# A Tracer-Based Algorithm for Automatic Generation of Seafloor Age Grids from Plate Tectonic Reconstructions

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# Highlights

- Open-source code to generate seafloor age grids from plate tectonic reconstructions.
- First published seafloor age grids of the Paleozoic's lost ocean basins.
- Paleozoic sea level change inferred from new paleo-oceanic lithosphere age models.

# Abstract

The age of the ocean floor and its time-dependent age distribution control fundamental features of the Earth, such as bathymetry, sea level and mantle heat loss. Recently, the development of increasingly sophisticated reconstructions of past plate motions has provided models for plate kinematics and plate boundary evolution back in geological time. These models implicitly include the information necessary to determine the age of ocean floor that has since been lost to subduction. However, due to the lack of an automated and efficient method for generating global seafloor age grids, many tectonic models, most notably those extending back into the Paleozoic, are published without an accompanying set of age models for oceanic lithosphere. Here we present an automatic, tracer-based algorithm that generates seafloor age grids from global plate tectonic reconstructions with defined plate boundaries. Our method enables us to produce novel seafloor age models for the Paleozoic's lost ocean basins. Estimated changes in sea level based on bathymetry inferred from our new age grids show good agreement with sea level record estimations from proxies, providing a possible explanation for the peak in sea level during the assembly phase of Pangea. This demonstrates how our seafloor age models can be directly compared with observables from the geologic record that extend further back in time than the constraints from preserved seafloor. Thus, our new algorithm may also aid the further development of plate tectonic reconstructions by strengthening the links between geological observations and tectonic reconstructions of deeper time.

#### 1. Introduction

The discovery of a method to determine the age of the present-day oceanic crust, using reversals of the Earth's magnetic field (Vine and Matthews, 1963), gave rise to the recognition that the seafloor is spreading, and ultimately to the development and broad acceptance of plate tectonics. In the halfcentury since the plate tectonic revolution, detailed age models of the present-day oceanic lithosphere have been constructed, and digital global oceanic age grids are continuously refined (Müller et al., 1997, 2008a, 2016). A wealth of information, mainly from marine geophysical data, but also from geology of continental margins, were used to reconstruct the extent and age distribution of oceanic lithosphere of the past, including portions that have been subducted (Müller et al., 2008b). These paleo-seafloor age grids present rich new opportunities for scientific inquiry, as a wide range of Earth processes can be further interrogated with the use of such age grids. Example applications include the estimation of paleobathymetry (spatial and temporal changes in ocean basin depth, which in turn is important for understanding past ocean currents and their effect on paleoclimate, e.g., Straume et al. 2019), sea level change (Müller et al., 2008b), global seafloor heat flow (Loyd et al., 2007; Crameri et al., 2019), and the subduction volume flux, which impacts geomagnetic reversals (Hounslow et al., 2018), the thermal structure of paleo-subduction zones (Maunder et al., 2019), transport of water (Karlsen et al., 2019a) and carbon (Merdith et al., 2019) to the deep mantle, and the slab pull force on tectonic plates (Conrad and Lithgow-Bertelloni, 2004; Faccenna et al., 2012). Seafloor ages for past times are also important as a boundary condition for global mantle convection models (Gurnis et al., 2012).

Present day age grid models are based on a set of isochrons (lines defined by equal seafloor ages) constructed using information from magnetic and gravity data available at various resolutions in most oceanic basins. Ages for seafloor locations between isochrons are computed based on rotation parameters that describe the plate motions for various time intervals. The isochrons and rotation parameters are linked to a specific geomagnetic timescale, and the choice of timescales will influence the calculated values of spreading velocities. To ensure a smooth grid of ocean floor ages that maintains sharp age discontinuities at fracture zones, Müller et al. (1997) designed an algorithm where they first created a set of densely interpolated isochrons along plate flow lines, and then used a minimum curvature routine to obtain age values on a regular grid. This method for reconstructing seafloor age from present-day seafloor age data and a plate kinematic model is time-consuming and requires significant human input, and consequently may be subjective or introduce errors. Because of this, seafloor ages are usually only determined after a plate reconstruction model has been finalized; they have not previously been computed "on the fly" from the plate kinematic model itself.

