

# Multidomain/multichannel Sequence Analysis

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Sequence Analysis Association (SAA)

<https://sequenceanalysis.org>

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# Outline

- 1 Multidomain/multichannel sequence analysis
- 2 Association between domains
- 3 Multidomain typologies
- 4 Visualizing groups of MD sequences
- 5 Illustration with Swiss biographic data
- 6 Dyadic/Polyadic Data

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# Classical Sequence Analysis

- Abbott and Tsay (2000) describe typical SA as a 3-step program
  - 1 Coding narratives as sequences
    - state tokens (eg., {work, no work} vs {Full-time, Part-time, Unempl., No activity} )
    - time granularity (e.g., yearly vs monthly )
  - 2 Computing pairwise dissimilarities between sequences
  - 3 Analyzing sequences based on their dissimilarities
    - building typologies by means of clustering
    - studying how clusters relate to covariates (sex, social origin, ...) by means of regression models

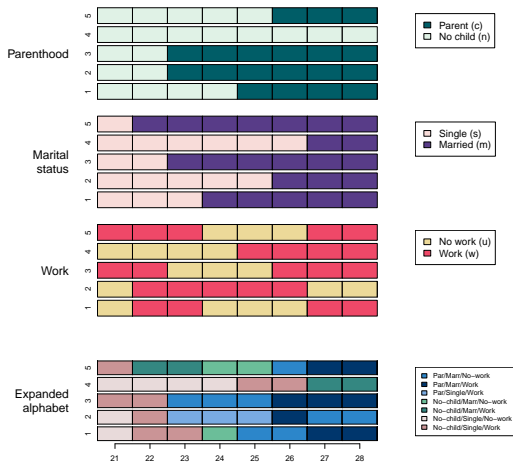
# SA is much more (Liao et al., 2022)

- Visualization
- Multichannel (Multidimensional) sequences
- Dissimilarity-based analysis different from clustering (SOM, Anova-like, regression trees, representative sequences, ...)
- Non-dissimilarity-based clustering (LCA, Mixture of HMM, ...)
- Characteristics of individual sequences (Entropy, Complexity, Badness, ...)
- Sequence network
- Probabilistic sequence models (HMM, PST, ...)
- Combining SA with related methods (EHA, HMM, Multistate models, ...)

# Multidomain SA: What is it?

- Analyzing how life trajectories:
  - in different domains: family, work, health, ... (Bernardi et al., 2019)
  - of linked people: partners, family members, ... (Elder et al., 2003) jointly unfold.
- Measuring association between domains of life (Piccarreta, 2017)
- Identifying spill over effects across domains (or linked people).

# Multidomain data: toy example data



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# Association between domains

- Joint analysis of domains of interest when domains are associated,  
i.e., when there are relationships between domains.
- Two types of association:
  - **State association**: when state tokens of one domain tend to co-occur with tokens of the other domain.
    - Measured by cross-tabulating token occurrences of the two domains
  - **Trajectory association**: when trajectories in one domain tend to co-occur with trajectories in the other domain.
    - Measured by correlation between pairwise distances in each of the domains.

# Contingency tables of state occurrences

Toy MD sequences

	<i>s</i>	<i>m</i>
<i>c</i>	3	16
<i>n</i>	14	7

$$v = 0.51^{***}$$

	<i>u</i>	<i>w</i>
<i>c</i>	8	11
<i>n</i>	9	12

$$v = 0.0076$$

	<i>u</i>	<i>w</i>
<i>s</i>	6	11
<i>m</i>	11	12

$$v = 0.13$$

Note: Degree of significance: \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Association between domains: Toy MD sequences

Toy MD sequences. Association between domains. State association (Cramer's  $v$ ) and trajectory association (Pearson's  $r$  correlation between pairwise INDELSLOG-based OM distances in each domain)

	Cramer's $v$	$p(v)$	Pearson's $r$	$p(r)$
Parent with Married	0.51	0.00	0.08	0.83
Parent with Work	0.01	0.96	0.05	0.88
Married with Work	0.13	0.43	0.35	0.32

# Computing association measures between domains

## Association between states

```
load(file = "data/toyMD.Rdata")
dnames <- c("Parent", "Married", "Work")
channels <- list(s.ch, s.mar, s.work)
seqdomassoc(channels, dnames = dnames)
```

```
##                df      LRT p(LRT)      v      p(v)
## Parent_with_Married  1 11.24061 0.0008 0.5140 0.00115
## Parent_with_Work    1  0.00231 0.9617 0.0076 0.96169
## Married_with_Work   1  0.63265 0.4264 0.1253 0.42802
```

