**A simulation study to compare Pb dating data analyses**

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The increasing interest to understand anthropogenic impacts on the environment have led to a considerable amount of studies that focus on sedimentary records of 100 - 200 years. Dating this period is often complicated by the poor resolution and large errors associated with radiocarbon (14C) ages, which is the most popular dating technique. Instead, sediment dating with lead-210 (Pb) is widely used it provides absolute and continuous dates for 100 – 150 years. The Pb dating method has traditionally relied on the Constant Flux (CF, also known as Constant Rate of Supply - CRS) model which uses the radioactive decay equation resulting in a restrictive model to approximate dates. In this work, we compare the classical approach to Pb dating (CF) and its Bayesian alternative (*Plum*). For this, we created simulated Pb profiles following three different sedimentation processes, complying the assumptions imposed by the CRS model, and analysed them with both approaches. Results indicate that the CRS model does not capture the true values even when the sediment is entirely dated, nor improves its accuracy as more information is available. On the other hand, the Bayesian alternative (Plum) provides consistently accurate results even with few samples, and its accuracy and precision constantly improves as more information is available.

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*Keywords:* Plum, Age-depth models, Chronology, Constant Rate of Supply, Comparison.

# Introduction

Lead-210 (Pb) is a radionuclide,part of the U decay chain, which forms naturally in the atmosphere as well as in sediments. This isotope, with a half-life of 22.23 years, is commonly used to date recent recently accumulated sediments ( to years). Unlike to other dating techniques such as C (radiocarbon dating), using a single measurement of Pb is not possible for dating; it is only when a suitable portion of the decay curve (the total inventory) is measured and with certain assumptions about the sedimentation process are met that a chronology can be established. In recent decades, increasing amounts of palaeoecological and pollution studies have focused on recent sediments in order to evaluate human impacts on the environment. These studies strongly rely on the accuracy of their chronologies in order to correctly assign dates to chemical, biological and ecological changes. That is, unlike other dating techniques, an analysis of a series (data set) of Pb measurements must be carried out in order to obtain meaningful dates. In a lake sediment, or any other, sedimentation process, samples are taken along a core at different depths, from which Pb activity is measured. The whole series of Pb measurements need to be analyzed to attempt to produce a coherent chronology, see .

A range of traditional data analyses, or so called “models”, are available for dating recent sediment using Pb, most notably are the Constant Rate of Supply (CRS), Constant Flux:Constant sedimentation (CF:CS) and Constant Initial Concentration (CIC) models . The CRS model, also known as Constant Flux - (CF) model is by far the most popular (see Figure [1](#fig:210models)) and has the most flexible assumptions. The CRS model assumes a constant supply of Pb to the sediment from the atmosphere and allows for changes in the sedimentation rate. In order to build a chronology, the CRS model uses a ratio between the complete “inventory” (the complete estimate of the radioactivity in the sediment column of the sediment between the surface and the “equilibrium depth” where Pb from the atmosphere can no longer be found) and the remaining inventory from depth (, where is the complete inventory and the decay constant of the Pb ).

Other, more restrictive models such as CF:CS and CIC requiere the assumption of a constant supply of Pb as well as other assumptions of the sedimentation process, as well as that of a constant supply of Pb. The flexibility of the CRS, regarding its assumptions, comes at the cost of the need to measure a sufficient portion of the inventory or the use of interpolation in order to properly estimate the complete inventory of Pb in the sediment.

The CRS model has undergone several revisions in the last decade in order to improve its accuracy and applicability. There are two types of revisions to this model: (1) revisions to its uncertainty estimates and (2) to its application where extra information is available, such as external independent dating markers (e.g. Cs dates) or laminated sediments .

A recent inter-laboratory model comparison experiment presented concerning results. A series of Pb measurements were send to 14 laboratories around the world. Each laboratory was ask to provide a chronology, given the same data. This experiment resulted in a wide range of chronologies not only when different models were used, but even when the same model was applied. The authors strongly recommended to the use of independent time markers (independent dating sources) to validate of the chronologies. This research clearly and critically shows the impact that user decisions have on the resulting chronologies, which becomes extremely important when trying to replicate and/or update the resulting chronologies. Users attending to do so will not only need access to the raw data but also to every user decision in constructing the chronology; unfortunately, these raw data sets and decisions are rarely reported.