The mid-ocean ridges constitute the locus of seafloor generation through time, while plate kinematics define the seafloor's subsequent journey until its destruction at a subduction zone. Thus, global plate tectonic reconstructions that define the motions of the plates and the locations and types of plate

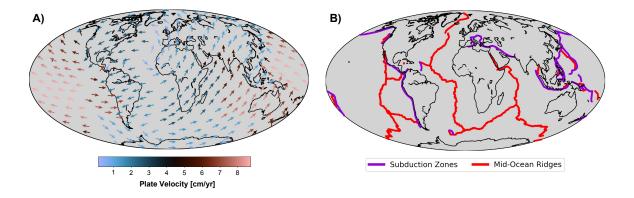


Figure 1: Present day global surface velocity field (A) and locations of mid-ocean ridges and subduction zones (B) derived from global plate tectonic reconstructions (Matthews et al., 2016).

boundaries (see Gurnis et al. 2012 for further descriptions of such models) also implicitly define the age and structure of oceanic basins. Global plate tectonic models with dynamic plate boundaries (Gurnis et al., 2012) have been constructed back into the mid-Paleozoic (410 Ma; Domeier and Torsvik 2014; Matthews et al. 2016), but published paleo-seafloor age grids are only available globally for the last 250 Ma (Müller et al., 2019). This timing discrepancy has, partly, occurred because the reconstruction of global paleo-seafloor age grids from a plate tectonic reconstruction presents a tedious and labor intensive task in the absence of an automatic method. Moreover, these reconstructions are subject to continuous changes as new geological information becomes available. It follows that an automatic and more efficient method for seafloor age determination is needed to allow the use of plate tectonic reconstructions to better decipher Earth's processes and dynamics through time. An automatic method also allows detecting inconsistencies in kinematic models, which would help to improve them. In this study we present such an automatic method for generating seafloor age grids, and introduce a specific implementation of it as an open-source Python code called *Tracer Tectonics* (or *TracTec*) (Karlsen et al., 2019b).

## 2. Methods

The algorithm specifically acts upon "self-closing" plate polygon models (Gurnis et al., 2012) that evolve over time from a set of dynamically evolving plate boundaries. Conventionally, such models comprise a rotation file, a plate boundary file and a continent polygon file. The rotation file (\*.rot file) represents a series of finite rotations that describe the time-dependent motions of each plate, which are identified by their associated Plate ID. The motion of plates by rotations about Euler poles (Greiner, 1999), and the surface velocity field (Fig. 1A), can be described at each point in time  $t \in [0, T]$ , where t = 0 defines the present-day and t = T is the earliest point in time for which the plate reconstruction

is defined. We prefer to refer to t as time, rather than age, to avoid confusion with the age of the seafloor.

The time-dependent motions described by the \*.rot file can be used to reconstruct any framework of points, polylines (continuous lines composed of one or more line segments) and polygons that are ascribed PlateIDs defined in the \*.rot file. The continent polygon file, for example, contains static polygons (tagged with metadata Plate IDs) that represent blocks of continental lithosphere that can be reconstructed and passively rotated through time by the \*.rot file. The plate boundary file contains the polylines that are used (in conjunction with the \*.rot file) to reconstruct the dynamic plate polygons at any time t, following the method of Gurnis et al. (2012). These polylines are associated with metadata tags that identify the type of plate boundary that they represent; in the following, we will be interested in those that are either identified as mid-ocean ridge or subduction zone (Fig. 1B).

We divide the computational approach for generating seafloor age grids from kinematic plate models into three modules: *preprocessing*, *main algorithm* and *post-processing* (Fig. 2). The Python scripts that we developed for each of these modules are part of the TracTec-package, and can be downloaded from: http://doi.org/10.5281/zenodo.3687548.

# 2.1. Preprocessing

As the time-dependent spatial distribution of mid-ocean ridges and subduction zones, together with plate kinematics, dictates the age of the ocean floor, we need to extract these properties from a given full-plate model and output them in a convenient format. To accomplish this, we use pyGPlates, which is a Python-based scripting interface to GPlates (Boyden et al., 2011) that allows for easy automation of such tasks. For each point in time t we extract plate boundaries labelled as mid-ocean ridges from the plate boundary file, and re-sample these polyline segments at user-defined intervals  $\Delta_R$  (50 km by default). Subduction zone plate boundary segments are extracted and sampled in the same manner ( $\Delta_S = 20$  km by default).

