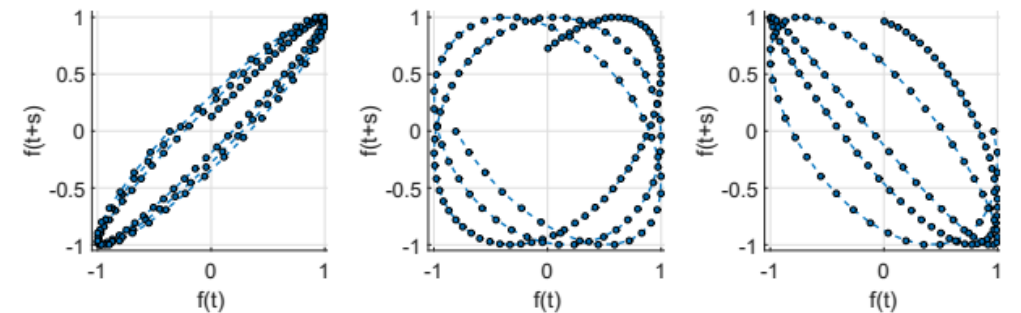
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y_t --- observation (time series) at the time t



(a) Time series

(b) FFT



(c) $s = 2$

(d) $s = 12$

(e) $s = 25$

d --- False nearest neighbor [\[M. Kennel et al., 1992\]](#)

s --- $2\pi \times d \times s \times \frac{f}{f_s} \equiv 0 \pmod{\pi}$ [\[J. A. Perea and J. Harer, 2013\]](#)