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1 **Taner filter settings and automatic correlation optimisation for cyclostratigraphic**  
2 **studies**

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11

12 **Abstract**

13 Cyclostratigraphy and astronomical tuning utilize the imprint of quasi-cyclic insolation  
14 changes in geological records to establish chronologies. In this context, filtering of time  
15 series in specific frequency bands is commonly applied to extract information on  
16 astronomical forcing from geological datasets. This approach is performed on specific  
17 insolation components (precession, obliquity or eccentricity) and sometimes also their  
18 amplitudes either in depth or time domain. In this study, we design and apply a simulation  
19 technique to determine the optimal Taner filter settings to extract precession-, obliquity- and  
20 eccentricity-related interference signals from astronomically tuned geological datasets. This  
21 is done by testing a variety of filter settings on several astronomical and artificial datasets.  
22 Based on our results, we propose specific filter designs (cut-off frequencies and roll-off  
23 rates) for the best extraction of astronomical (interference) signals from tuned geological  
24 datasets. Focus here lies on datasets shorter than ca. 1 million years and interference  
25 patterns between astronomical components.

26 A second step utilizes these filter settings for an automated alignment, where geological data  
27 on a tuned time scale are matched to a suite of astrochronologic correlation targets. This is  
28 done by aligning filter minima and maxima to astronomical targets. This approach is  
29 particularly useful for the determination of the relative contributions of astronomical  
30 parameters in a specific dataset and allows for the automatic determination of phase shifts  
31 between well expressed insolation components in datasets.

32

33 **Key words**

34 cyclostratigraphy; automated tuning; automated time scale optimisation; Taner filter settings;  
35 correlation

36

37 **Competing interests statement:**

38 No author has a competing interest

39

40 **Author contributions\***

41  
42  
43  
44 \*CZ carried out the programming and designed the study. SK helped applying methods to  
45 several test datasets, discussed these, and ensured suitable implementation. FH and JL  
46 oversaw the study and discussed the implementation and documentation. All authors  
47 contributed to the manuscript creation by writing, discussing changes for clarity and technical  
48 usefulness and correctness.

49  
50 **1 Introduction**

51 Cyclostratigraphy relates quasi-cyclic patterns in sediments to astronomical characteristics  
52 which in turn are used for time scale reconstructions (e.g. Hinnov, 2000; Hinnov and Hilgen,  
53 2012; Hilgen et al., 2015). These are then further aided by other dating techniques (i.e. Ar/Ar  
54 and U/Pb dating). Generally, the quasi-cyclic patterns are extracted from geological datasets  
55 by filtering the data in the depth or time domain (e.g. Valero et al., 2014, 2016; Martinez and  
56 Dera, 2015; Da Silva et al., 2016). During the Neogene, direct astronomical tuning is often  
57 possible at the scale of precession (~ 20 kyr; e.g. Shackleton and Crowhurst, 1997; Lourens  
58 et al., 2001; Abels et al., 2009; Zeeden et al., 2014). In particular, precession filtering of  
59 geological data and the extraction of their respective amplitudes have been used to  
60 reconstruct the astronomical imprint within geological datasets (e.g. Ding et al., 2002;  
61 Lourens et al., 2010). These filtered data patterns and amplitudes are also applied to test  
62 tuned time scales (Shackleton et al., Meyers, 2015; 1995; Zeeden et al., 2015). Yet, filter  
63 settings are commonly chosen quite arbitrarily, and we are aware of only one study which  
64 systematically investigates the effect of filter settings (Li et al., 2018).

65 Hence, we focus on filtering individual precession- obliquity and eccentricity cycles in a most  
66 representative way. We highlight that the filter settings suggested in this study are unsuitable  
67 for extracting amplitude variations of cycles, as for such investigations the full span of  
68 astronomical forcing must be included (see e.g. Hinnov, 2000), and wider filters must be  
69 applied (Zeeden et al., 2015).

70 Here, we focus on two different aspects related to cyclostratigraphy and time scale  
71 reconstructions: (1) Taner filters and (2) automated alignment of filter extremes to correlation  
72 targets. We focus on Taner filters (Taner, 1992) as they (a) are available in the ‘astrochron’  
73 R package (a widely used method in cyclostratigraphy; Meyers, 2014; R Core Team, 2017)  
74 and in matlab (<http://mason.gmu.edu/~lhinnov/cyclostratigraphytools.html>), and allow for  
75 automated application to various datasets, (b) their filter properties such as high and low  
76 filter cut-off frequencies and the roll-off rate, a parameter for the steepness of the filter  
77 boundaries, can easily be adjusted, and (c) they enjoy increasing popularity in the  
78 cyclostratigraphic community, (e.g. Wu et al., 2013; Bouilila et al., 2014, 2015; Meyers, 2015;  
79 Laurin et al., 2016; Martinez et al., 2017). An intuitive visualisation of the Taner filter and its  
80 properties is given in Kodama (2015; their Fig. 4.5.). Comparing filters of geological data and  
81 astronomical targets, and especially their amplitudes, can give a direct (visual) impression of  
82 similarity.

83 Automated correlation is often regarded as a helpful tool and several methods have been  
84 proposed and published (Olea, 1994; Lisiecki and Lisiecki, 2002; Pälike, 2002; Huybers and  
85 Aharonson, 2010; Lin et al., 2014; Kotov et al., 2016; Edwards et al., 2018), and used (e.g.  
86 Lisiecki and Raymo, 2005; Pälike et al., 2006b; Necula and Panaiotu, 2008; Lang and Wolff,

87 2011; Liebrand et al., 2011). Generally, these methods aim at a high-resolution correlation  
88 based on an initial time scale. However, they do not enjoy great popularity within the  
89 geosciences community. This may be because no-easy-to-use open source application was  
90 made available for cyclostratigraphic and geochronologic applications until recently (Kotov et  
91 al., 2016), while also care must be taken to take additional constraints from integrated  
92 stratigraphy into account (Hilgen et al., 2014) to avoid errors in automated correlation. The  
93 fear of geoscientists to be replaced by algorithms may also contribute to limit the application  
94 of such methods.

95 We suggest that automatic correlation approaches become especially useful when their  
96 freedom in changing sedimentation rates is limited, and an initial (tuned) time scale is the  
97 basis for further improvements. Here we describe and provide algorithms that (a) create an  
98 ensemble of tuning targets based on variable contributions of the individual astronomical  
99 parameters as well as variable phase relationships between them (Laskar et al., 2004), and  
100 (b) automatically optimise a correlation between a geological record and a specific target or  
101 an ensemble of tuning targets by aligning filtered data extremes (i.e., minima and maxima) to  
102 those in the targets.