#### 2.2. Main Algorithm

To simulate and track the journey of oceanic lithosphere through space and time, from its creation along a mid-ocean ridge until its destruction at a subduction zone, we use tracers. These are numerical particles on which quantities of interest are tracked. The essential property we want to track with the tracers is the age of the oceanic lithosphere. A secondary property we track is the Plate ID associated with each tracer, i.e. the plate to which each tracer currently belongs. We track Plate IDs to determine when a tracer crosses a plate boundary (recognized by a change in its associated Plate ID), which in turn is useful to detect subduction of tracers. At a given time t, the number of tracers is  $N_T(t)$ , their positions are  $\mathbf{x}_n(t)$ , their ages  $\tau_n(t)$ , and their associated Plate IDs are  $p_n(t)$ , for  $n = 0, 1, ..., N_T(t) - 1$ .

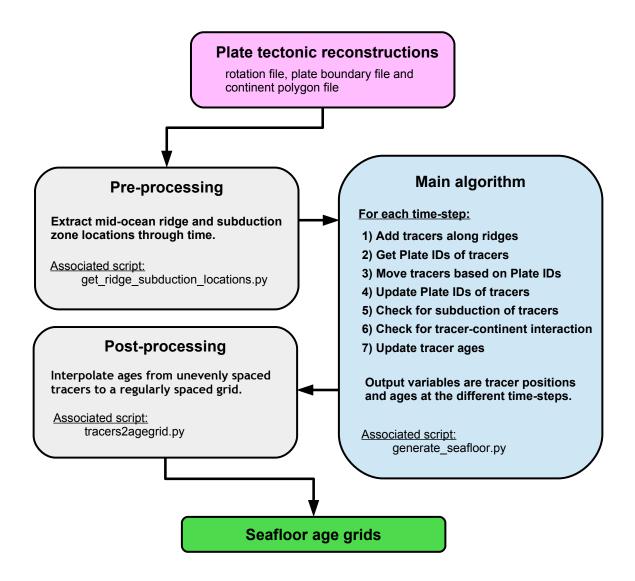


Figure 2: Flow chart of TracTec (Karlsen et al., 2019b) showing an overview of the algorithm and the steps used to generate seafloor age grids from plate tectonic reconstructions.

We break the algorithm into individual sub-steps (see below, 1-7) that are completed at every integer time-step  $t \in [0,T]$ , before moving on to the next. Although our algorithm is capable of operating at any arbitrary time-step, we fix the time-step size  $\Delta t$  to 1 Myr in accordance with the inherent temporal resolution that modern plate tectonic reconstructions are made. The use of larger time-steps would yield sparse age-grids and omit meaningful kinematic data, whereas smaller time-steps could result in erroneous behavior owing to deficiencies in the input plate model below the standard assumed temporal resolution of 1 Myr.

## 1) Add tracers

At the beginning of each time-step we add tracers at intervals given by  $\Delta_R$  on each side of the midocean ridges (Fig. 3A). The tracers are added at a small offset distance  $\epsilon_R$  (1 km by default) from the ridge to ensure that they are within the polygons that define the two spreading plates. Their initial ages  $\tau_n(t)$  are set to zero. The uncertainty introduced by adding tracers at this offset distance is given by  $\epsilon_R/v_s$ , where  $v_s$  is the half-spreading rate. For an average mid-ocean ridge ( $v_s \sim 26$  mm/yr)  $\epsilon_R/v_s \approx 0.04$  Myrs.

# 2) Get tracer Plate IDs

Based on the tracers' positions  $\mathbf{x}_n(t)$ , we use pyGPlates' point-in-polygon algorithm to assign them their Plate IDs  $p_n(t)$ .

#### 3) Move tracers

To determine how the tracers move from t to  $t + \Delta t$  (Fig. 3 B-C), we use pyGPlates to query the rotation file to obtain the finite rotation associated with each tracer (which is based on their Plate IDs  $p_n(t)$ ). Next, by applying that rotation to their current positions  $\mathbf{x}_n$ , we obtain their new positions  $\mathbf{x}_n(t + \Delta t)$ .

## 4) Update tracer Plate IDs

Based on the tracers' new positions  $\mathbf{x}_n(t + \Delta t)$ , we use pyGPlates' point-in-polygon algorithm to assign them their new Plate IDs  $p_n(t + \Delta t)$ .

# 5) Check for subduction

After having moved the tracers from  $\mathbf{x}_n(t)$  to  $\mathbf{x}_n(t+\Delta t)$ , we check if any tracers have been subducted (Fig. 3 H-I). This is accomplished by first determining which tracers have changed Plate ID by comparing  $p_n(t)$  against  $p_n(t+\Delta t)$ . Given the sub-set of tracers that have changed Plate ID, we determine if any are within distance  $R_{min}$  of a subduction zone boundary. The default value of  $R_{min}$  is set to 100 km, which we have found to be appropriate. Any tracers that fulfill both of these criteria are considered subducted and are thus terminated at this time-step.