![Frequency of ^{210}Pb dating models used in papers between 1964 and 2017. Data gathered by  from a literature review of 271 papers. The models include CF:CS model The Constant Flux - Constant Sedimentation;, CIC (Constant Initial Concentration)  and CRS - Constant Rate of Supply;. ](data:application/pdf;base64,)

Frequency of Pb dating models used in papers between 1964 and 2017. Data gathered by from a literature review of 271 papers. The models include CF:CS model , CIC (Constant Initial Concentration) and CRS - .

Recently presented an alternative to these classical models, by introducing *Plum*, a Bayesian approach to Pb dating. This model treats every data point as originating from a system that includes the sedimentation process as well as the radioactive decay process. It also incorporates an important variable to the inferred processes, namely, the levels of supported Pb, which naturally forms in the sediment and is normally threaded as a hindrance variable. *Plum* assumes that there exists an (unknown) age-depth function that relates depth with calendar age . Conditional on , the following model is assumed for measured Pb for the sediment section form depths to

Here is the supported Pb in the sample and the supply of Pb to the sediment, see for details. The age-depth model is based on a piece-wise linear model constrained by prior information on the sediment’s accumulation rates .

This treatment of the data allows for a formal statistical inference and by using a Bayesian approach all the parameters of the model can be inferred. This differs from the CRS model as the latter uses the decay equation to obtained the age-depth function resulting in a more restrictive age-depth model, removes assumed values of supported Pb before modelling, and does not provide a formal statistical inference. *Plum* has shown to provide accurate results with a realistic precision depending on the different case scenarios - both in simulations as well as for real cores. Under optimal conditions *Plum* and the CRS model have shown to provide similar results with *Plum* providing more realistic uncertainties, with minimal user interaction.

presented a comparison between classical and Bayesian age-depth models construction, both for real and simulated C-dated cores. They concluded that Bayesian age-depth models provide a more accurate result and more realistic uncertainties under a wide range of scenarios. Similarly, *Plum* has shown to provide accurate results with a realistic precision depending on different scenarios, both in simulations as well as for real cores. Under optimal conditions, *Plum* and the CRS model have shown to provide similar dates , with *Plum* providing more realistic uncertainties with minimal user interaction. In this study, we compare Pb dates and uncertainties from the widely applied CRS model (by far the most popular age-depth model for Pb) against *Plum* using simulated cores, i.e. sedimentation “scenarios”. The objective of this study is to test whether the results obtained by , concerning the accuracy and precision of the Bayesian approach, are maintained in a more complex modelling situation, such as the construction of Pb-based age-depth models. We also wish to observe the learning process of each of the models and estimate the amount of information is needed to obtained a reasonable chronology for each model.

The paper is organized as follows: first section sets the tools for the comparison, it describes the simulations of the three different scenarios and we described a parameter which will facilitate the comparison called information percentage. Section 3 describes the comparison for both the overall chronologies and by single depths. Lastly section 4 presents the conclusions and discussion of the results obtained in section 3.

# Experiment Setup: Simulations

In order to observe the accuracy and precision of any model, a known true age-depth function is required. presented a methodology for simulating radiocarbon dates and their uncertainties, while presented an approach for simulating Pb data given an age-depth function . It is important to note that these simulations follow the equations presented by guaranteeing that the CRS assumptions are met. By using the approach presented by for simulating Pb data and the structure of uncertainty estimation presented by , reliable Pb simulated data can be obtained.