103

## 104 **2 Experimental design and computer code**

### 105 **2.1 Taner filter settings for cyclostratigraphy: concept**

106 To find the best Taner filter settings for astrochronologic applications, we test the frequency  
107 range and steepness of various Taner filter settings. This is done by using the precession,  
108 obliquity and eccentricity from the La2004 astronomical solution (Laskar et al., 2004). In  
109 addition, a set of signal- and time-distortions are imposed on the solution, and the extraction  
110 of astronomical signals is tested on these datasets (see Table 1 for an overview). The details  
111 of these tests are outlined below, and the used R code is available in the Supplementary  
112 Materials.

113

### 114 **2.2 Taner filter settings for cyclostratigraphy: implementation**

115 For this study we utilized the R software (R Core Team, 2017) environment and the Taner  
116 (Taner, 1992) filter incorporated within the 'astrochron' R package (Meyers, 2014), because  
117 of the free availability and user flexibility of astrochron, and its increasing application in the  
118 astrochronology community. The La2004 astronomical solution (Laskar et al. 2004) is used  
119 at a 1 kyr resolution as basis for our experiments with artificial datasets. For these datasets,  
120 we used the last one million years of the solution and a mix of precession and obliquity/tilt  
121 signals mimicking northern and southern hemisphere climate variability. In this context  $p-0.5t$   
122 represents standardized precession minus  $0.5 \times$  standardized tilt (Lourens et al., 1996), and  
123  $0.5p-t$  and  $p-t$  represent similar signals with a higher contribution of obliquity. The  $-(p-0.5t)$  is  
124 used for testing precession and obliquity filter settings. For reconstructing eccentricity, a mix  
125 of  $p+0.5t+e$  was used. The use of a one million year-long test dataset does not imply that  
126 results are only valid for time scales of this length. We choose this length as a case study  
127 because (a) it provides enough cycles to give statistically useful results for precession,  
128 obliquity and the  $\sim 100$  kyr eccentricity components, (b) longer datasets would use more  
129 computing time, and (c) over such a length the frequency of precession and obliquity can be  
130 considered unchanged.

131 The experimental setup is designed to test filter settings where a clear expression of an  
132 astronomical imprint is present in data, and also in more complicated cases which may be  
133 regarded suitable for mimicking real data. For testing precession filter settings, initially we  
134 test the possibility to extract precession from a  $-p+0.5t$  (experiment #1) and a  $p-0.5t$   
135 (experiment #2) dataset using the last 1 million years. After performing these experiments for  
136 idealized astronomical signals, we included 50, 100 and 200% AR1-noise to the  $-p+0.5t$   
137 dataset (experiments #3-5) and to the  $p-t$  dataset (experiments #6-8), which has a lower  
138 precession contribution and represents a less ideal case for extracting precession  
139 information. To test filter settings in cases where age models are imperfect, we include noise  
140 in artificial datasets (as in experiments #3-5). We also include uncertainty in the artificial  
141 stratigraphic (time) domain (experiments #9-11) by replacing age with a gamma distribution  
142 with shape and scale parameters of five and one respectively. Gamma distributions are  
143 limited to positive values and can therefore model a growing stratigraphic column and have  
144 been previously used for this purpose (Martinez et al., 2016). Finally, experiments #12-14  
145 are similar to #9-11 but are clipped at zero to ensure only positive values. This represents a  
146 specific case of nonlinear response to astronomical forcing. Table 1 summarizes the  
147 experiments for precession, a supplementary R script performs these tests. For the  
148 optimisation experiments the procedure may be described as follows:

- 149 (1) Maximize  $\text{abs}(\text{correlation between (original signal, filtered signal)})$ ,
- 150 (2) Subject to {lower filter boundary, upper filter boundary, filter roll-off rate}.

151 Testing filter settings for obliquity and eccentricity is done in a similar way, with a given basis  
152 for the eccentricity filter reconstruction of  $p+0.5t+e$ . The full code of all the conducted  
153 individual experiments is attached as supplementary R scripts. Results for experiments with  
154 obliquity and eccentricity are available as Supplementary Tables 1 and 2.

155

### 156 **2.3 Automatic generation of correlation target ensembles for cyclostratigraphic** 157 **studies**

158 Although a single correlation reference dataset is often applied and may represent a suitable  
159 target, it might not be appropriate as the exact mix of astronomical parameters in a  
160 geological dataset is usually unknown, and also may vary with time. Therefore, we suggest  
161 testing a geological dataset against an ensemble of astronomical correlation targets with  
162 changing contributions of two of the three astronomical parameters precession (P),  
163 obliquity/tilt (T) and eccentricity (E). To facilitate this, three R functions are implemented,  
164 which generate correlation targets of (a) eccentricity and tilt (b) precession and tilt and (c)  
165 eccentricity and precession. The amount of different combinations, and the maximal phase  
166 lag of the longer period component can be set.

167 In this study, we abstained from including an experiment where all three astronomical  
168 parameters are mixed since: (a) often only one or two of the astronomical parameters are  
169 dominating geological datasets and hence are used as tuning targets (Kuiper et al., 2008;  
170 Liebrand et al., 2011; Zeeden et al., 2013; Boulila et al., 2015; e.g. Kaboth et al., 2016), (b)  
171 the relative phase information can be more directly retrieved when limiting the investigation  
172 to only two parameters, and (c) often eccentricity is present in geological records as  
173 amplitude modulation of the precession signal. Where a mix of precession, obliquity and  
174 eccentricity may be useful as correlation target, a correlation target can be easily generated  
175 using the established 'etp' function of the 'astrochron' R package (Meyers, 2014; R Core  
176 Team, 2017) on the basis of the La2004 solution. Also the amplitude modulation of obliquity  
177 is not used here, as it has no direct influence on insolation.

178

## 179 **2.4 Generation of correlation targets: implementation**

180 The functions 'createETtargets', 'createPTtargets' and 'createEPTargets.R' automatically  
181 create a specific amount (**N**; parameters to be set in the R functions in **bold**) targets with  
182 changing contributions of two astronomical parameters from a starting time (**TimeMin**) to a  
183 maximum age (**TimeMax**) in kyr before present with a temporal spacing of **Dt** kyr. The  
184 influence of hemispheres can be set for the 'createPTtargetes.R' and 'createPETargets.R'  
185 functions, and maximal phase shifts of the obliquity relative to precession and eccentricity  
186 can be set using the **phaseShift** parameter, which represents the maximal phase shift in kyr.  
187 For the function 'createPTtargets.R' also a nonlinear reaction to precession and eccentricity  
188 may be tested by including setting the **nc** (nonlinear combinations) parameter to >1. This is  
189 done by applying the exponential function  $\exp(\text{precession} * 5 * \text{nc})$  where **nc** represents the  
190 number of nonlinear combinations of amplified precession and obliquity. The multiplication  
191 by 5 is somewhat arbitrary chosen; it gives a useful difference of output reference datasets.  
192 This nonlinear amplification of precession may be useful in cases where precession is  
193 nonlinearly amplified in geological records during eccentricity maxima.

