

#### Analyzing Sequential User Behavior on the Web

Tutorial @WWW2016

### **About Us**











#### **Tutorial Website and Material**

Website:

# sequenceanalysis.github.io

- Slides (to be uploaded)
- Jupyter notebooks:
  - Download and run/edit on your own computer
  - View the result on nbviewer
  - Virtual environment on mybinder

#### Structure of this Tutorial

- Introduction & Overview
- Sequential Pattern Mining



- Break -

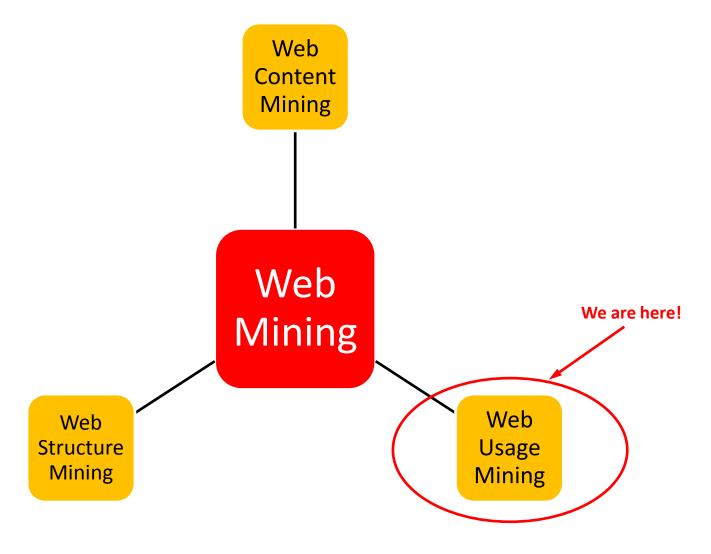
- Markov Chain Modeling
- Comparison of Hypotheses on Sequences



#### Part 1

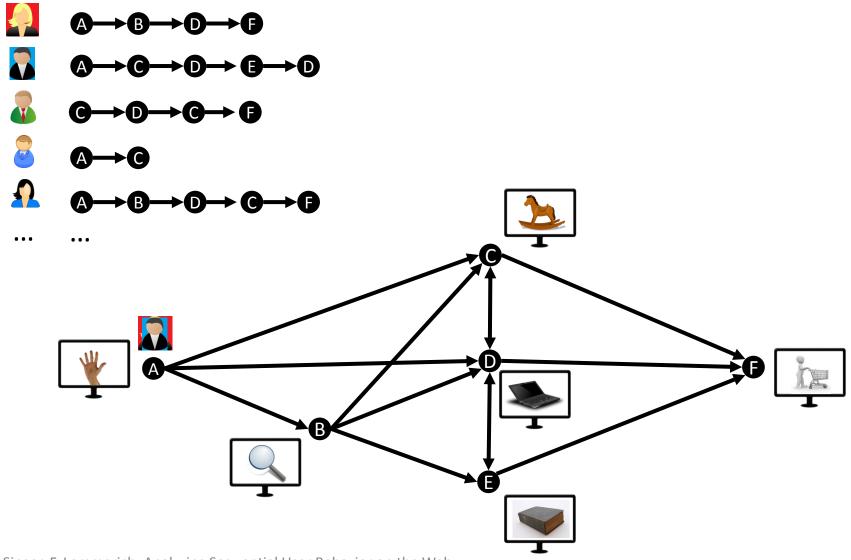
A Short Introduction to Categorical Sequences on the Web