## Simulation Construction

Three different scenarios (see Table [1](#tab:sim_param)) were chosen for our sedimentation simulations, each with their own age-depth functions and parameters. These scenarios were selected as they provide three key challenges for the models: Scenario 1 presents an age-depth function which is quite common for recent sediments, with less compaction toward the surface at 0 cm depth; Scenario 2 presents a challenging core structure as the function replicates a drastic and rapid shift in sediment accumulation behaviour around depth 15 cm depth; and lastly Scenario 3 presents a cyclic and periodic change in accumulation rates. Using the age-depth functions and defined parameters defined in Table [1](#tab:sim_param), we obtain the Pb activity, or concentration, at any given depth or interval by integrating the age-depth curve for that interval. These concentrations may be interpreted as error-free measurements, see Figure [2](#fig:true_210) . Because Pb activity measurement is subject to error, we need to replicate the measurement errors. presents error structure for radiocarbon dates. We can use this structure to our Pb measurements as both measurements are subject to similar measurement problems.

Simulated age-depth function and parameters used in each scenario

|  |  |  |  |
| --- | --- | --- | --- |
| Label | Age-depth |  | Supported Pb |
|  | function | () | () |
| Scenario 1 |  | 100 | 10 |
| Scenario 2 |  | 50 | 25 |
| Scenario 3 |  | 500 | 15 |

[tab:sim\_param]

![Simulated sedimentation scenarios with their corresponding ^{210}Pb profiles. Left: Age-depth functions for the three different scenarios (Table 1). Right: Corresponding ^{210}Pb activity profiles in relation to depth.](data:application/pdf;base64,)

Simulated sedimentation scenarios with their corresponding Pb profiles. Left: Age-depth functions for the three different scenarios (Table [1](#tab:sim_param)). Right: Corresponding Pb activity profiles in relation to depth.

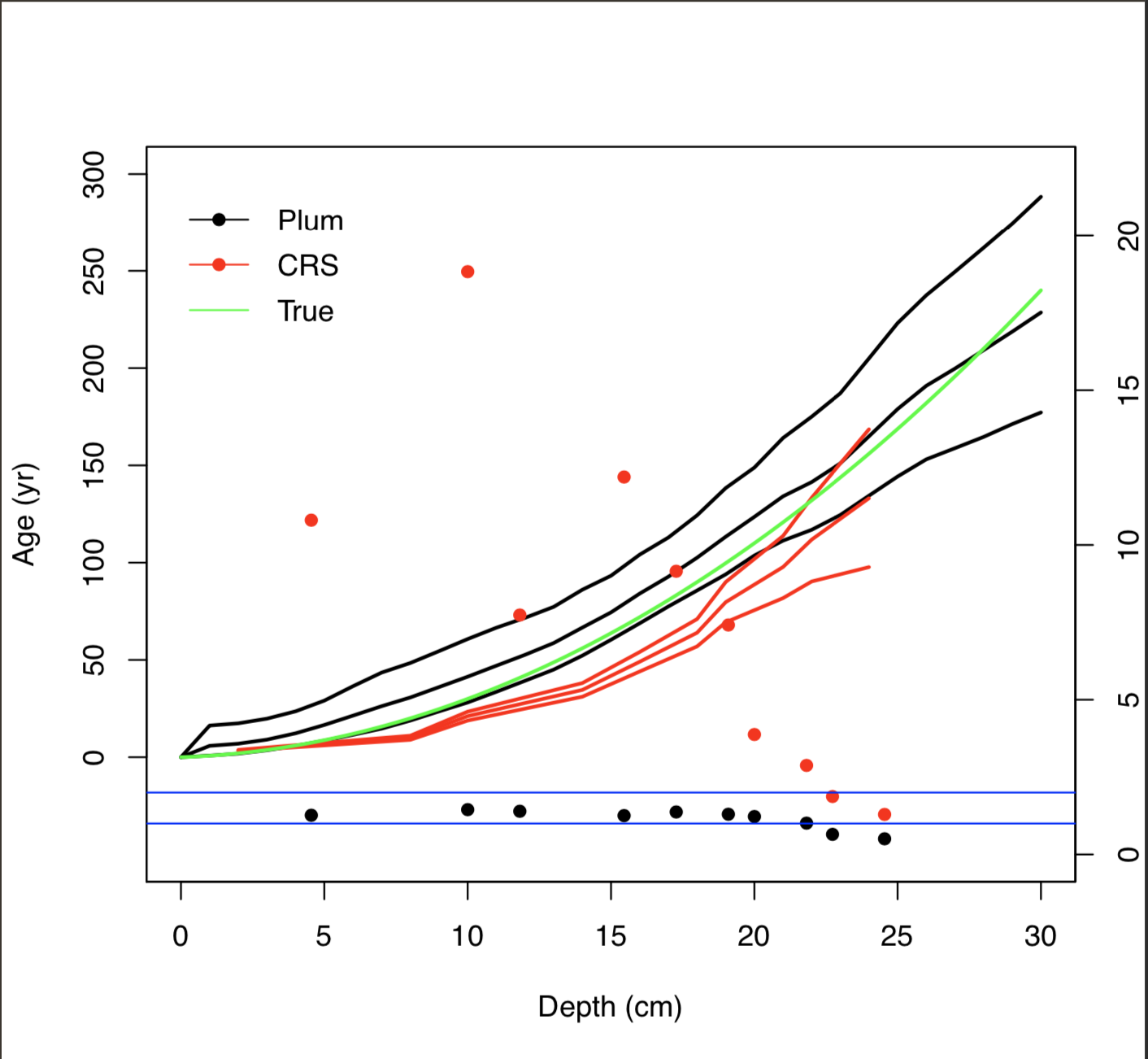
Let be the true Pb concentration in the interval , given the age-depth function and parameters and in each scenario. To simulate disturbances in the material, we can introduce scatter centred around the true value, , where is the amount of scatter for this variable (in this case ). Now, to replicate outliers, a shift from the true value () is defined, which occurs with a probability . This results in a new variable which is defined as