194

## 195 **2.5 Adjustment of correlative age models**

196 Several methods of automatic alignment between signals have been suggested which are all  
197 based on the principle of fully automatic alignment of datasets in high resolution (e.g. Lisiecki  
198 and Lisiecki, 2002; Lin et al., 2014; Kotov et al., 2016). We propose a different  
199 implementation in this study and assign maxima and minima of filtered geological data to  
200 corresponding maxima and minima in correlation targets. This limits the temporal flexibility of  
201 the method presented here to less than an investigated (precession, obliquity, eccentricity)  
202 cycle prominently present in data and used for age model construction.

203 The presented method requires high-resolution geological proxy data that record  
204 astronomical-scale changes with high fidelity. A data resolution of 1-3 kyr is suggested here  
205 for precession related variations allowing for the identification of precession/obliquity  
206 interference patterns. This data resolution and a tuned time scale are required to lead to  
207 useful results. Where a delayed response of proxy data may be expected, as e.g. in proxies  
208 for ice volume and soil development in the Quaternary (e.g. Thompson and Goldstein, 2006;  
209 Marković et al., 2015), care must be taken to initially obtain and tune to a useful (precession)  
210 phase. This may not be possible in all cases, as adjusting the phase of both precession and  
211 obliquity is not anticipated directly with this method.

212 The here proposed method uses a geological dataset, one or a set of correlation targets,  
213 and the temporal resolution as input parameters. It filters the data and target, aligns maxima  
214 (and minima) of these filter results, and selects the target resulting in a best fit. Linear  
215 interpolation of the age/depth relationship is applied between tie points.

216 We test the presented approach by applying the automatic correlation optimisation to two  
217 Miocene marine datasets from the Mediterranean and equatorial Atlantic. Both dataset show  
218 precession-obliquity interference patterns and both records were used for establishing  
219 integrated time scales (Hüsing et al., 2009; Zeeden et al., 2013, 2014; Wotzlaw et al., 2014).  
220 The first dataset is from Monte dei Corvi, Italy. This section was investigated for its  
221 integrated stratigraphy using astrochronology, magnetostratigraphy and also radiometric  
222 dating (Hilgen et al., 2003; Hüsing et al., 2007, 2009, 2010; Wotzlaw et al., 2014; Zeeden et  
223 al., 2014). A dataset of lightness as obtained by calorimetric measurement directly on

224 outcropping sediments has previously been used to estimate the phase between precession  
225 and obliquity (Zeeden et al., 2014). Here, we test the existing tuning results through the  
226 application of our correlation algorithm. The second example is a colour reflectance record  
227 from ODP Leg 154, Site 926 in the equatorial Atlantic (Curry et al., 1995; Shackleton and  
228 Crowhurst, 1997; Zeeden et al., 2013). The colour reflectance variability has been  
229 interpreted to represent input of terrestrial material from the Amazon into the equatorial  
230 Atlantic (Harris et al., 1997; Dobson et al., 2001), and ODP Leg 154 shows a rather clear  
231 imprint of astronomical climate forcing, which was used in numerous studies (Shackleton  
232 and Crowhurst, 1997; Shackleton et al., 1999; Pälike et al., 2006a; Wilkens et al., 2017).  
233 Here we analysed the time interval from 8.6 to 9 million years ago of this dataset using the  
234 **corOptMinMax** function.

235

## 236 **2.6 Implementing the adjustment of correlative age models**

237 The function performing the correlation optimisation 'corOptMinMax.R' can be used to test if  
238 the alignment improves or deteriorates the correlation. By setting the parameter  
239 **deselectTiePoints** to "T" (vs. "F"), tie points decreasing correlation can be deselected  
240 automatically until a minimum of 2 tie points is left. The **resolution** parameter can be set to  
241 focus on precession ("p"), obliquity ("o"), eccentricity ("e") or combinations ("po" for a  
242 combination of precession and obliquity and "eo" for a combination of eccentricity and  
243 obliquity). The function **corOptMinMax** uses a geological dataset (**Data**) and a target  
244 (**target**) as basic input, as well as the temporal resolution as **Dt** parameter. The maximum  
245 temporal difference between the filtered data maxima and minima to the correlation target  
246 can be set by the **maxDiff** parameter; this parameter limits the flexibility of the algorithm.  
247 The **corOptMinMax** function returns resulting correlation coefficients (here the spearman  
248 rank correlation to allow for nonlinearity in the relationship between data and targets) of the  
249 original time scale, the correlations resulting from adjustments and resulting tie points. The  
250 **corOptMinMax** function produces a list containing (a) the original correlation between  
251 optimal target and data (b) a variable including the optimal correlation, its contribution of  
252 obliquity and precession, the tuning target contributions of precession and obliquity and the  
253 lag of a second astronomical component if applicable (c) the re-tuned data and (d) the used  
254 tie points for re/tuning data.

255

## 256 **3 Results**

### 257 **3.1 Filter settings and generation of correlation targets**

258 Table 1 summarizes the results from the experiments investigating different filter settings for  
259 precession and Supplementary Tables 1 and 2 contain the results for obliquity and  
260 eccentricity. Please note that these results are partly based on resampling procedures and  
261 noise generation from specified distributions. To make results reproducible we set a seed in  
262 the R code. For all experiments, we investigate whether the ideal filter properties represent  
263 real optima in the settings, or if these lie in a range of values giving similar results. Especially  
264 the dependency of correlation results on the cut-off frequencies is for several experiments  
265 rather arbitrary (especially for precession and obliquity filters), as exemplary shown in  
266 Supplementary Fig. 1. In such cases, we estimate the narrow end of filter settings resulting  
267 in the best correlation; these estimates based on the inspection of relationships between  
268 parameters and correlation are also displayed in Table 1 and Supplementary Tables 1 and  
269 2.

270 Experiments show that for precession, filters are optimally set at frequencies from 0.043 to  
271 0.054 [1/kyr], using a roll-off rate of  $10^{28}$ . These consistently perform well in a set of  
272 experiments with artificial data (Table 1). This includes testing its robustness in non-ideal  
273 circumstances as: (a) changing amount of noise, (b) different contributions of precession and  
274 obliquity, and (c) non-constant sedimentation rate. For obliquity, we suggest setting upper  
275 and lower filter limits at 0.022 and 0.029 [1/kyr] and the roll-off rate at  $10^3$ . Eccentricity of the  
276 last 1 million years is best reconstructed by setting filter boundaries to 0.003 and 0.012  
277 applying a roll-off rate of  $10^3$ .

278 The corresponding R scripts (see section 2.4 and Supplementary Materials) automatically  
279 generate ensembles of correlation targets and are tested for a best fit with geological data  
280 hereafter.