### Web Mining [Srivastava 2000]

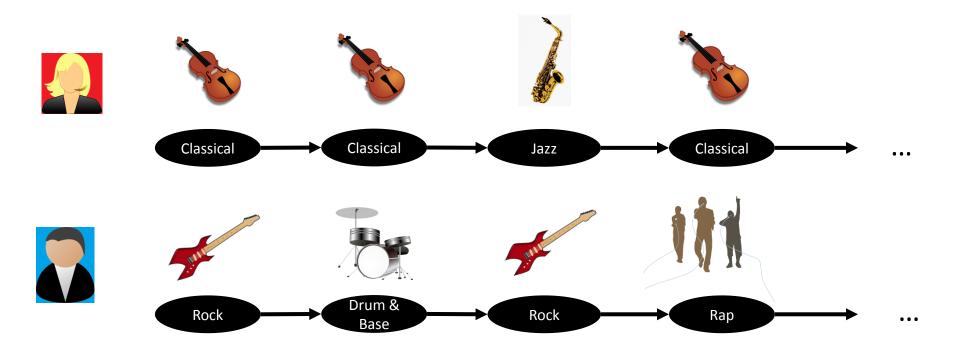




### Example: Navigation through the Web



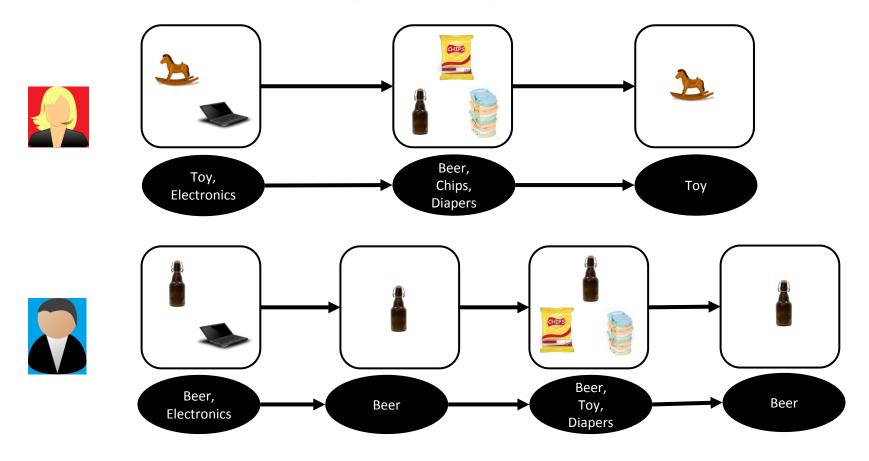
### **Example II: Listening History**





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### **Example III: Shopping History**

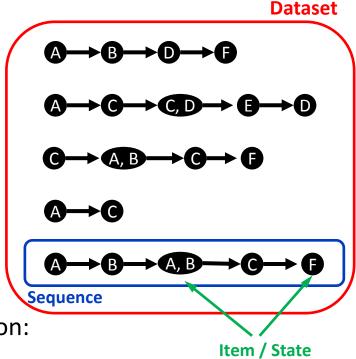




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#### Data Covered in this Tutorial

- Dataset is given by a set of sequences
- Each sequence contains several events X
- Each event in a sequence has...
  - Exactly one categorical variable (state)
    (Modeling, Hypotheses Comparison)
  - Multiple Binary variables (items)
    (Sequential Pattern Mining)
- We do not cover methods using more information:
  - Numeric/ordinal variables each event
  - No time stamps (only ordering)
  - == NO time series analysis
  - Text



#### **Data Sources**

Web Server Logs (e.g., Apache logs)

- Cookies
- Explicit user input
- Client-side tracking (modified browsers, eye-tracking)
- Web APIs (e.g., greddit or Wikipedia) or scraping:
  - Maybe not capture user actions directly
  - Results/edits form sequences

#### Data Pre-processing of Web Logs [Chitraa et al. 2010]

- Data Cleaning, e.g.
  - Remove access to single images
  - Errorneous requests (http errors)
- User identification (usually based on IP address)
- Session identification
  - Time-oriented heuristics
  - Navigation-oriented heuristics
- Path completion: accounts for proxy / caching effects

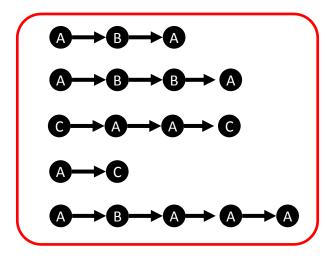
#### Tasks for Sequential Data

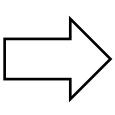
- Sequence Clustering
- Sequence Classification
- Sequence Prediction
- Sequence Labeling
- Sequence Segmentation
- Sequential Pattern Mining
- Sequence Modeling
- Hypotheses Comparison on Sequences