#### 6) Check for tracer-continent interaction

To ensure that tracers don't end up on continents (which could happen in nature, e.g. ophiolites, but shouldn't happen in conventional dynamic plate polygon models, but nevertheless occurs due to inexorable flaws in such models), we delete all tracers that end up inside continent polygons (checked using pyGPlates' point-in-polygon algorithm against the continent polygon file).

#### 7) Update tracer ages

Before moving on to the next time-step, which practically means returning to step 1) of the algorithm, we update the age of the tracers that are left after the two filtering steps 5-6), by simply by adding  $\Delta t$  to their current age, obtaining  $\tau_n(t + \Delta t)$ .

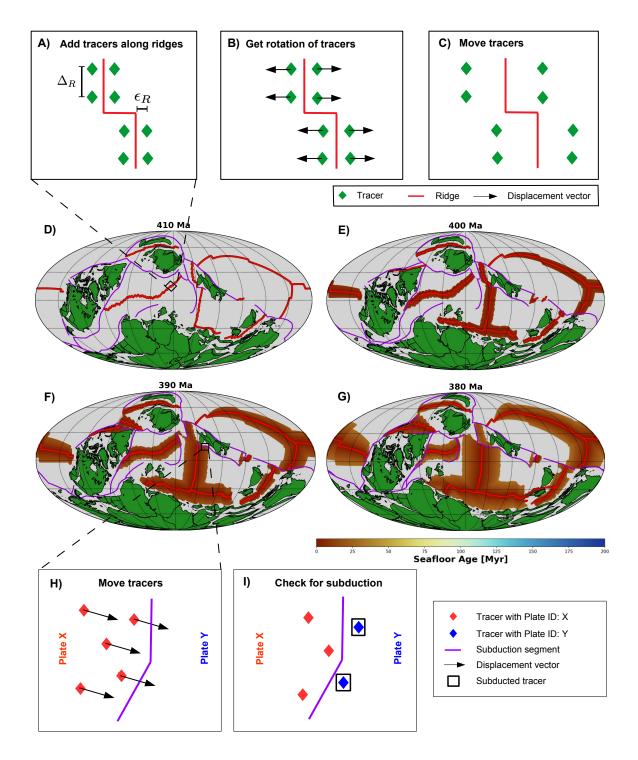


Figure 3: Schematics illustrating how tracers are added along mid-ocean ridges (A), moved (B-C), and checked for subduction (H-I). The repeated application of these steps over time leads to an ocean basin gradually filled with tracers that track the age of the ocean floor (D-G). In this example the model of Matthews et al. (2016) is used. Note that (A-C, H-I) are not to scale.

## 2.3. Post-Processing

The output from steps 1-7) is unevenly distributed tracer positions  $\mathbf{x}_n(t)$  associated with ages  $\tau_n(t)$ . However, for most applications, it is convenient to express the seafloor ages on a regular grid, rather than at arbitrary points. This calls for a post-processing step that interpolates seafloor ages from tracer positions onto a regular grid. There are countless ways of doing this, ranging from simple nearest neighbor algorithms, to weighted means, to splines etc. Depending on the application, smoothing of resulting age grids may be preferred or required. In our online repository we provide an example of a simple post-processing script that uses GMT's linear interpolation algorithm (Wessel et al., 2013) to obtain seafloor ages on a regular grid.

The Python scripts to generate age grids based on the steps described above using Tracer Tectonics are provided in the Zenodo repository (Karlsen et al., 2019b). A summary and a general overview of the algorithm are shown in Figure 2.

#### 2.4. Initial Condition

Running the algorithm without any initial condition, as in the example of Figure 3 D-G, we see that it takes some tens of millions of years before the ocean basins are entirely covered by tracers. Technically, this is the time it takes for the predicted seafloor ages to result solely from the plate kinematics of the input model. Alternatively, one could apply an initial condition that incorporates some educated guess of the seafloor ages for the initial time step. This is straightforward to include in the framework presented (steps 1-7 of the main algorithm), by simply initializing  $\mathbf{x}_n(t=T)$  and  $\tau_n(t=T)$ . As with any time-dependent model, one should be aware of the assumptions implicit to the chosen initial condition, and its effects on the model output. For this algorithm, it is straightforward to track the effect of an applied initial condition. This can be achieved by simply tracking the fraction of initial tracers through time (Fig. S1); at some point, all the initial tracers will be eliminated, from which point the output will no longer depend on the initial condition.