281

### 282 **3.2 Adjustment of correlative age models**

283 The optimisation of correlation between two geological datasets and a mix of precession and  
284 obliquity is applied to two marine example datasets (see sections 2.5, 2.6). Figures 2-5  
285 represent the automatically generated output from the application of the 'corOptMinMax.R'  
286 function with some visual adjustments. For both datasets, we apply filter properties including  
287 precession and obliquity by setting the **resolution** parameter to "po". In this way we obtain  
288 information on the optimal mix of precession and obliquity, the obliquity phase, and improve  
289 the correlation between geological data and astronomical target.

290 The grey scale colour dataset from Monte dei Corvi for the interval from 9.6 to 9.0 Ma (Fig.  
291 2) is best resembled by a combination of 0.48\*precession and 0.52\* obliquity in this time  
292 interval, and thus a clearly higher obliquity contribution than previously used (inverted p-0.5t)  
293 for correlation to astronomic targets when considering longer intervals (Hilgen et al., 2003;  
294 Hüsing et al., 2010; Zeeden et al., 2014). This shows the varying contribution of precession  
295 and obliquity in long term eccentricity maxima and minima, where the p-0.5t gives a good  
296 overall fit of the Monte dei Corvi Record. An obliquity offset of 1 kyr results in the best  
297 correlation by increasing the correlation coefficient slightly from 0.64 to 0.66. Next, we  
298 consider two time intervals with clearly different precession and obliquity contributions (Figs.  
299 3-4), from 9.6-9.3 and 9.3-9.0 Ma. The interval from 9.3-9.0 Ma (Fig. 3) is dominated by  
300 precession (74%) where obliquity (36%) plays a subordinate role. The correlation increases  
301 slightly from 0.71 to 0.74 with changing obliquity phase, but this phase might not be very  
302 reliable as the obliquity contribution is relatively weak. The interval from 9.6-9.3 Ma (Fig. 4) is  
303 dominated by obliquity (66%) while precession plays a lower role (34%). Obliquity is leading  
304 the signal by 2 kyr, and again the correlation can be improved a bit, here from 0.66 to 0.70.

305 Using filter properties including precession and obliquity on the dataset from Ceara Rise  
306 suggests that a mix of 44% precession and 56% obliquity provide the best correlation target  
307 (Fig. 5). An obliquity offset of 1 kyr results in a correlation coefficient of 0.66 (0.57 based on  
308 the original chronology). Results support the original tuning in this case, but propose a  
309 higher contribution of obliquity. It needs to be noted that this is a statement for this specific  
310 time interval only, and the overall good fit between the Ceara Rise record and the originally  
311 used target (inverted p-0.5t) is not questioned here.

312

## 313 **4 Discussion**

### 314 **4.1 Filter settings**



315 Here we suggest specific filter settings for the optimal reconstruction of astronomical  
316 parameters from geological datasets. We propose to use cut-off frequencies of 0.043 and  
317 0.054 [1/kyr] for filtering precession related signals and 0.022 and 0.029 [1/kyr] for obliquity.  
318 For eccentricity, the filter results (0.003 and 0.012 [1/kyr]) describe the periodicities around  
319 100 kyr well. In case of lower and very low frequency eccentricity components (e.g. Boulila  
320 et al., 2012; Martinez and Dera, 2015) wider band-pass or low-pass filters will be more  
321 appropriate than the ones proposed in this study. While rather low roll-off rates perform best  
322 for eccentricity and obliquity, precession is best reconstructed when using high roll-off rates  
323 are used. This may be explained by most of precession components being in the filter range,  
324 while some low-frequency components of obliquity and eccentricity will only partly be  
325 captured by lower roll-off rates or alternatively wider filters, which then will also incorporate  
326 more non-astronomical data variability.

327 The filter properties designed in this study are adapted to recent precession and obliquity  
328 frequencies and are therefore particularly well suited for the Quaternary and Neogene time  
329 periods. However, because obliquity and precession periods were shorter in the past (Berger  
330 et al., 1992; Hinnov, 2000; Laskar et al., 2004), different filter settings will be optimal for pre-  
331 Neogene time intervals. Corresponding filter properties for older times are listed in Table 2.  
332 To determine the best filter settings for time intervals, the supplementary scripts can be  
333 adjusted to test pre-Neogene time intervals still covered by the La2004 solution by changing  
334 the timing (code lines 18-22). This script then automatically generates the best filter settings  
335 for an individual time interval, based on the precession and obliquity frequencies from Laskar  
336 et al. (2004). Table 2 provides an estimation of corresponding frequencies according to the  
337 evolution of frequencies by Laskar et al. (2004).

#### 338 **4.2 Using large ensembles of correlation targets**

339 The method described here allows a direct correlation between proxy data and a large  
340 ensemble of correlation targets of which the best fit can automatically be selected.  
341 Additionally, it can be directly combined with further data analysis in R, as e.g. provided by  
342 the 'astrochron' package (Meyers, 2014; R Core Team, 2017). It requires an initial time  
343 scale, and the maximum deviation of tie points from this initial time scale can be set, hence  
344 deliberately reducing the change in timing and possibility for misinterpretations. It may be  
345 argued that setting maximal deviation from original tuning is prescribing results. We agree  
346 that this may represent an issue. However, we see the possibility to avoid automatically  
347 implemented misinterpretations and therefore provided the option to limit the time offset.  
348 Additionally, the functions provided allow for the testing of various phase relationships  
349 between astronomical parameters automatically. Beside the automated tuning-optimisation  
350 our proposed application also includes the possibilities of optimal parametrization of ice  
351 volume models. In addition, the ensembles of targets and corresponding tunings allow  
352 stochastic approaches to investigate the effect of interpretations based on age models.

#### 353 **4.3 Correlation optimisation and exemplary application**

354 The results from our proposed automated alignment approach are less arbitrary than manual  
355 tuning and produce higher correlation coefficients in the examples shown, although the aim  
356 here is to reconstruct the relative contributions of precession and obliquity. The automatic  
357 alignment of minima and maxima of filtered data requires an initial tuning (on the time scale  
358 to be achieved by automatic alignment, e.g. precession, obliquity or eccentricity), a clear  
359 expression of the astronomical parameters used for alignment and a constant phase  
360 relationship of proxy data to astronomical forcing. This method requires more guidance than  
361 other automatic alignment methods based on dynamic time warping and dynamical  
362 programming (e.g. Lisiecki and Lisiecki, 2002; Lin et al., 2014; Kotov et al., 2016), and is not

363 meant to create a fully unguided correlation. The presented approach is developed to  
364 optimise a tuning, and mainly find the best fitting correlation target in large ensembles of 10s  
365 to 100s of options, the relative contributions of astronomical parameters and the phase of  
366 obliquity in precession and obliquity driven records. To our knowledge the methods  
367 described and the appended R scripts are the only open available and free-to-use possibility  
368 to test a dataset against large ensembles of correlation targets.