Finally, to simulate the uncertainty provided by the laboratory, we can define the simulated measurements as , where is the standard deviation reported by the laboratory. is defined as , where is the minimum standard deviation assigned to a measurement. This variable differs between laboratories,we use a default value of . Finally, is the analytical uncertainty (default .01) and an error multiplier (default 1.5). The default parameters were set in accordance with .

For this this study we created a data set for each of the three simulation by integrating in intervals of 1 cm, for depths from 0 to 30 cm where radioactive equilibrium was guaranteed . The complete simulated Pb data sets can be found in the Supplementary Material [5](#sec:supp_mat).

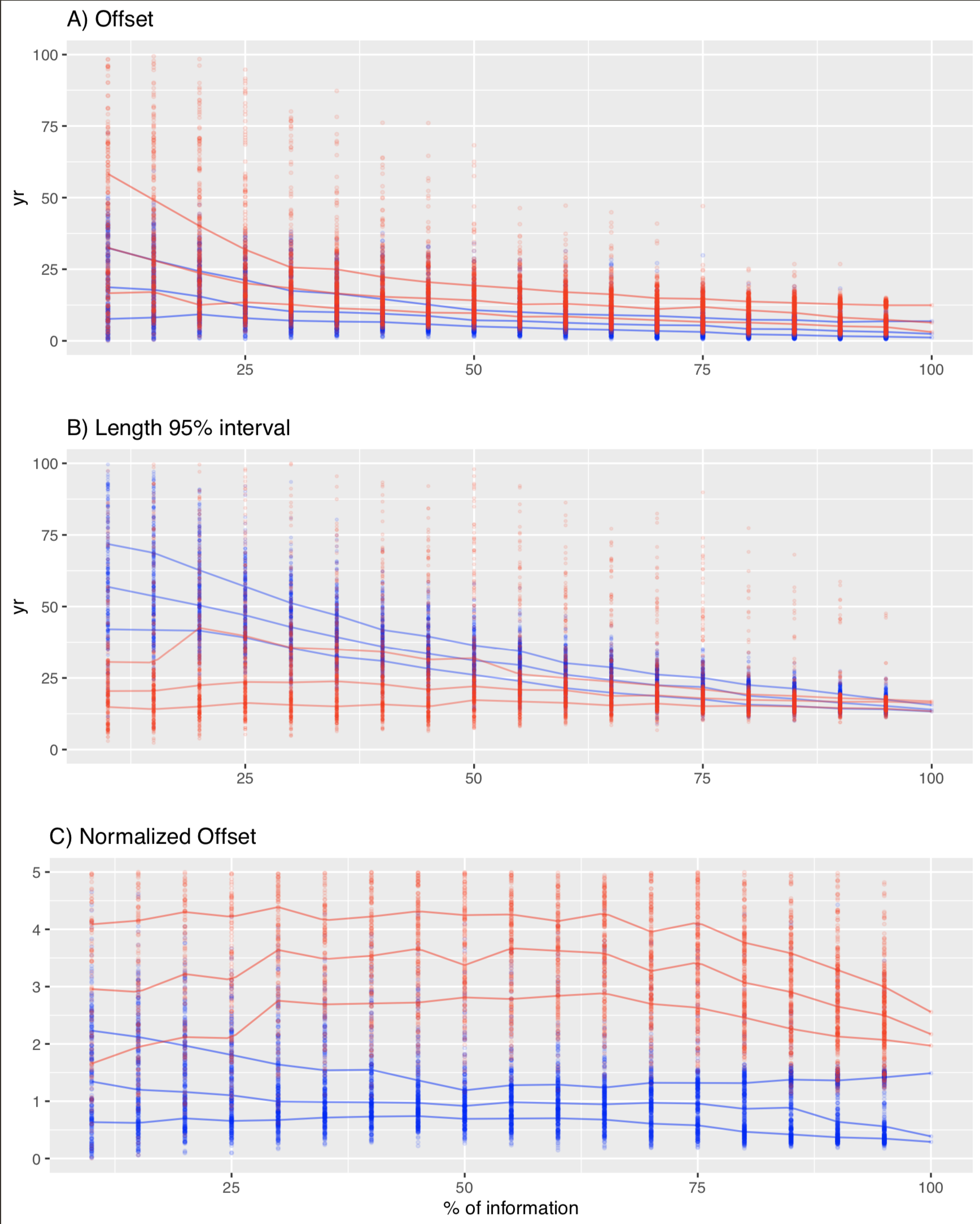
# Model Comparison

To allow for a reasonable comparison between models, and to evaluate the effect that different percentages of information may have on the accuracy and precision of Pb models, we used our three simulated data sets (see previous section). For these simulated cores, samples were randomly selected given a percentage of information (e.g. for a 20% information data set, 6 random 1-cm samples were selected of a possible total 30 1-cm samples). As the CRS model assumes that background has been reached, in order to reduce user manipulation, we decided to fix the last sample (30 cm depth) for every case. This step not only guarantees the consistent application the CRS model, it also provides the model with single bottom-most depth to be removed as it is common practice when using the CRS model. 