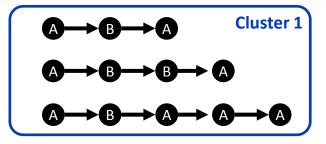


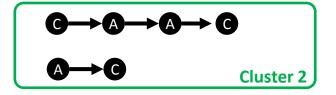
#### Sequence Clustering: Task

"Find groups in the sequence dataset such that sequences within one group are similar and sequences in different groups are dissimilar"



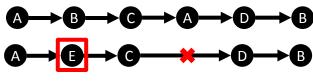






#### Sequence Clustering: Method Overview [Xu & Wunsch, 2005]

- Clustering based on sequence similarity
  - E.g., edit distance (Levenshtein distance):
    Number of transformation operations



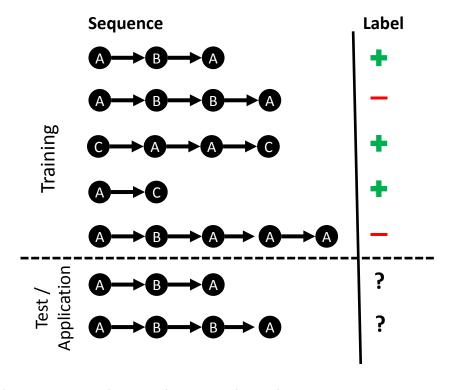
Edit distance: 2

- Can apply hierarchical clustering, density-based clustering, ...
- Indirect clustering: Extract features first
  - Features: all n-grams, sequential patterns
  - Use (classical) vector-spaces clustering on these features
- Statistical sequence clustering / model based clustering
  - Use set of Hidden Markov Models (HMM)
  - Each model "generates" the sequences of one cluster
  - EM algorithm optimizes clusters and sequence-cluster mapping



### Sequence Classification: Task

"Given a training dataset of labeled sequences, predict the labels of future sequences"



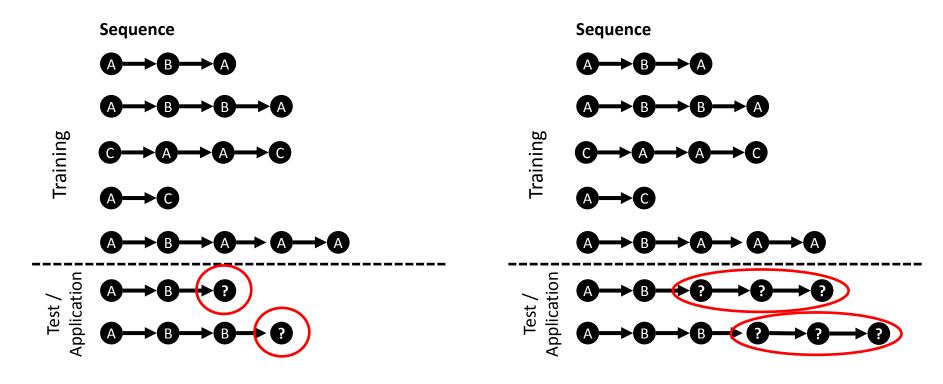
#### Sequence Classification: Methods [Xing et al. 2010]

- Use sequence similarity measure
  - See sequence clustering
  - Apply k-nearest-neighbor for classification
- Indirect classification: extract features first
  - See sequence clustering
  - Apply any classification method
  - SVM with string kernels:
    do not compute the features explicitly, but only use a kernel instead
- Model-based classification
  - Discriminatively trained Markov Models
  - Different variations of Hidden Markov Models



### Sequence Prediction / Sequence Generation: Task

"Given a set of sequences and some incomplete sequences, how will the new sequences continue?"



