

Aoristic research in R

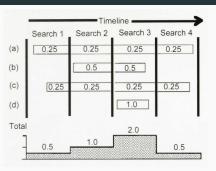
Correcting temporal categorizations in archaeology

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Motivation

What is aoristic?



Principle of aoristic analysis (Mischka 2004)

- Widespread and intuitive tool
- Production of time series from a collection of evidence with start and end date
 - in arch. situation (estimation of eg. site abundancy per time slice), these are mostly typological phases (based on typochronological artefacts)
 - in anthropological cases (estimation of eg. age distribuition), these are mostly individuals with anthropological age determination

Aoristics in arch. applications

- background for a lot of archaeological reasoning without proper reference to the method
- earliest explicit application in archaeology (to the best of my knowledge):
 - Mischka 2004
- recently reference eg. by:
 - Crema 2012
 - Kolář et al. 2016
 - Orton et al. 2017
 - Palmisano et al. 2017

Aoristic approaches in anthropology

 "proportional method": "It distributes the weight of single observations uniformly over the age-intervals into which they have been assigned" (Boldsen 1988, 335f.)

A1₁ and A2₁: the lower and upper margins of the age interval into which the i'th skeleton has been age determined.

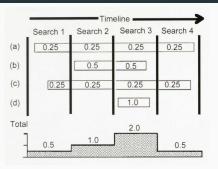
$$\alpha_{ij} = \begin{cases} 1 & \text{if } A1_i \leq j \leq A2_i \\ 0 & \text{else} \end{cases}$$

N: the number of skeletons in the analyzed sample. Using the proportional method:

$$F(t) = \frac{1}{N} \sum_{i=1}^{N} \sum_{j=0}^{t} \left[\alpha_{ij}/(A2_i - A1_i + 1)\right]$$

Boldsen 1988, 336

Ordinal scales and overlapping categories: the problem



Principle of aoristic analysis (Mischka 2004)

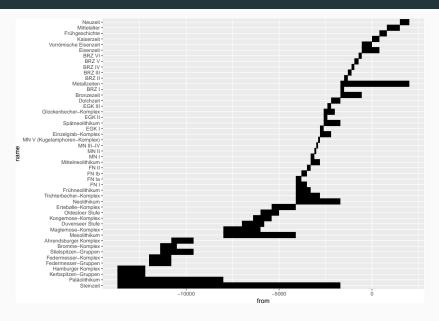
- ideal world: ordinal scales "involve discrete non- overlapping categories" (Byers 2017, 15)
- reality, due to fragmentary nature of data:
 - archaeological phasing is most often neither exclusive nor non-overlapping
 - anthropological age categories show different grades of resolution

Resulting problem:

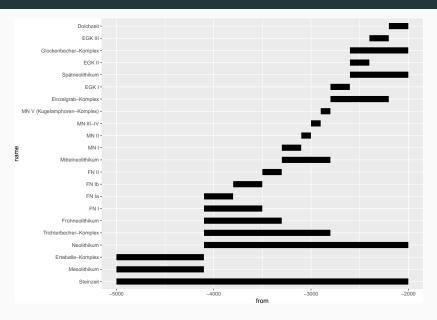
values where categories overlap tend to get overemphasized

Case study 1: Archaeological context

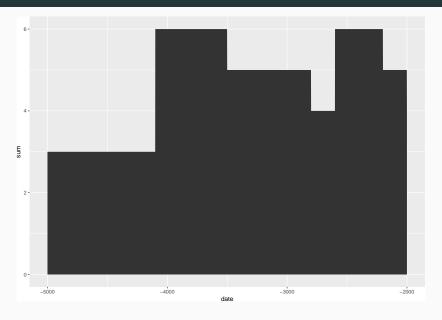
The archaeological situation



The archaeological situation, a bit more reduced



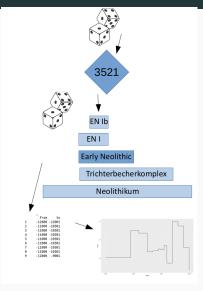
Number of overlapping periods



Simulating the typological dating process

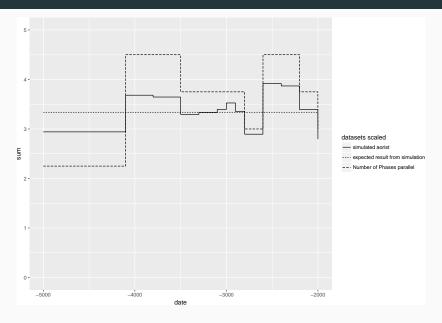
Simulation

- Equal distribution of data by randomly selecting a date within the time interval
- Preservation of objects is independent from their typochronological diagnostical value → random selection of possible archaeological dating
- Resulting dates analysed aoristically