## 3. Discussion

## 3.1. Validation and Benchmarking

To validate our algorithm and its implementation, we compare the seafloor age grids generated by our algorithm against published present-day seafloor age models (Benchmark #1, Fig. 4), direct point observations of present-day seafloor ages (Benchmark #2, Fig. 5) and against the time-dependent age-area distribution of the oceanic lithosphere for the last 230 Myr (Benchmark #3, Fig. 6) based on M16 (Müller et al., 2016).

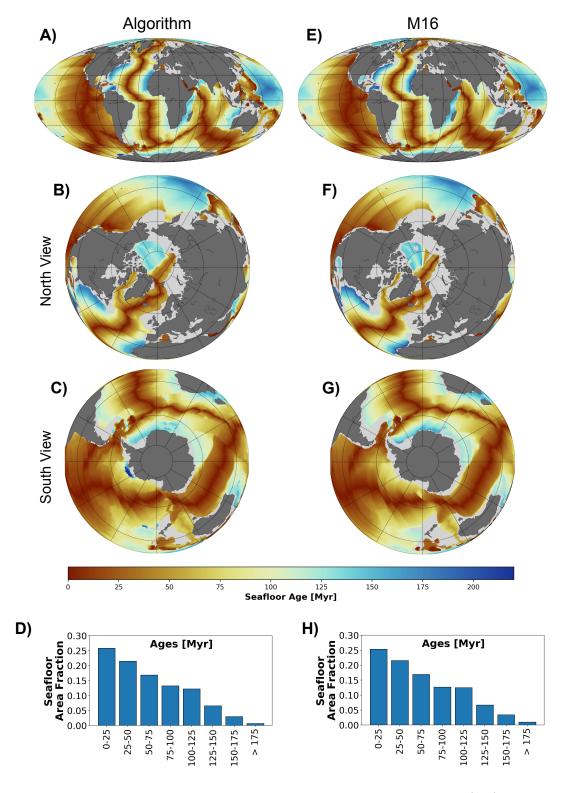


Figure 4: Benchmark #1: Maps comparing present-day seafloor ages generated by our algorithm (A-C) to those of M16 (Müller et al., 2016) model (E-G), and their corresponding age-area distributions (D and H). The input plate model used here was Matthews et al. (2016).

To generate the seafloor age grids used to evaluate the performance of our algorithm (with its default settings), we employed Matthews et al. (2016) as the input plate model with the initial condition shown in Figure 7A. Notably, the effect of an initial condition applied at 400 Ma is very small already by 300 Ma (only 11 % of the initial tracers remain), and zero at the present-day (Fig. S1). The initial condition and the age grids are available from our online repository (Karlsen et al., 2019b).

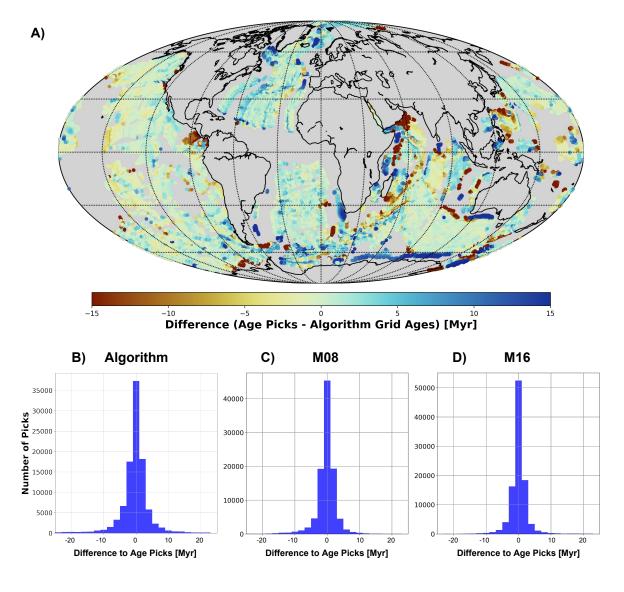


Figure 5: Benchmark #2: Map (A) comparing the present-day age grid generated by our automatic algorithm to ages inferred from magnetic reversal picks (Seton et al., 2014). Distributions (B-D) show the number of pick ages that fall within a certain deviation from (B) our algorithm-generated age grid, (C) M08 - Müller et al. (2008a) and (D) M16 - Müller et al. (2016). The total number of pick ages is 101418.