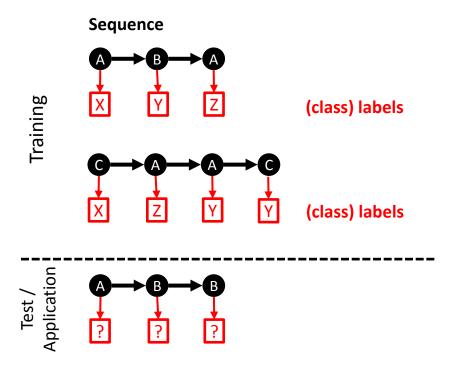
#### Sequence Prediction: Methods

- Apply (Hidden) Markov Models
- (Partially ordered) Sequential rules (based on sequential patterns)
- Recurrent Neural Networks (RNNs)



### Sequence Labeling: Task

"Given a set of sequences with labels for each event, predict the labels of new (unlabeled) events"



#### Sequence Labeling [Nguyen & Guo 2007]

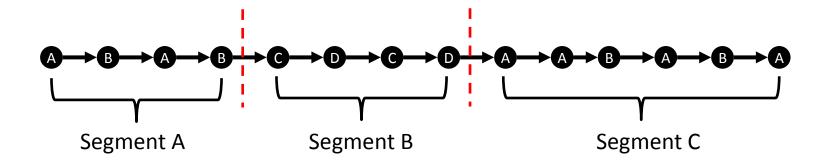
More typical for Natural Language Processing
 E.g., part of speech tagger, reference extraction, ...

- Methods:
  - Hidden Markov Models [Rabiner 1989]
  - Conditional Random Fields [Laferty et al. 2001]
  - SVM-Struct [Tsochantaridis et al. 2005]
  - **—** ...



#### Sequence Segmentation

"Partition a sequence into segments such that the segments are as homogeneous as possible"



#### Sequence Segmentation: [Terzi & Tsaparas 2006]

- Applications:
  - Detect behavioral stages of web users
  - DNA segmentation
  - Text segmentation
- Methods:
  - Given time information: similar to discretization
  - Models + MDL [Kiernan & Terzi 2009]
  - Set of models, optimizes (log-) likelihood [Yang et al. 2014]

#### Tasks for Sequential Data

- Sequence Clustering
- Sequence Classification
- Sequence Prediction
- Sequence Labeling
- Sequence Segmentation
- Sequential Pattern Mining
- Sequence Modeling
- Hypotheses Comparison on Sequences

#### **Human Navigation**

- User Navigation from Web logs [Catledge & Pitkow 1995]
- Strong regularities in WWW surfing [Huberman et al. 1998]
- Mining longest repeating subsequences for prediction [Pitkow & Pirolli 1999]
- Information scent theory [Chi et al. 2001]
- Navigation in Wikipedia
  - Human wayfinding in information networks [West & Leskovec 2012]
  - Automatic vs. Human Navigation [West & Leskovec 2012-2, Trattner et al. 2012]
  - Memory and structure [Singer et al 2014]

#### Detecting a-typical Surfing Behavior

- Characterizing (a-)typical user behavior [Sadagopan & Li 2008]
  - Model sequences with Markov chains
  - Detect improbable sequences
  - Characterize outliers manually
- Sybil (Fake identity) [Wang et al 2013]
  - Visualize transition probabilities in Markov chains
  - Use SVM/similarity based approaches for classification

#### Further Application Areas [Facca & Lanzi 2005]

- Improved website design
- Personalization of web content [Pehtaa et al 2012, Andersson 2002, Eiriniki et al 2003]
  - Recommending links
  - Personalized site maps
- Pre-fetching and caching [Patil & Patil 2015, Wu & Chen 2002]
- E-commerce / customer relation ship management [Bounsaythip & Rinta-Russala 2001, Ansari et al. 2001, ]
- Identifying relevant websites [Bilenko & White 2008]

• ...

#### Privacy: Ethical and Legal issues

- Ethical issues:
  - Web Usage Mining exploits user data, often no (conscious) agreement
  - User are judged based on group characteristics instead of individual merit

[Van Wel & Royakkers 2004]

- Legal issues: [Velazquez 2013]
  - Depends on the country
  - "contracts are the main legal tool to protect users' privacy, therefore affirming informed consent as being the key concept in deploying a suitable privacy policy."
  - Key question:
    Is the IP address personal data (personally identifiable information):
    Example Germany: "potentially personal data"
  - Also different regulations for academic and commercial use

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