369 The Italian based Miocene example gives convincing results and correlation is similar to  
370 previously published results (Zeeden et al., 2014). In addition, we obtain information on the  
371 relative contributions of precession and obliquity, and the obliquity phase. The obliquity  
372 phase offset of 1 kyr is consistent with cross-spectral analysis of a tuning to the p-0.5\*t  
373 reference, which resulted in an obliquity phase offset of up to 1.5 kyr (Zeeden et al., 2014).

374 The dataset from Miocene drill cores recovered during ODP Leg 154 in the equatorial  
375 Atlantic shows a correlation coefficient of 0.61 when optimising the correlation. Improved  
376 results are in this case achieved when filter properties include precession and obliquity  
377 (versus filter properties only comprising precession, not shown), therefore we regard these  
378 results most useful.

379 In both experiments with real geological data, better results are obtained when filter settings  
380 include both obliquity and precession. This is probably because the datasets used are  
381 influenced by Northern Hemisphere insolation (Shackleton and Crowhurst, 1997; Hilgen et  
382 al., 2003) which comprises both precession and obliquity. Importantly, we show that the  
383 tuning target previously used are not necessary optimal for all time intervals, and that fitting  
384 short intervals to more suitable targets leads to a better correlation and an in detail better  
385 and more reproducible age model. However, it is important to highlight that here we focus on  
386 short intervals of much longer datasets. We here do not question the originally used  
387 correlation targets, which indeed give a good overall fit with the datasets.

388 The relative contribution of obliquity and precession are not constant over time, but depend  
389 on both astronomical forcing with changing parameter amplitudes (Laskar et al., 2004) and  
390 climate feedback (e.g. Rutherford and D'Hondt, 2000; Lisiecki and Raymo, 2005; Meyers  
391 and Hinnov, 2010; Zeeden et al., 2013). This limits the applicability of our approach to long  
392 time series, because better fits of data and targets can be expected for shorter intervals with  
393 similar forcing and feedback. For example the Monte dei Corvi record (Hilgen et al., 2003;  
394 Hüsing et al., 2009), part of which is used here as example, shows a good fit with the p-0.5t  
395 astronomical target when the overall record is considered. However, shorter intervals  
396 especially in long term eccentricity minima show a higher obliquity influence, as may be  
397 expected.

398 The information on the relative influence of astronomical parameters and their phase, which  
399 is obtained here by fitting numerous targets with changing influence of parameters, can be  
400 regarded the main aim and outcome, whereas a small increase in correlation does not  
401 provide much additional information.

402

## 403 **5 Conclusions**

404 Here we propose specific Taner filter settings to extract astronomical scale variations from  
405 geological tuned time series. Our experiments suggest the following filter properties for  
406 Quaternary and Neogene studies: For precession, filter boundaries are optimally set at  
407 frequencies of 0.043 and 0.054 using a roll-off rate of  $10^{28}$ . These consistently perform well  
408 in a set of experiments with artificial data. For obliquity, we suggest setting upper and lower

409 filter limits at 0.022 and 0.029 and the roll-off rate at  $10^3$ . The components around 100 kyr of  
410 eccentricity are best reconstructed by setting filter boundaries to 0.003 and 0.012 and by  
411 applying a roll-off rate of  $10^3$ . At least for precession, lower roll-off rates ( $\sim 10^2$ ) perform better  
412 in practice because even tuned age models are usually imperfect and therefore require  
413 investigating a wider frequency band. For deep-time records, different cut-off frequencies are  
414 necessary due to the evolution of precession and obliquity frequency through Earth history  
415 (see Tab. 2).

416 These investigated filter settings are in turn applied in an automatic alignment of filtered  
417 geologic data, which we provide as a correlation optimisation method for testing geological  
418 datasets against sets of correlation targets. Two geological datasets of Miocene marine  
419 successions are directly compared with large ensembles of targets (425 reference datasets,  
420 respectively). This guided alignment is preventing high deviations from an original age  
421 models, and therefore may be preferred over fully automatic alignments with less guidance.  
422 This method leads to a better understanding of the contribution of different astronomical  
423 parameters to a geological dataset, and can estimate phase of obliquity at the same time.  
424 We provide all code as R functions and scripts for further use in the supplementary  
425 materials.

426

#### 427 **Acknowledgements**

428 CZ is financed through an PSL post-doctoral fellowship. All datasets used in this study are  
429 available in the Pangaea database, computer code is available as supplementary  
430 information. Four reviewers are thanked for their helpful comments which helped to improve  
431 this study.

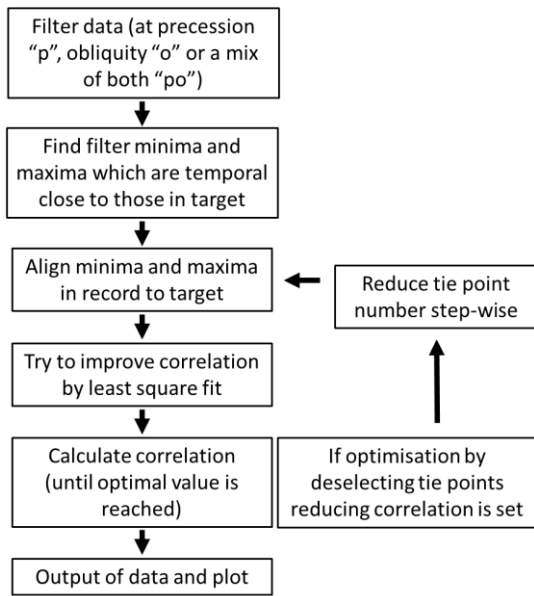
432

#### 433 **Computer Code Availability**

434 All computer code described here is available as code in the R language in supplementary  
435 materials. It was designed by C. Zeeden, IMCCE, Observatoire de Paris, PSL Research  
436 University, CNRS, Sorbonne Universités, UPMC Univ. Paris 06, Univ. Lille, 75014 Paris,  
437 France. Phone: +33 1 4051 2038. The code requires a decent computer, and the R  
438 software including the 'astrochron' package installed.

