

# Surrogate Loss Learning for Dynamic Time Warping(DTW)

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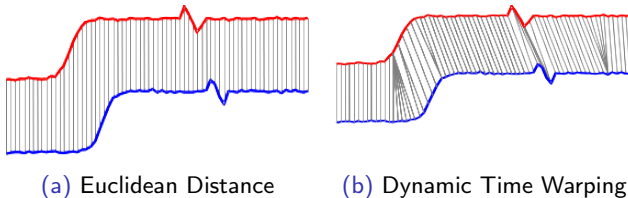
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# Agenda

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# Motivation



- **Dynamic Time Warping (DTW)** is a technique used to quantify the similarity between two time sequences varying in speed and length.
- DTW is used as an evaluation metric in time series forecasting.
- Can be applied to irregular time series.

# Motivation

DTW Method:

- Calculate for each point in the series an easy distance (e.g. Euclidean).
- Iterate back through the series and add also the distance to the nearest a adjacent point.

For any two time-series,

$$X = (x_1, x_2, \dots, x_N)$$

and

$$Y = (y_1, y_2, \dots, y_M)$$

we compute a distance(cost) matrix as:

$$c_{i,j} = d(x_i, y_j) + \min\{c(i-1, j-1), c(i, j-1), c(i-1, j)\}$$

# Motivation

Research has manifested that we can use a metric as a loss function.

**Problem:** *DTW is non-differentiable because of ' $\min$ ' operation.*

**Solution:** *Surrogate Losses are a superior option.*

Goal: Develop a surrogate model which depicts DTW loss function as a meta-level neural network.

- Surrogate Loss network will be trained before forecasting model.

# Literature Review

In time series forecasting following loss functions are used:

- MAE:
- MAPE
- MSE
- soft-DTW \*

\*soft-DTW: a differentiable loss function version of DTW which is highly dependent on hyperparameter  $\lambda$ .

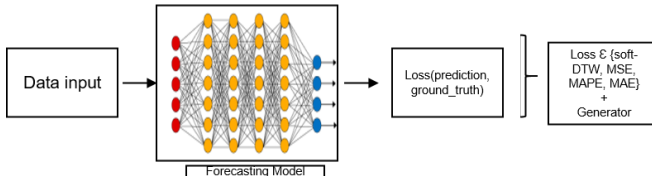
Given as:  $dtw_{\gamma}(x, y) := \min_{\gamma} \{ \langle A, \Delta(x, y) \rangle, A \in A_{n,m} \}$ .

# Research Idea

## Pre-processing

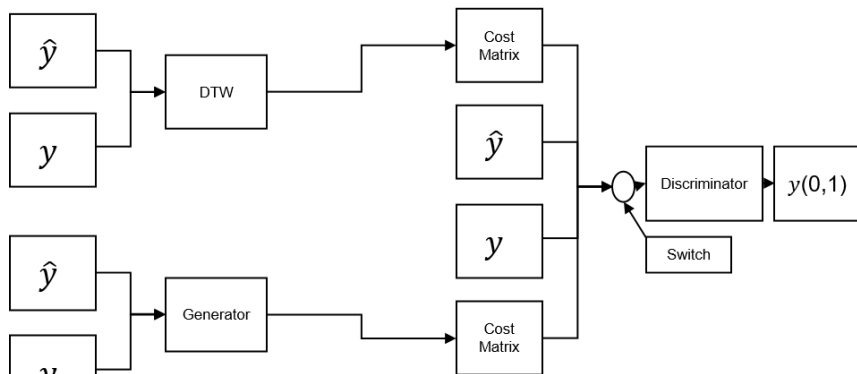


## Baseline Idea



# Research Idea

Proposed Model for Surrogate Loss Network- GAN:

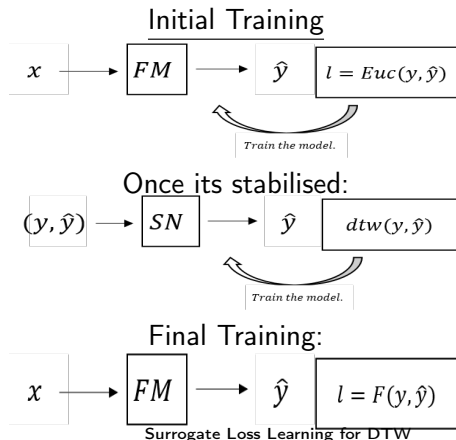




# Research Idea

*Meta-data for training Surrogate Loss Network:*

Forecasting Model  $\rightarrow$  FM, Surrogate Function  $\rightarrow$  SF, Input  $\rightarrow$  X,  
Prediction  $\rightarrow$  Y



# Data Foundation

Public Data-sets:

- Traffic
- Electricity
- Retail

Private/ Company Data-set:

RBB Data-streams

# Data Foundation

Dataset	Description	Prediction horizon	Properties	Pre-processing
Electricity Dataset	UCI Electricity Load Diagrams Dataset, containing electricity consumption of 370 customers	Use past week's data to forecast over the next 24 hours.	Target Type: R No. Of columns: 370 No. Of samples: 500k	Aggregate on hourly level.
RBB Data-streams	Time-related sensed data from production machinery.	Use past year's data to forecast over the next week.	Target Type: R No. Of columns: 165 No. Of samples: 17k	Aggregate on daily level.
Traffic Dataset	UCI PEM-SF Traffic Dataset which describes occupancy rate of 440 San Francisco Bay Area Freeways.	Use past week's data to forecast over the next 24 hours.	Target Type: [0, 1] No. Of columns: 370 No. Of samples: 500k	Aggregate on hourly level.
Retail Dataset	Favorita Grocery Sales Dataset from the Kaggle competition is a metadata for different products and the stores sampled at the daily level.	Use 90 days of past data to forecast product sales of 30 days into the future.	Target Type: R No. Of columns: 130k No. Of samples: 500k	Use daily data.

# Evaluation

## Performance Evaluation:

- We evaluate the performances of proposed model, and compare it against above mentioned baselines.
- The results are evaluated using following metrics:
  - 1 DTW
  - 2 MSE

# References

- Cuturi, Marco, and Mathieu Blondel. Soft-DTW: a differentiable loss function for time-series, International Conference on Machine Learning. PMLR, 2017.
- Lima Bryan, Sercan O. Arık, Nicolas Loeffb, Tomas Pfister. Temporal Fusion Transformers for Interpretable Multi-horizon Time Series Forecasting, <https://arxiv.org/pdf/1912.09363.pdf>.
- Grabocka, Josif, Randolph Scholz, and Lars Schmidt-Thieme. Learning surrogate losses, arXiv:1905.10108 (2019).
- Lamb Alex et al, Professor Forcing: A New Algorithm for Training Recurrent Networks 2016(<https://arxiv.org/abs/1610.09038>)
- Electricity Data-set:  
<https://archive.ics.uci.edu/ml/datasets/ElectricityLoadDiagrams20112014>

# References

- Traffic Data-set: <http://pems.dot.ca.gov/>
- Favorita Data-set:  
<https://www.kaggle.com/c/favorita-grocery-sales-forecasting/data>
- RBB-Dataset: MunichRE

# Timeline

