

SIMPDA 2017 - Neuchâtel (Switzerland)

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Discovering Customer Journey Maps using a Mixture of Markov Models

Customer journey map

XES - Standard

Challenge

E-M Markov models

Case study

Feature 1

Feature 2

Feature 3

Feature 4

Conclusion

Customer Journeys.

Path followed by a customer to consume a service.





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Customer Journeys: Capture the Customer Experience.

Customer journeys are becoming increasingly complex

- Numerous channels / devices
- Increasing number of interactions

« Improving understanding of customers has been ranked one of the most important research challenges in the coming year [1] »



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CJMs: A customer-centric approach.

Customer Journey Maps (CJMs) are used to understand, discuss, or improve the main paths in the usage of a service

Mostly used as a design thinking tool / strategic tool



Getting information Trying

Buying

Sharing experience





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Strategy vs. Reality.

« People don't behave like robots, and no matter how well we craft an experience, they will not perceive exactly as we anticipate or hope » [2]



[2] Richardson, A.: Series on customer journey: Using customer journey maps to improve customer experience, using customer journey maps to improve customer experience, touchpoint bring the customer experience to life. Harb Bus Rev (2010)



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XES: Discovering CJMs from Event Logs.

- We can leverage these traces to provide fact-based insights on customer journeys Inspired from Process Mining
- INPUT: The XES standard is adequate to store CJMs [3]

Journey	Activity	Timestamp	Customer
1	Getting information	10.05.2016	Z.Davis
	Trying	10.05.2016	Z.Davis
	Buying	10.05.2016	Z.Davis
	Sharing experience	10.05.2016	Z.Davis
2	Getting information	11.05.2016	O.Palmer
	Trying	11.05.2016	O.Palmer
	Buying	11.05.2016	O.Palmer
	Comlaining	12.05.2016	O.Palmer

• When customers interact with a service, it (often) leaves traces

[3] Bernard, G., & Andritsos, P. (2017). A Process Mining Based Model for Customer Journey Mapping.





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1 Event logs (e.g., XES) В \square Α 2 3 4 5 E C G A B D A B D F E C G C A

Challenge: Summarize the Logs using K Representative Journeys.





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Customer journey Customer journey map **XES - Standard** Challenge E-M Markov models **Case study** Feature 1 Feature 2 Feature 3 Feature 4 Conclusion

Markov models: Modelling sequences.

« In probability theory, a Markov model is a stochastic model used to model randomly changing systems. It is assumed that future states depend only on the current state, not on the events that occurred before it. » - Paul A. Gagniuc



First probability vector



Gagniuc, Paul A. (2017). Markov Chains: From Theory to Implementation and Experimentation. USA, NJ: John Wiley & Sons. pp. 1–256. ISBN 978-1-119-38755-8.





Expectation-Maximization on a Mixture of Markov Models.



Optimizing the log-likelihood of the data



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Case study: The Dataset.

An activity-based travel survey conducted in the Chicago metropolitan area over a demographic representative sample of its population.

29,542 journeys - **2,381** are unique

"being at home" \rightarrow "going to school" \rightarrow "having a meal" \rightarrow "being at home"

16 types of events

Average number of events per journey: 4.8

http://www.cmap.illinois.gov/data/transportation/travel-survey





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Case Study: Results.

Activities:

LEGEND





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Feature 1: Each Model is Responsible for each Journey.



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Feature 2: Limited Number of Parameters.

As input:



Parameters set by a human can have an **important impact** on the representative journeys. The E-M on a mixture of Markov models allows to discover the representative journeys much more **naturally**.



K - Which can potentially be computed upfront thanks to information criterion technics



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Feature 3: Computing the Next Most Probable Event.

Activities:

Visiting friends

Entertainment

Work

Home activities





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Feature 4: Handling Shifting Behaviors - Soft Clustering.

С

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Model 3





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Feature 4: Handling Shifting Behaviors - Soft Clustering.







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Kudelski application

Probabilistic approach robustness

Research in progress!

Q&A