Our algorithm reproduces the present-day seafloor well (Fig. 4 A-C), as can be seen from the maps comparing the resulting age grids to those of M16 (Fig. 4 E-G). The characteristic triangular present-day seafloor age-area distribution (Sclater et al. 1980; Cogné and Humler 2004), which shows the fraction of the ocean floor that falls within a certain age range, is also well reproduced. On the regional scale, some minor differences can be seen, for example there are three narrow bands of artificially young seafloor branching from the Mid-Atlantic Ridge near the Caribbean and the southwest of Iberia. These are merely consequences of the underlying plate model (Matthews et al., 2016), for which these plate boundaries are erroneously designated as mid-ocean ridges, either at present (Fig. 1), or in the recent past. This demonstrates that a combined work-flow for developing plate tectonic reconstructions, which incorporates the generation and analysis of seafloor age grids, can reveal flaws and inconsistencies in the plate model. In the case of the aforementioned errors in the mid-Atlantic, simply re-labeling the offending boundaries as transform features and re-running the age grid algorithm addresses the issue.

To further evaluate the performance of our algorithm, we compare the generated present-day age grid against ages inferred from magnetic reversal picks (Fig. 5). This global dataset provides by far the most comprehensive, direct sources of oceanic lithosphere ages, and is available from the Global Seafloor Fabric and Magnetic Lineation (GSFML) database (Seton et al., 2014). We see that  $\sim 37$ % of the 101418 pick ages are within  $\pm 1$  Myr of the ages from our age grids, and  $\sim 84$ % are within  $\pm 5$  Myr (Fig. 5B). These numbers are slightly higher for models of the present-day ocean floor that are based on more labor-intensive methods (described in Section 1) such as Muller et al. (2008b), for which they are  $\sim 45$ % and  $\sim 92$ % respectively (Fig. 5C), and  $\sim 52$ % and  $\sim 93$ % for M16 (Fig. 5D).

As pointed out in several studies (e.g. Becker et al. 2009; Coltice et al. 2012; Sim et al. 2016), the triangular age-area distribution of the present-day ocean floor is unlikely to be a constant feature through Earths history. In particular, large fluctuations are predicted to have occurred in the rates of seafloor spreading and global subduction (e.g. Hays and Pitman 1973), which should preclude a constant age-area distribution of the seafloor. Moreover, the age-area distribution of the seafloor is an important feature of our planet because it exhibits a first-order control on e.g. bathymetry, sea level, and planetary cooling through regulation of surface heat flow (i.e. loss of mantle heat). Therefore, as a third benchmark, we compare the time-dependent age-area distributions of the seafloor as generated by our algorithm with those published in Müller et al. (2016).

We observe a clear time-dependence in the age-area distributions computed from our age grids, and this time-dependence is nearly identical to that predicted by M16 (Fig. 6). Given the broad observed trend toward relatively younger seafloor ages between 140-80 Ma, we anticipate that some of these seafloor age variations are related to the widespread development of new ridge systems during the Cretaceous. The initiation of some of these new ridge systems (including the southern mid-Atlantic) was associated with late-stage Pangea breakup, whereas other new ridges appeared in the

Paleo-Pacific basin, probably related to the emplacement of large igneous provinces (Torsvik et al., 2019). The rapid development of these new ridges, producing juvenile oceanic lithosphere, must have been synchronously balanced with increased subduction that would have accelerated the destruction of relatively old seafloor. These two processes combined thus explains the tendency toward relatively younger seafloor during the Cretaceous (Fig. 6).

In summary, our algorithm reproduces detailed reconstructions of seafloor ages like those of M16. We observe only minor regional differences between the algorithm-generated present-day seafloor age grid and the M16 model. These differences mainly occur in regions where Matthews et al. (2016) inferred plate boundary locations that deviate from the interpreted isochrons of M16. This merely demonstrates that our algorithm is an automatic and fast way to detect inconsistencies between data and models. Global features like the age-area distribution, which is important for many geodynamic applications, are robustly reproduced over time. Thus, as plate tectonic reconstructions improve, so will the reliability and regional resolution of the age-grids produced by our algorithm.

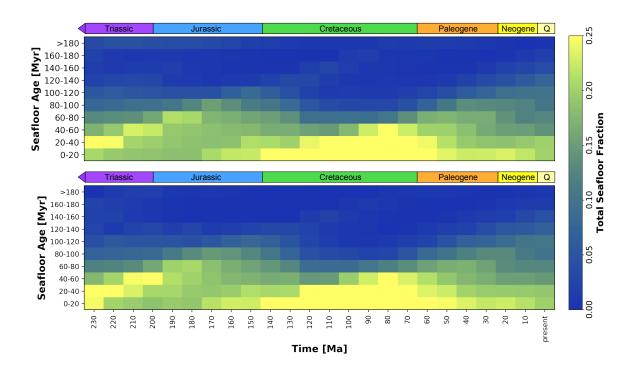


Figure 6: Benchmark #3: Comparison of the age-area distribution of the seafloor through time for our algorithm created age grids (top) and the Müller et al. (2016) (M16) model (bottom).