100 different samples were randomly selected for information percentages from 10% to 95% at 5% intervals (i.e., 10%, 15%, 20%,...,95%) and the complete sample was also used (i.e 100% percentage of information sample). After a random sample was selected, both the CRS model and *Plum* were applied to the data set. To have an objective comparison, both models were run with their default configuration (*Plum* with default settings and CRS estimates and uncertainties as described in ). Both sets of outputs were then compared against the true known age value.



Comparison between *Plum* and the CRS model against the true age-depth model using 50% of the information percentage (using 1-cm samples at depths 2, 8, 10, 14, 16, 18, 19, 21, 22, 24, 25, 27, 28, 29, 30). Lines show the age estimates with the 95% credible intervals (*Plum*) and the 95% confidence interval (CRS). Dots show the normalized offset, distance between the inferred age and the true age in relation to the standard error (the standard deviation in the case of the CRS and the length of the confidence interval divided by 4 in the case of *Plum*).

Figure [3](#fig:comparison1r) shows an example of the comparison between the Pb models against the true value. As we are dealing with a total of 5333 simulations, in order to evaluate the overall precision and accuracy of bothmodels, we decided to calculate the mean offset to the true age-depth model (in yr), the mean of length of the 95% intervals (in yr), as well as the mean normalized accuracy indicating the distance of modelled ages from the true value given the model’s own uncertainty at each depth.