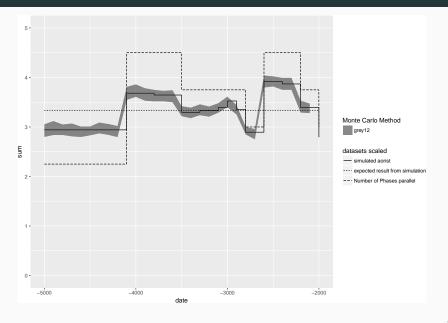


Monte Carlo simulation archaeological periods

Result of a naive aoristic analysis



Result of monte carlo method (Crema 2012 etc., using archSeries package)



Solution to the problem

	3499-3400	3399-3300	3299-3200
Phase 1	1	1	1
Phase 2	0	1	0
Phase 3	1	1	0

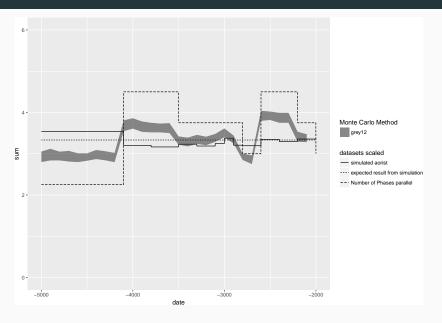
Solution

- Correct the aoristic weight by the number of parallel phases:
- Aoristic weight for each interval is 1/n parallel phases
- Normalize to row sum = 1

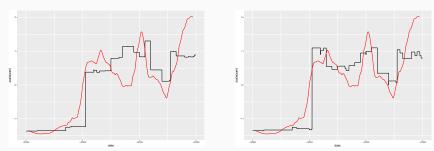
	3499-3400	3399-3300	3299-3200
Phase 1	0.5	0.3333333	1
Phase 2	0.0	0.3333333	0
Phase 3	0.5	0.3333333	0

	Phase 1	Phase 2	Phase 3
Site 1	0.2727273	0.1818182	0.5454545
Site 2	0.0000000	1.0000000	0.0000000
Site 3	0.6000000	0.4000000	0.0000000

Result of a corrected aoristic analysis



Real World: Neolithic at Jutland



Real World Example: Aoristic analysis of the Neolithic od Northern Jutland. Left uncorrected, right corrected. Black aoristic result, red sum calibration, smoothed by 100 year moving window.



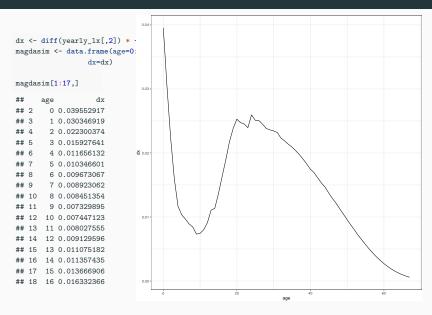
A simulated "real-world" anthropological case study

```
# Data available at
                                                    # http://data.princeton.edu/eco572/datasets/brassrlm1.dta
magdalenenberg
                                                    rlm1 <- read.dta("brassrlm1.dta")
                          SC
                                                   vfit1 <- predict(lf, newdata=rlm1)</pre>
                          ##
                                        1x
                                               1s
##
               Dx
          a
                                                   yearly_lx \leftarrow cbind(1:69, 1/(1 + exp(2 * yfit1[1:69])))
                                0.96585586 0.8499
        0-4 3.79
## 1
                                                   yearly_lx[1:17,]
                                0.92423423 0.7691
## 2
        5-9 4.62
                                0.88333333 0.7502
      10-14 4.54
## 3
                                                          [,1]
                                                                    [,2]
      15-19 4.21
                                0.84540541 0.7362
                                                    ## 1
                                                             1 0 9796006
                                0.71036036 0.7130
## 5
      20-24 14.99
                                                    ## 2
                                                             2 0.9400476
                                0.52468468 0.6826
                          ## 6
## 6
      25-29 20 61
                                                             3 0.9097007
                                                    ## 3
                                0.36972973 0.6525
## 7
      30-34 17.20
                                                    ## 4
                                                             4 0 8874003
                                0.24009009 0.6223
      35-39 14.39
## 8
                                                    ## 5
                                                             5 0.8714727
                               0.17990991 0.5898
## 9
      40-44 6.68
                                                    ## 6
                                                             6 0.8598166
                          ## 10 0.14351351 0.5534
## 10 45-49 4.04
                                                    ## 7
                                                             7 0.8494700
## 11 50-54 5.49
                          ## 11 0.09405405 0.5106
                                                    ## 8
                                                             8 0.8397969
                          ## 12 0.04252252 0.4590
## 12 55-59 5.72
                                                    ## 9
                                                             9 0.8308738
                          ## 13 0.02126126 0.3965
## 13 60-x 4.72
                                                    ## 10
                                                            10 0.8224225
                                                    ## 11
                                                            11 0.8150926
                                                    ## 12
                                                            12 0.8076455
                                                    ## 13
                                                            13 0.7996179
                                                            14 0.7904883
                                                    ## 14
                                                    ## 15
                                                            15 0.7794131
                                                    ## 16
                                                            16 0.7680557
                                                    ## 17
                                                            17 0 7543888
```