## 3.2. An Example Application: Paleo Sea Level

In this section we will demonstrate how our new algorithm and paleo-age grids (computed as described in Section 3.1) can be used to study tectonic mechanisms for sea level variations during the last 400 Myr. We would like to stress that the age-grids generated by our algorithm are bound to the input plate model, and as is the case for all plate models operating in times earlier than 200 Ma, the kinematics of all oceanic plates are necessarily synthetic (because no in situ oceanic lithosphere older than 200 Ma has survived to the present-day). This naturally implies that the construction of age-grids for earlier times is much more uncertain, and interpretation of them should be done with caution and care; here we only use these synthetic age grids to consider some global, first-order trends for the sake of demonstration.

From the generated seafloor age grids, we compute bathymetry by applying the age-depth relation of Crosby and McKenzie (2009) (alternative age-depth relations are explored in Figure S2). Next, we use these bathymetry grids (e.g., Fig. 7C) to compute the change in average ocean basin depth relative to the present-day. To account for temporal changes in sediment thickness (which depend directly on seafloor age as well as on latitude) we follow Straume et al. (2019). Finally, we compare how these changes in ocean depth would affect sea level, and relate them to the sea level history (Fig. 7D) as inferred from the sedimentary record (Hallam, 1992; Haq and Al-Qahtani, 2005; Haq and Schutter, 2008). Although many other processes affect sea level fluctuations on tectonic time scales, the age-area distribution of the seafloor (through thermal subsidence of the ageing oceanic lithosphere) exhibits the first-order control, and is the only process that shows a direct correlation with the sea level record (Müller et al., 2008b; Conrad, 2013; Karlsen et al., 2019a). Therefore, an automatic method for generating seafloor age grids (from which ocean depth can be computed) enables sea level to be used as a first-order order deep-time constraint on plate tectonic reconstructions. A selection of our new seafloor age grids from the Paleozoic, with corresponding bathymetry, is shown in Figure S3.

Our prediction of sea level fluctuations caused by ocean depth changes inferred from our 400 Myr reconstruction of age grids shows a first-order agreement with established sea level records (Hallam, 1992; Haq and Al-Qahtani, 2005; Haq and Schutter, 2008). Such a correlation has been noted previously into the Mesozoic (e.g., Müller et al. 2008b; Conrad 2013), but our algorithm applied to the plate reconstructions of Matthews et al. 2016 shows that this correlation may extend back into the late Paleozoic. Our age models predict a clear peak in sea level during the late stages of Pangea assembly (~ 340 Ma, Fig. 7D), in agreement with the early predictions of sea-level change based on conceptual models of a supercontinental cycle (Worsley et al., 1985, 1986; Nance et al., 1986). Moreover, the consistency between predicted and observed sea level indicates that the tectonic rates (seafloor spreading and its counterpart, seafloor subduction) in the underlying plate tectonic model might be reasonable for this period of time.

