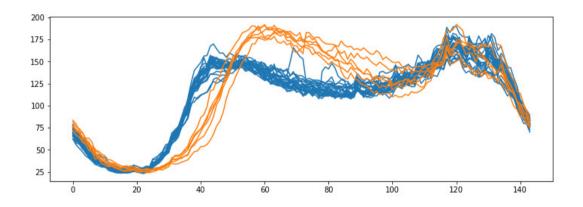
Advanced Data Mining

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- □ How to classify time series?
 - see the jupyter python notebook with some examples



- **APPROACH 1:** (for regular time series)
 - define patterns of daily (or weekly, monthly, annual) profiles for each class
 - each day (or week, month, year) time series can be matched to these patterns
 - the time series represents the class of the best matched pattern

□ How to classify time series?

APPROACH 2:

- consider the time series as a vector of numbers, apply one of classic classification algorithms, e.g. kNN, SVM, etc.
- **□** Difficulties:
 - time series may be of various length
 - similar, but shifted or rescaled time series will lead to dissimilar vectors

APPROACH 3:

as above, but use the DTW distance measure

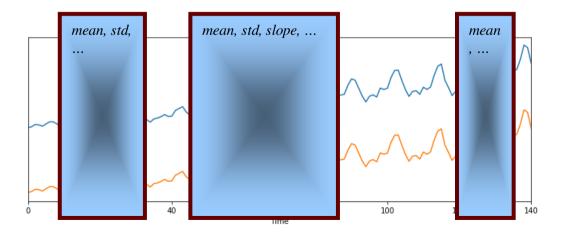
□ How to classify time series?

APPROACH 4:

- consider the time series in a feature-based representation
- **□** Time Series Forest

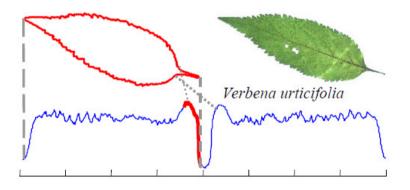
H. Deng, G. Runger, E. Tuv, M. Vladimir, "A time series forest for classification and feature extraction". Information Science 239, 2013, pp.142-153.

- TSF extends the **Decision Tree** and **Random Forest** classifiers
- the time domain of the time series is splitted into a number of intervals
- for each interval, a given set of features is evaluated (e.g. mean, std, slope)
- this gives a vector of features describing the time series
- similar approaches based on XGBoost or DNN



- □ How to classify time series?
 - APPROACH 5:
 - shapelet-based representation

L. Ye, E. Keogh, "Time Series Shapelets: A New Primitive for Data Mining". International Conference on Knowledge Discovery and Data Mining, 15, 2009, pp.947-956.



- ☐ Generally: *a shapelet is a characteristic pattern in time series*
- □ Formally:
 - \blacksquare a shapelet S is a sequence $s_1, s_2, ..., s_L$
 - for each time series X, we can evaluate the distance between S and X, e.g. DTW(X, S), to verify whether the shapelet S is characteristic to the time series X (and how much)
- Therefore, for a given set of shapelets, time series may be encoded in the shapelets-based representation.

□ How to classify time series?

APPROACH 5:

brute-force algorithm for discovering valuable shapelets

L. Ye, E. Keogh, "Time Series Shapelets: A New Primitive for Data Mining". International Conference on Knowledge Discovery and Data Mining, 15, 2009, pp.947-956.

- consider a train dataset D composed of N time series $X_1, X_2, ..., X_N$ labelled with target values $y_1, y_2, ..., y_N \in \{1, 2, ..., C\}$, where C is the number of classes
- for the sake of simplicity, consider the binary classification problem (C = 2)
- similarly to the **Decision Tree** classifier, for a shapelet S and a distance threshold d, the entire train dataset may be splitted into two parts with the condition

- and the information gain (e.g. based on entropy) may be evaluated
- for a shapelet S, the optimal split point is the distance threshold maximizing the information gain among all distance thresholds
- the information gain of a shapelet S is the information gain for the shapelet S and its optimal split point
- the brute-force algorithm considers the set of all possible shapelets of a given length and searches for the shapelet with the maximum information gain
- it may be easily extended to find more than one shapelet and to general classification problems (C > 2)

- □ How to classify time series?
 - APPROACH 5:
 - **brute-force algorithm for discovering valuable shapelets**

L. Ye, E. Keogh, "Time Series Shapelets: A New Primitive for Data Mining". International Conference on Knowledge Discovery and Data Mining, 15, 2009, pp.947-956.

```
candidates := Candidate-Shapelets(maxlen, minlen)
best_gain := 0
for each S in candidates
    gain := Information-Gain(S)
    if gain > best_gain then
    best_gain := gain
    best_shapelet := S
return best_shapelet
```

possible improvements by subsequence distance early abandon and admissible entropy pruning

□ How to classify time series?

APPROACH 6:

□ try to define the shapelets in a learning process

J. Grabocka, N. Schilling, M. Wistuba and L. Schmidt-Thieme, "Learning Time-Series Shapelets". International Conference on Data Mining, 14, 2014, pp.392-401.