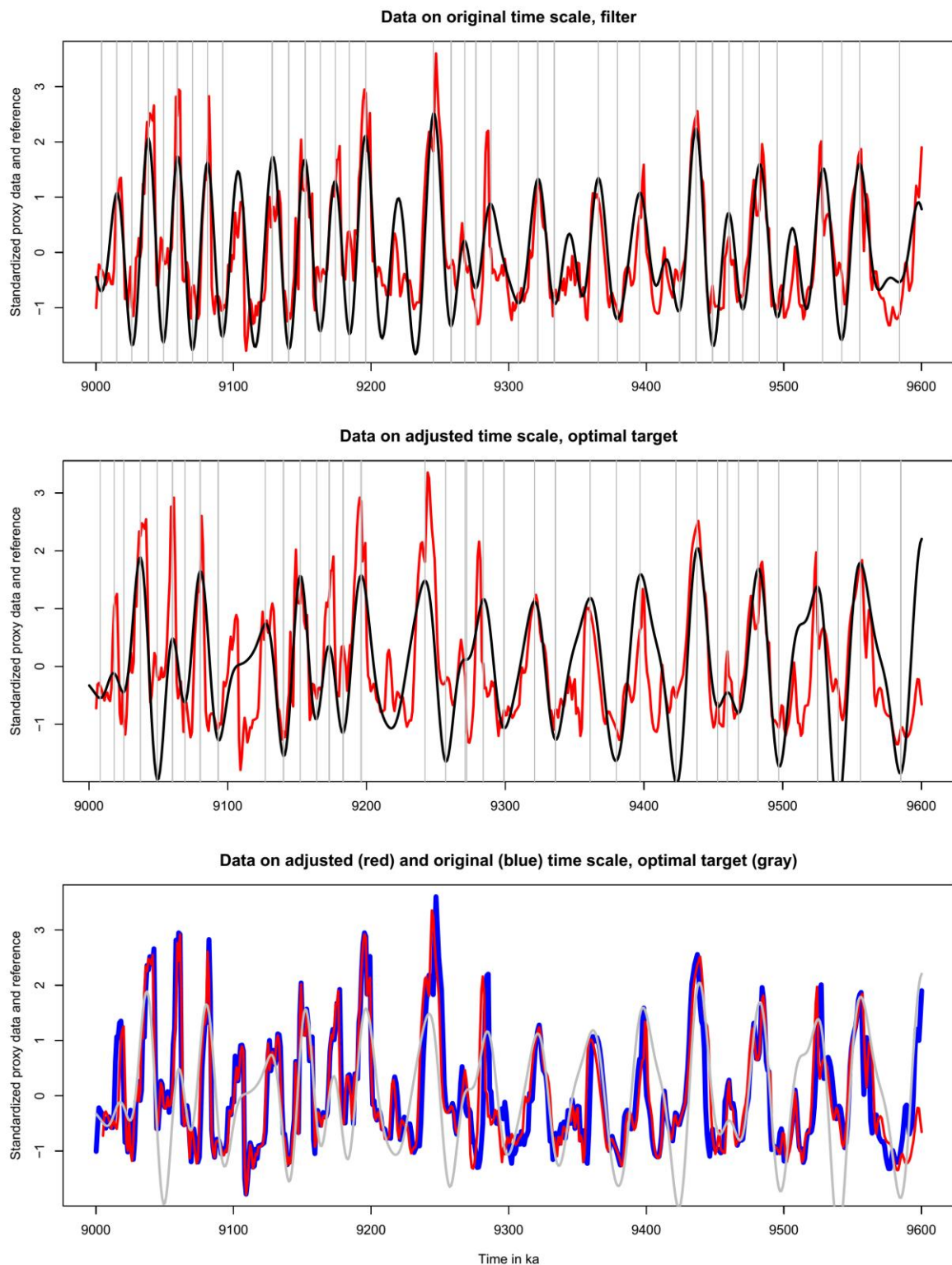
439

440 **Figures and Tables**

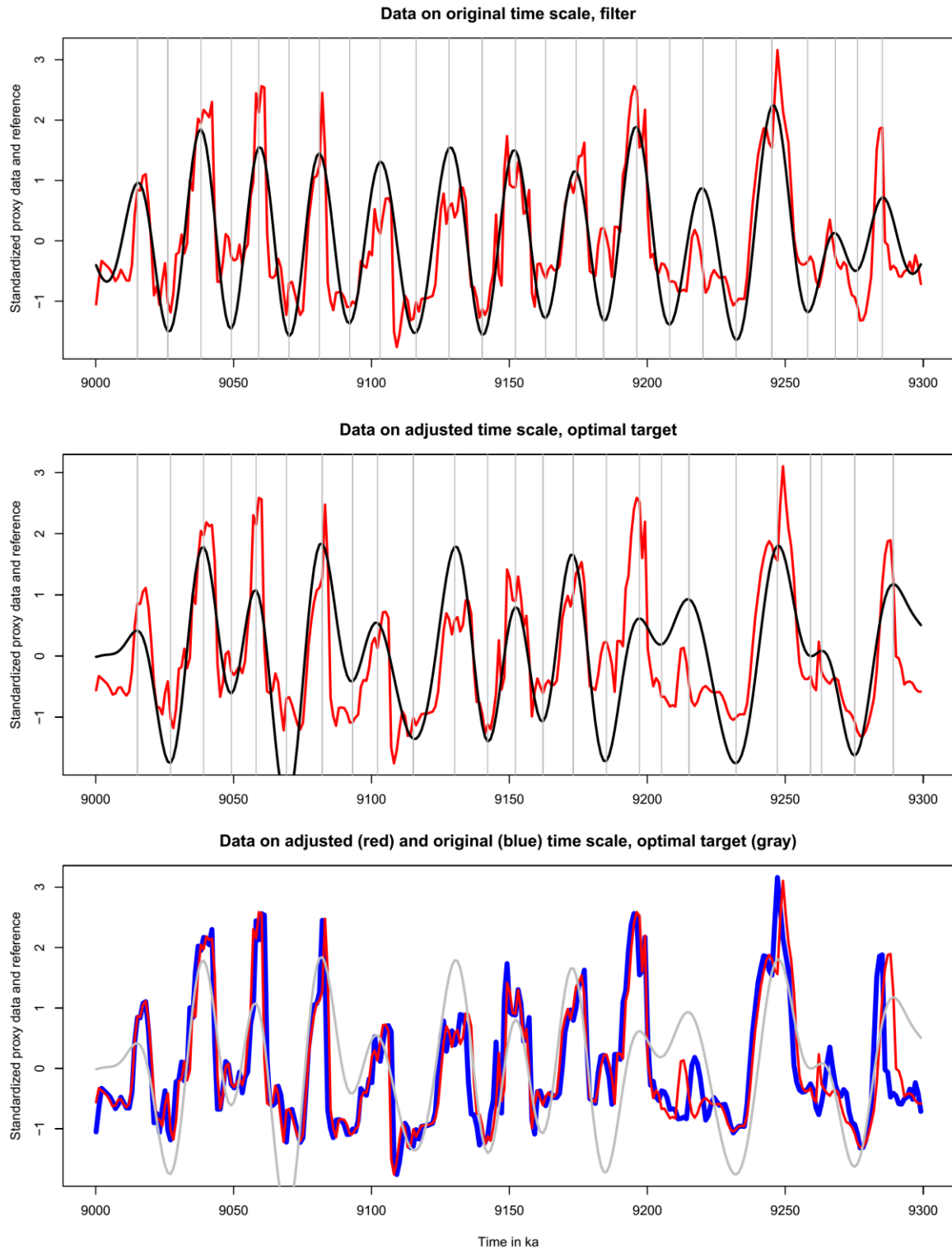


441

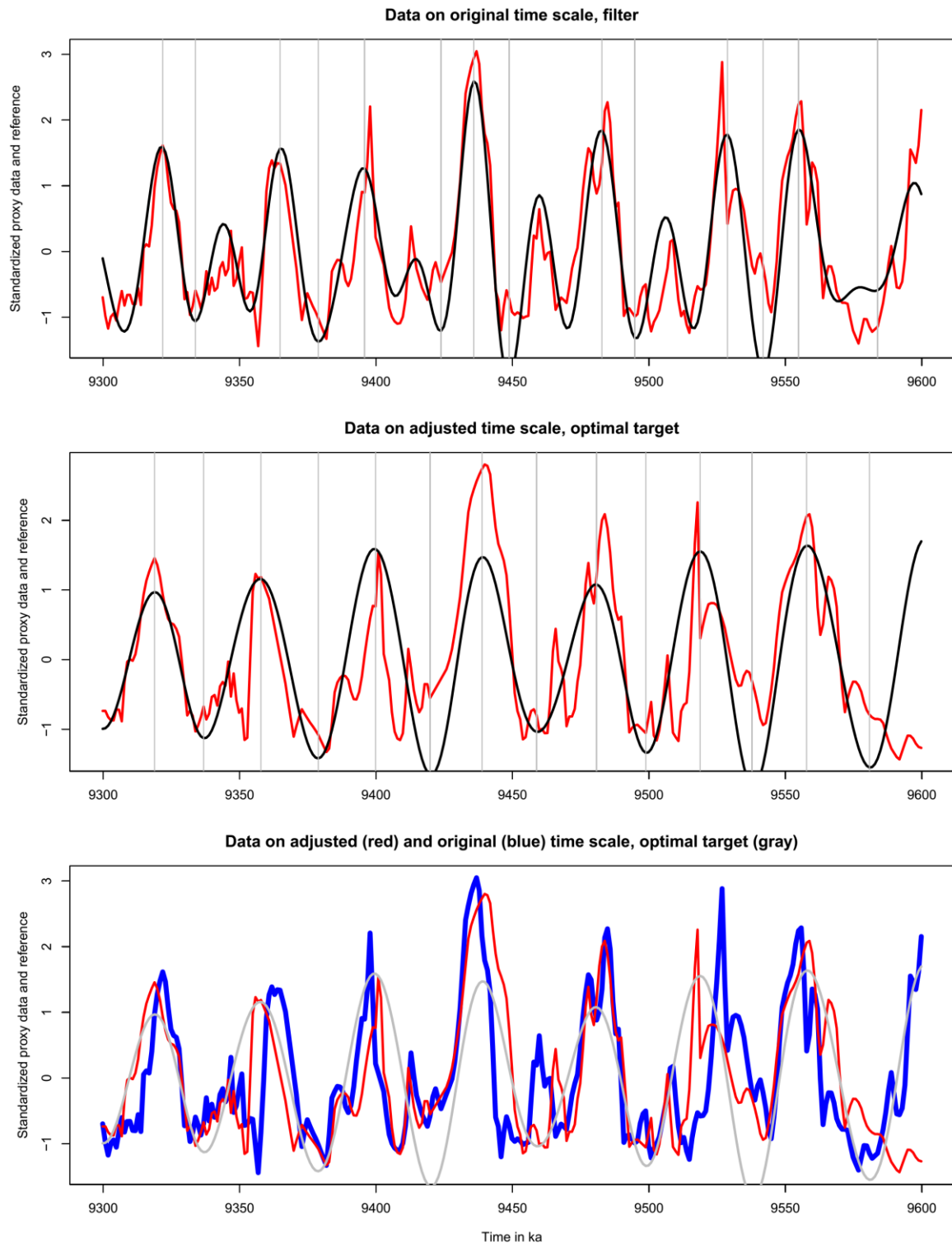
442 Figure 1: Flow chart of the automatic alignment method. Please note that filtering can also  
443 be done in the frequency band of eccentricity and obliquity-eccentricity (see paragraph 2.2).



445  
 446 Figure 2: Grey scale colour data form Monte dei Corvi (Zeeden et al., 2014) on the original  
 447 age scale (top), the adjusted age scale (using a filter comprising both precession and  
 448 obliquity; middle panel), as well as a comparison of data on original and adjusted time scales  
 449 (bottom panel) including the target fitting data best (grey). Vertical lines represent tie points  
 450 between filtered data and the best-fitting correlation target. (for printing: black and white)



452  
 453 Figure 3: Younger and precession dominated part of figure 2. Grey scale colour data form  
 454 Monte dei Corvi (Zeeden et al., 2014) on the original age scale (top), the adjusted age scale  
 455 (using a filter comprising both precession and obliquity; middle panel), as well as a  
 456 comparison of data on original and adjusted time scales (bottom panel) including the target  
 457 fitting data best (grey). Vertical lines represent tie points between filtered data and the best-  
 458 fitting correlation target. (for printing: black and white)