sc <- mutate(sc, yx = 0.5*log((1-lx)/lx), ys = 0.5*log((1-ls)/ls))

lf <- lm(yx ~ ys, data=sc)

simulating the Magdalenenberg Population



Real world Age Cohorts

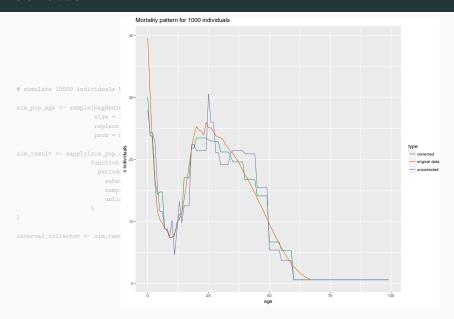
A typical "real-world" example of anthropological age categories (adapted from Moghaddam et al. 2016 on the cemetery of Münsingen-Rain)

```
##
      from to
## 1
         3 4
## 2
         7 8
         9 10
## 3
         9 11
## 4
## 5
        10 10
## 6
        12 13
## 7
        12 14
## 8
        13 14
## 9
        15 19
        18 27
## 10
## 11
        18 29
## 12
        20 25
## 13
        20 29
## 14
        20 49
## 15
        25 34
## 16
        25 39
## 17
        25 44
## 18
        30 39
## 19
        30 44
## 20
        30 49
## 21
        34 49
## 22
        35 44
## 23
        35 49
## 24
        40 49
## 25
        40 54
## 26
        40 59
## 27
        45 59
## 28
        50 99
## 29
        60 99
## 30
         0 0
## 31
         0 6
## 32
         1 2
## 33
         3
         1 3
## 34
```

The simulation

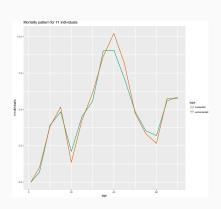
```
# simulate 10000 individuals based on their dx
sim_pop_age <- sample(magdasim$age,
                                                            # from the ages
                     size = 10000.
                                                            # take 10000 individuals
                     replace = T,
                                                            # multible times is ok
                     prob = magdasim$dx)
                                                            # with probability according to their dx
sim_result <- sapply(sim_pop_age,
                                                            # for all individuals
                    function(x){
                      periods_reduced %>%
                                                            # from the periodes
                        subset(from <= x & to >= x) %>% # take those within which the age falls
                        sample_n(size = 1, replace = T) %>% # take on of those randomly
                                                            # reformat
                        unlist.
interval_collector <- sim_result %% t() %>% as.data.frame() # reformat the result
head(interval_collector)
```

The simulation



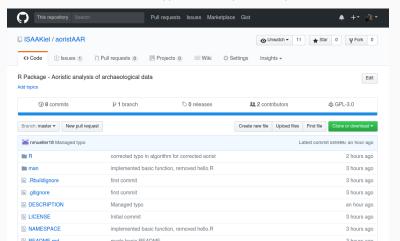
A real "real-world" example

the cemetery of Münsingen-Rain (processing with the R-package "mortAAR", data after Moghaddam et al. 2016)



Conclusion

- Aoristic analysis is still a valuable tool, might need a bit calibration
- Monte Carlo method ala Crema 2012 etc. is unfortunately not sufficient for removing bias of overlapping phases
- R package for doing (corrected!) aoristic analysis is available at the Github Repository of ISAAK: https://github.com/ISAAKiel/aoristAAR



Thank you



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The code of this presentation is available at:

 $https://github.com/MartinHinz/aoristAAR_pres$