# Computing association measures between domains

## Association between trajectories

```

cost.meth <- "INDELSLOG"
toy.met = "OM"
d.ch <- seqdist(s.ch, method = toy.met, sm = cost.meth)
d.mar <- seqdist(s.mar, method = toy.met, sm = cost.meth)
d.work <- seqdist(s.work, method = toy.met, sm = cost.meth)
domdiss <- list(d.ch, d.mar, d.work)
disssdomassoc(domdiss, dnames = dnames)

## $Pearson
##      Parent Married  Work
## Parent 1.0000  0.080 0.0549
## Married 0.0800  1.000 0.3481
## Work    0.0549  0.348 1.0000
##
## $p.pearson
##      Parent  Married  Work
## Parent 1.69e-131 8.26e-01 8.80e-01
## Married 8.26e-01 1.99e-128 3.24e-01
## Work    8.80e-01 3.24e-01 2.97e-128
##
## $Pearson.Rsquare
##      Parent Married  Work
##      0.0047 0.0638 0.0621
##
## $dnames
## [1] "Parent" "Married" "Work"
##
## attr( "class" )
## [1] "list" "ddomassoc"

```

# Summary of domain association between domains

```
dass.traj <- disssomassoc(domdiss, dnames = dnames, what = "all")  
summary(dass.traj)
```

##	Pearson	p.Pearson	Spearman	p.Spearman
## Parent_with_Married	0.0800	0.826	0.1325	0.715
## Parent_with_Work	0.0549	0.880	0.0248	0.946
## Married_with_Work	0.3481	0.324	0.3676	0.296

# Outline

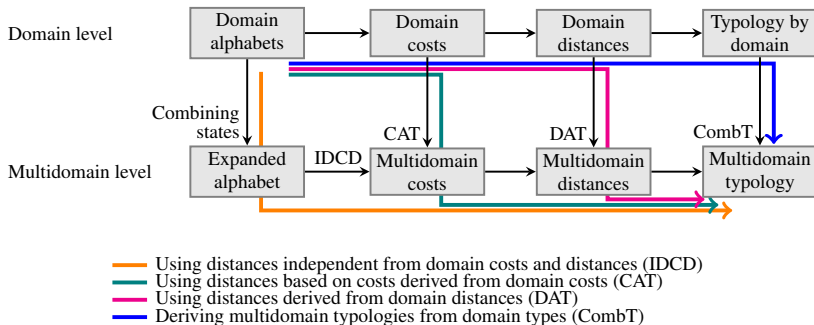
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# Methods of multidomain SA

- As in classical SA, multidomain SA proceeds by building a typology
- In multidomain SA, we need a **joint typology** of the different domains
- Multidomain SA consists then in studying how the trajectories in the different domains are related within the different types (clusters).
  - For example: insecure work trajectories may typically co-occur with unstable family life trajectories and bad health pathways.
- Comparing the trajectories of the different domains within the clusters is generally done graphically: need for specific **visualization tools** of multidomain sequences.



# Strategies for building a multidomain typology



- Other approaches based on domain distances: MD distances derived from quantitative representations (principal coordinates, MDS) of sequences for each domain. (Example: GIMSA, Robette, Bry, and Lelièvre, 2015)

# Independence constraints of additive tricks (CAT, DAT)

- CAT (Pollock, 2007; Gauthier et al., 2010)  
assumes independence across state tokens of the different domains:
  - Substitution cost between “single + no child” and “married + child” same as between “single + child” and “married + no child”
  - When domain costs are constant (e.g., all equal to 2), resulting multidomain substitution costs are proportional to number of non-matching domains.  
“single + no child” and “married + no child” differ on one domain.
- DAT (Han and Moen, 1999)  
assumes independence between trajectories:
  - Distance between  $(s+n, s+n, \dots, s+n)$  and  $(m+c, m+c, \dots, m+c)$  same as between  $(s+c, s+c, \dots, s+c)$  and  $(m+n, m+n, \dots, m+n)$

# Building MD sequences with expanded alphabet

```
(s.toyMD <- seqMD(channels = channels))
```

```
## Sequence
```

```
## 1 n+s+u-n+s+w-n+s+w-n+m+u-c+m+u-c+m+u-c+m+w-c+m+w
```