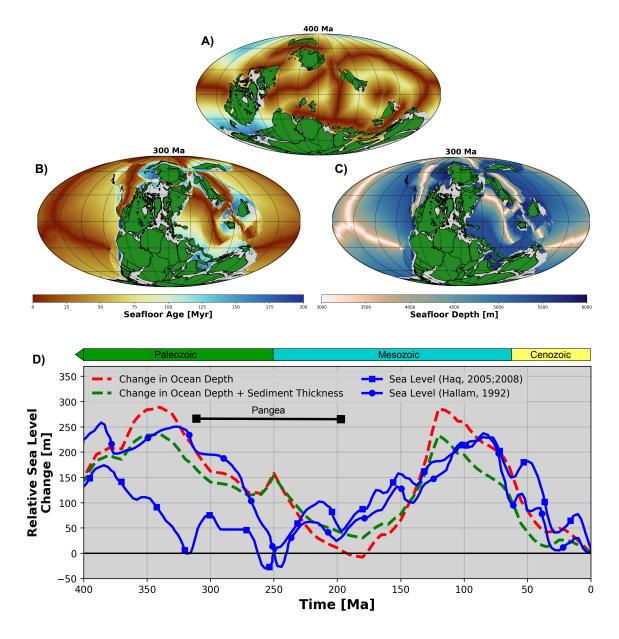


Figure 7: Initial condition (A) used to generate seafloor age grids from the plate reconstruction of Matthews et al. (2016) from 400 Ma to the present. A snapshot at 300 Ma shows seafloor ages (B) and bathymetry (C) computed by applying the age-depth relation of Crosby and McKenzie (2009). Panels (A) and (B) use the same colorbar. From the bathymetry models we compute changes in mean ocean depth and isostatically compensate them (Pitman III, 1978) to compare sea level changes relative to the present day (C, red dashed line) to Phanerozoic sea level reconstructions (C, blue lines) inferred from the sedimentary record (Hallam, 1992; Haq and Al-Qahtani, 2005; Haq and Schutter, 2008). The effect of time-variations in sediment thickness on the sea level prediction (C, green dashed line) is computed after equation 2a in Straume et al. (2019). Thick black bar shows the approximate duration of Pangea.

#### 3.3. Limitations

The uncertainties in the generated seafloor age grids are directly linked to, and controlled by, the underlying plate tectonic model. Thus, the generated age-grids are dependent on global plate motions and the locations of mid-ocean ridges and subduction zones, for which the assignment of formal errors is impractical to impossible. For these reasons, we expect uncertainty in the age grids to be proportional to those of the underlying plate tectonic reconstructions, with negligible additional uncertainty introduced through the rather straightforward computational steps of our algorithm (Fig. 2). A final word of caution is that the generated age grids will never be better than the input plate model. We thus advise users to familiarize themselves with the uncertainties tied to the underlying plate tectonic reconstructions.

# 4. Conclusions

The development of plate tectonic reconstructions in the past decade has been rapid, with the introduction of powerful new tools (Boyden et al., 2011; Clark et al., 2012; Gurnis et al., 2012; Wu et al., 2015; Shephard et al., 2017; Gurnis et al., 2018) and the augmentation of global kinematic datasets (Seton et al., 2014; Gaina and Jakob, 2019) propelling a concomitant proliferation of new and emergent full-plate models, of both regional (Shephard et al., 2013; Zahirovic et al., 2014; Gibbons et al., 2015; Domeier, 2016; Zahirovic et al., 2016; Domeier, 2018; Torsvik et al., 2019) and global scope (Seton et al., 2012; Domeier and Torsvik, 2014; Müller et al., 2016; Matthews et al., 2016; Merdith et al., 2017; Müller et al., 2019). Part of the power of these latest full-plate models is that they implicitly provide information on plate kinematics across the entire global (or regional) surface (Gurnis et al., 2012), and so can be used to derive seafloor age grids to test and evaluate them, as well as to potentially explore other geophysical questions that relate to seafloor ages. Unfortunately, while the information needed to compute seafloor ages is implicitly available, the presently-established workflows to retrieve and process this information are time-consuming, laborious and not publicly available. This emphasizes the need for a method that can automatically generate seafloor age grids from full-plate tectonic reconstructions.

In this study we have presented an algorithm for generating seafloor age grids that robustly predicts the known present-day ocean floor ages, as well as a previously published model for the time-dependent age-area distribution. This new method has been applied to generate the first set of publicly available oceanic lithosphere age grids that approximate the first order age distribution of the Paleozoic oceans. Application of our generated seafloor models to estimate past sea level changes reveals a general agreement with observations from the sedimentary stratigraphy record (Hallam, 1992; Haq and Al-Qahtani, 2005; Haq and Schutter, 2008), and provides a possible explanation for a peak in sea level during the assembly phase of Pangea. We hope that our automated algorithm will enable such comparisons between age-grid predictions and existing geological constraints to become a routine

procedure of the full-plate model development process. Such an improved workflow should ultimately lead to better and more self-consistent combined reconstructions of both plate tectonics and seafloor ages.

# Computer Code Availability

Name of code: Tracer Tectonics (TracTec)

Developer: Krister S. Karlsen (e-mail: k.s.karlsen@geo.uio.no; phone: +47 22 85 40 80).

Year first available: 2019.

Hardware required: No requirements.

Program language: Python 2.7.

Software dependencies: GMT (only post-processing) and the following Python libraries: pyGPlates,

scipy, numpy.

<u>Program size:</u> 86 MB (including benchmark age grids) <u>Source code:</u> http://doi.org/10.5281/zenodo.3687548

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