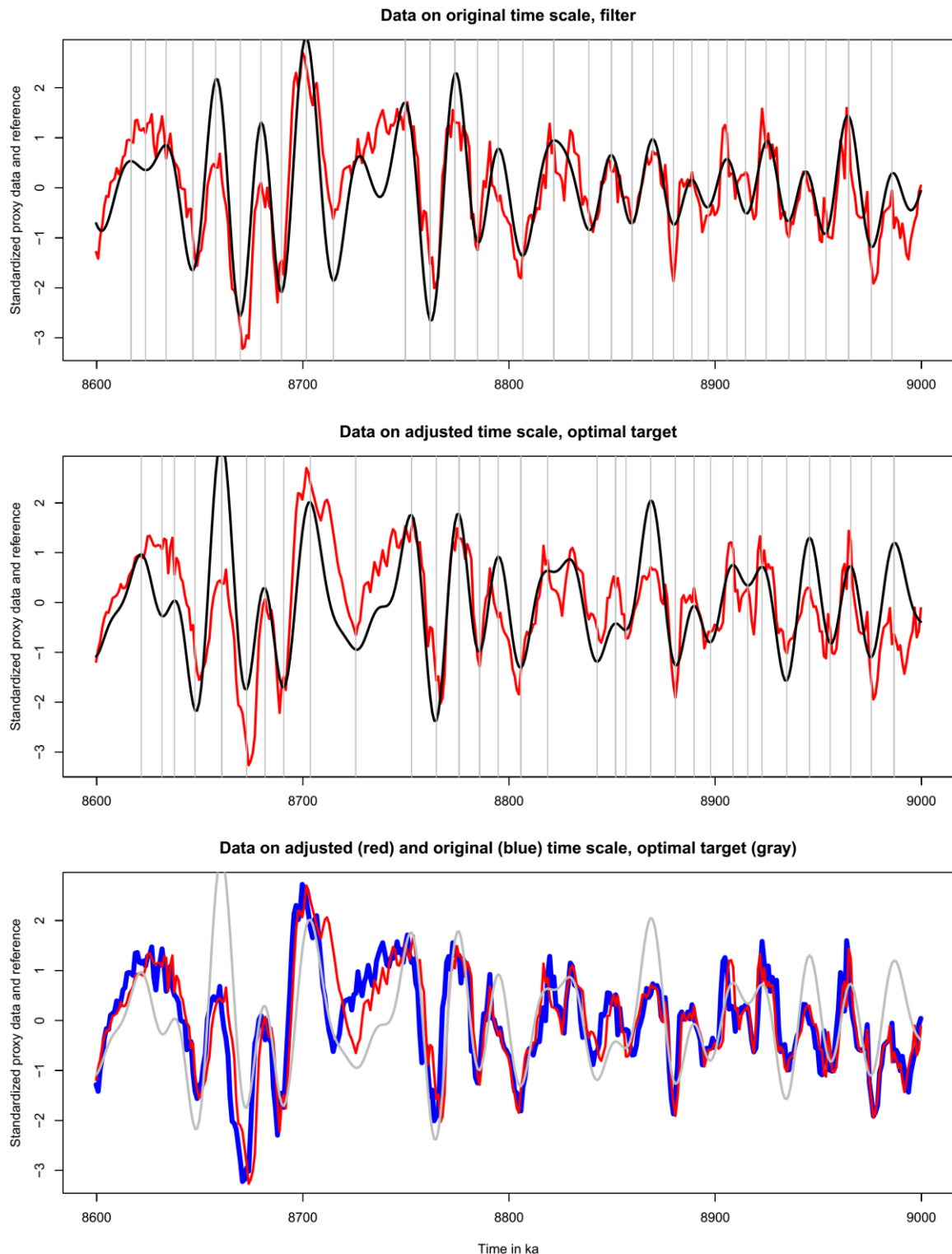


459

460 Figure 4: Older part of figure 2 showing precession-obliquity interference, an obliquity target  
 461 approximates data best. Grey scale colour data from Monte dei Corvi (Zeeden et al., 2014)  
 462 on the original age scale (top), the adjusted age scale (using an obliquity filter comprising  
 463 both precession and obliquity; middle panel), as well as a comparison of data on original and  
 464 adjusted time scales (bottom panel) including the target fitting data best (grey). Vertical lines  
 465 represent tie points between filtered data and the best-fitting correlation target. (for printing:  
 466 black and white)

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Figure 5: Grey scale colour data from the equatorial Atlantic Ceara Rise (Zeeden et al., 2013) on the original age scale (top), the adjusted age scale (using a filter comprising both precession and obliquity; middle panel), as well as a comparison of data on original and adjusted time scales (bottom panel) including the target fitting data best (grey). Vertical lines



474 represent tie points between filtered data and the best-fitting correlation target. (for printing:  
 475 black and white)

476

477 Table 1: Experiments and their respective results for testing precession filter properties. p  
 478 denotes precession, t tilt/obliquity and e eccentricity. Results from testing filter properties are  
 479 displayed in columns 2 to 4. For most results widening filters above a certain level gives  
 480 similarly good correlation. Narrowing these results and avoiding similar correlation  
 481 coefficients as from wider filters seems useful in real cases. Therefore, the narrowest filters  
 482 with high correlation are given in columns 5 to 7. Filter boundaries (upper- and lower  
 483 frequency range boundaries) are given in 1e-3 [1/kyr].

tested data	as by experiments			adjusting results to useful values		
	lower frequency	upper frequency	roll-off rate	lower frequency	upper frequency	roll-off rate
-p+0.5t	0.036	0.080	0.027	0.041	0.054	27
p+0.5t	0.045	0.050	0.040	0.041	0.054	27
-p+0.5t+0.5*AR noise	0.039	0.055	0.022	0.041	0.054	21
-p+0.5t+ equal AR noise	0.041	0.054	0.031	0.041	0.054	31
-p+0.5t+2*AR noise	0.041	0.054	0.031	0.041	0.054	31
p-t+0.5*AR noise	0.047	0.051	0.042	0.041	0.054	31
p-t+ equal AR noise	0.047	0.051	0.042	0.041	0.054	31
p-t+2*AR noise	0.047	0.051	0.042	0.041	0.054	31
p+0.5t+0.5*AR noise; time domain distorted by gamma distribution	0.036	0.051	0.029	0.050	0.054	27
p+0.5t+ equal AR noise; time domain distorted by gamma distribution	0.036	0.051	0.029	0.050	0.054	27
p+0.5t+2*AR noise; time domain distorted by gamma distribution	0.036	0.051	0.029	0.050	0.054	27
p+0.5t+0.5*AR noise; time domain distorted by gamma distribution; clip data at 0	0.041	0.054	0.057	0.041	0.054	27
p+0.5t+ equal AR noise; time domain distorted by gamma distribution; clip data at 0	0.041	0.054	0.063	0.041	0.054	27
p+0.5t+2*AR noise; time domain distorted by gamma distribution; clip data at 0	0.041	0.054	0.063	0.041	0.054	26
mean result	<b>0.041</b>	<b>0.054</b>	<b>0.039</b>	<b>0.043</b>	<b>0.054</b>	<b>28</b>

484

485

486 Table 2: Evolution of suggested filter boundaries through Geological past, using the  
 487 evolution of astronomical frequencies as in (Laskar et al., 2004).

time in Ma	precession low filter boundary	precession high filter boundary	obliquity low filter boundary	obliquity high filter boundary
0	0.043	0.054	0.022	0.029
50	0.044	0.055	0.023	0.030
100	0.045	0.056	0.024	0.031
150	0.046	0.057	0.025	0.033
200	0.048	0.059	0.027	0.034
250	0.049	0.060	0.028	0.035
300	0.051	0.062	0.030	0.037
350	0.052	0.063	0.031	0.038
400	0.054	0.065	0.033	0.040
450	0.056	0.067	0.035	0.042
500	0.057	0.068	0.036	0.044
550	0.059	0.070	0.038	0.045
600	0.061	0.072	0.040	0.047
650	0.063	0.074	0.042	0.050
700	0.065	0.077	0.044	0.052
750	0.068	0.079	0.047	0.054
800	0.070	0.081	0.049	0.056
850	0.072	0.083	0.051	0.059
900	0.075	0.086	0.054	0.061
950	0.078	0.089	0.057	0.064
1000	0.080	0.091	0.059	0.066

488

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