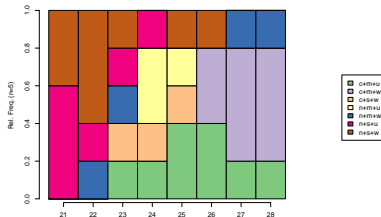
```
## 2 n+s+u-n+s+w-c+s+w-c+s+w-c+s+w-c+m+w-c+m+u-c+m+u
```

```
## 3 n+s+w-n+s+w-c+m+u-c+m+u-c+m+u-c+m+w-c+m+w-c+m+w
```

```
## 4 n+s+u-n+s+u-n+s+u-n+s+u-n+s+w-n+s+w-n+m+w-n+m+w
```

```
## 5 n+s+w-n+m+w-n+m+w-n+m+u-n+m+u-c+m+u-c+m+w-c+m+w
```

```
seqdplot(s.toyMD, with.legend = "right")
```



# Retrieving CAT costs

```
(CATcost <- seqMD(channels, what = "cost", sm = "INDELSLOG"))
```

```
##           c+m+u c+m+w c+s+w n+m+u n+m+w n+s+u n+s+w
## c+m+u 0.000 0.578 1.156 0.576 1.154 1.154 1.731
## c+m+w 0.578 0.000 0.578 1.154 0.576 1.731 1.154
## c+s+w 1.156 0.578 0.000 1.731 1.154 1.154 0.576
## n+m+u 0.576 1.154 1.731 0.000 0.578 0.578 1.156
## n+m+w 1.154 0.576 1.154 0.578 0.000 1.156 0.578
## n+s+u 1.154 1.731 1.154 0.578 1.156 0.000 0.578
## n+s+w 1.731 1.154 0.576 1.156 0.578 0.578 0.000
## attr("indel")
## [1] 0.882 0.782 0.882 0.849 0.749 0.949 0.849
## attr("alphabet")
## [1] "c+m+u" "c+m+w" "c+s+w" "n+m+u" "n+m+w" "n+s+u" "n+s+w"
## attr("cweight")
## [1] 1 1 1
```

# IDCD, CAT, and DAT distances

```
(dIDCD <- seqdist(s.toyMD, method = "OM", sm = "INDELSLOG"))
```

```
##      1      2      3      4      5
## 1 0.00 4.53 2.20 5.52 3.39
## 2 4.53 0.00 5.48 6.76 6.90
## 3 2.20 5.48 0.00 6.47 4.48
## 4 5.52 6.76 6.47 0.00 5.50
## 5 3.39 6.90 4.48 5.50 0.00
```

```
(dCAT <- seqdist(s.toyMD, method = "OM", sm = CATcost, indel = attr(CATcost,
"indel")))
```

```
##      1      2      3      4      5
## 1 0.00 5.06 2.31 6.35 2.31
## 2 5.06 0.00 5.20 6.44 7.37
## 3 2.31 5.20 0.00 7.50 3.26
## 4 6.35 6.44 7.50 0.00 7.23
## 5 2.31 7.37 3.26 7.23 0.00
```

```
(dDAT <- d.ch + d.mar + d.work)
```

```
##      1      2      3      4      5
## 1 0.00 3.84 2.31 4.99 2.31
## 2 3.84 0.00 3.77 5.97 5.67
## 3 2.31 3.77 0.00 6.82 2.78
## 4 4.99 5.97 6.82 0.00 6.15
## 5 2.31 5.67 2.78 6.15 0.00
```

# CombT: Deriving MD types from domain types

- Combining domain types  $\Rightarrow$  (too) many combined types, many scarce types
- Optimally merging combined types (so as to minimize quality loss of the partitioning)
  - Function `dissmergegroups()`

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# Rendering group of MD sequences

- Two ways of plotting MD sequences:
  - Plot of sequences of combined states
    - Large MD alphabet: Select contrasting colors for 10-12 most frequent tokens and display color legend for those tokens only.
  - For each group, separate plots by domains (groups by rows and domains by columns or conversely).  
TraMineR function `seqplotMD` since version 2.2-7
    - For plots depending on characteristics such as sequence order or representative sequences, characteristics must be determined at the MD level and applied to each domain level.



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# SHP biographic data (2000)

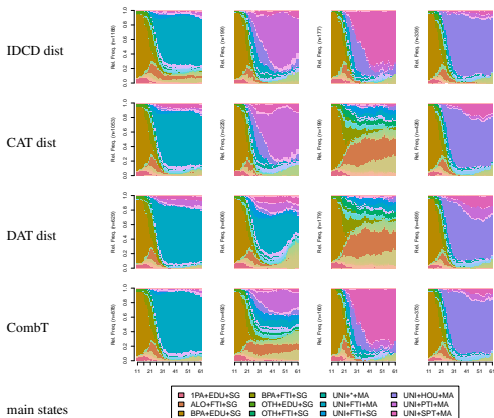
## Swiss Household Panel

- Three domains: living arrangement (LA, 9 tokens), work life (WL, 10 tokens), civil status (CS, 5 tokens)
- 1903 MD life sequences, ages 11 to 61
- 229 observed combined tokens
- Pairwise domain association

	Cramer $v$	$p(v)$	Pearson $r$	$p(r)$
LA with WL	0.29	0.0000	-0.002	0.0375
LA with CS	0.46	0.0000	0.509	0.0000
WL with CS	0.34	0.0000	-0.001	0.2577

# 4-cluster solution using different strategies

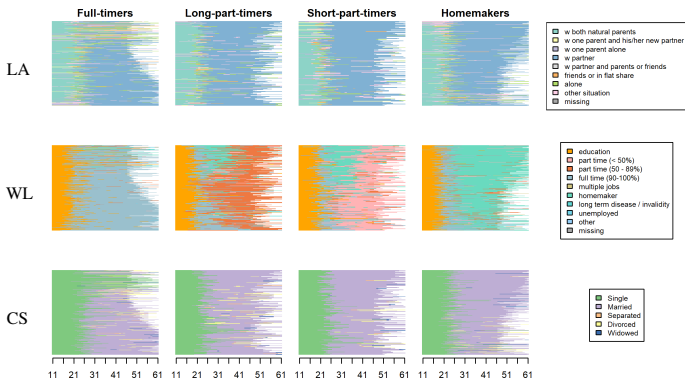
SHP. Three domains: living arrangement (LA), work life (WL), civil status (CS).  
1903 MD life sequences, ages 11 to 61, 229 combined tokens



(Figure from Ritschard, Liao, and Struffolino, 2023)

# IDCD 4-cluster solution by type (col) and domain (row)

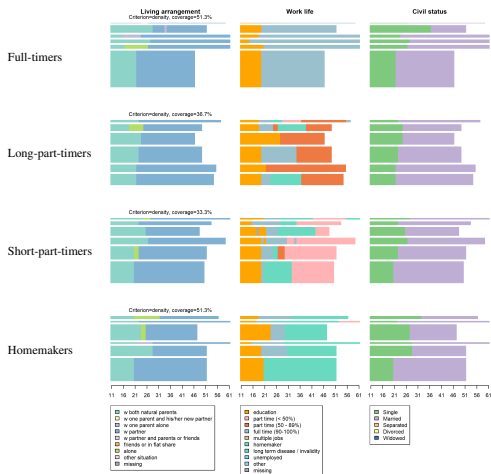
SHP. Three domains: living arrangement (LA), work life (WL), civil status (CS).  
1903 MD life sequences, ages 11 to 61, 229 combined states



(Figure from Ritschard, Liao, and Struffolino, 2023)

## IDCD 4-cluster solution: by type (row) and domain (col)

Representative MD sequences (Gabadinho and Ritschard, 2013)



(Figure from Ritschard, Liao, and Struffolino, 2023)

# Aspects under development

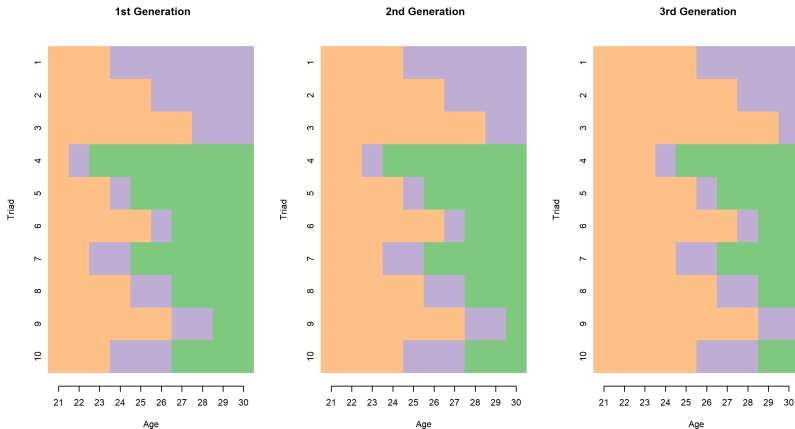
- Tools to help identifying combinations of domain characteristics unique to each MD types.
- How does association between domains depend on covariates?
- Visualization:
  - Automatize efficient color selection for multidomain state tokens.
    - Degree of darkness proportional to mean domain rank of combined tokens
    - Contrasting colors for a subset of main tokens (`seqmaintokens`)

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# Dyadic/Polyadic Data

Example from help page of TraMinerExtras::seqpolyads





# Polyadic analysis

(Liao, 2021)

- Polyadic data is special case of multichannel data where all channels have same alphabet.
- This allows to compute pairwise dissimilarities between channels for each polyad.
- for measuring how strongly sequences are linked within polyads.
- Liao (2021) proposed two statistics:
  - $U$ : difference between observed polyadic distance and mean distance of  $T$  randomized polyads
  - $V$ : proportion of the  $T$  randomized polyads with greater polyadic distance than observed polyad

# Polyadic Analysis

Example from help page of TraMinerExtras::seqpolyads

```
seqL <- list(seqGrandP, seqChild)
(seqG3.Tim <- seqpolyads(seqL, method = "HAM", a = 1, T = 100))

## $mean.dist
## Obs Rand
## 3.10 4.42
##
## $U
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 2.42 2.42 2.42 1.42 1.42 1.42 0.42 0.42 0.42 0.42
##
## $U.tp
##      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]
## [1,] 4.02e-18 4.02e-18 4.02e-18 1.02e-08 1.02e-08 1.02e-08 0.0673 0.0673
##      [,9]      [,10]
## [1,] 0.0673 0.0673
##
## $V
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 0.84 0.84 0.84 0.66 0.66 0.66 0.52 0.52 0.52 0.52
##
## $V.95
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 0 0 0 0 0 0 0 0 0 0
##
## $observed.dist
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
## [1,] 2 2 2 3 3 3 4 4 4 4
```

**Thank you!**  
**Questions?**

# References I

- Abbott, A. and A. Tsay (2000). Sequence analysis and optimal matching methods in sociology, Review and prospect. *Sociological Methods and Research* 29(1), 3–33. (With discussion, pp 34-76).
- Bernardi, L., J. Huinink, and R. A. Settersten, Jr (2019). The life course cube: A tool for studying lives. *Advances in Life Course Research* 41, 100258.
- Elder, Jr, G. H., M. K. Johnson, and R. Crosnoe (2003). The emergence and development of life course theory. In J. T. Mortimer and M. J. Shanahan (Eds.), *Handbook of the Life Course*, pp. 3–19. Boston, MA: Springer.
- Gabadinho, A. and G. Ritschard (2013). Searching for typical life trajectories applied to childbirth histories. In R. Levy and E. Widmer (Eds.), *Gendered life courses - Between individualization and standardization. A European approach applied to Switzerland*, pp. 287–312. Vienna: LIT-Verlag.
- Gabadinho, A., G. Ritschard, N. S. Müller, and M. Studer (2011). Analyzing and visualizing state sequences in R with TraMineR. *Journal of Statistical Software* 40(4), 1–37.

## References II

- Gauthier, J.-A., E. D. Widmer, P. Bucher, and C. Notredame (2010). Multichannel sequence analysis applied to social science data. *Sociological Methodology* 40(1), 1–38.
- Han, S.-K. and P. Moen (1999). Clocking out: Temporal patterning of retirement. *American Journal of Sociology* 105(1), 191–236.
- Liao, T. F. (2021). Using sequence analysis to quantify how strongly life courses are linked. *Sociological Science* 8(3), 48–72.
- Liao, T. F., D. Bolano, B. Cornwell, C. Brzinsky-Fay, A. Fasang, S. Helske, R. Piccarreta, M. Raab, G. Ritschard, E. Struffolino, and M. Studer (2022). Sequence analysis: Its past, present, and future. *Social Science Research* 107, 102772.
- Piccarreta, R. (2017). Joint sequence analysis: Association and clustering. *Sociological Methods & Research* 46(2), 252–287.
- Pollock, G. (2007). Holistic trajectories: A study of combined employment, housing and family careers by using multiple-sequence analysis. *Journal of the Royal Statistical Society A* 170(1), 167–183.

## References III

- Ritschard, G., T. F. Liao, and E. Struffolino (2023). Strategies for multidomain sequence analysis in social research. *Sociological Methodology* 53(2), 288–322.
- Robette, N., X. Bry, and E. Lelièvre (2015). A 'global interdependence' approach to multidimensional sequence analysis. *Sociological Methodology* 45(1), 1–44.