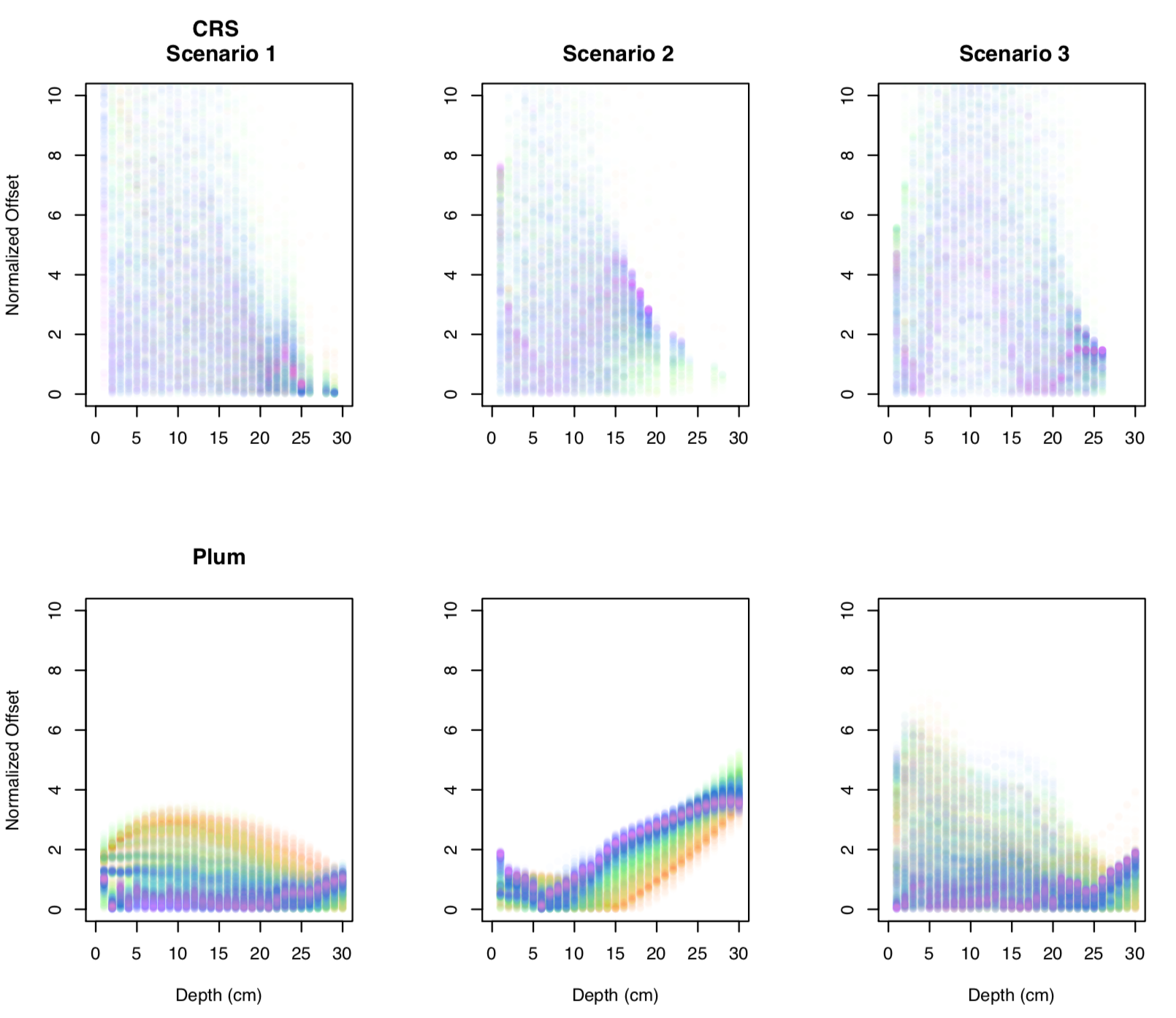


Top panel A) shows the offset between the modelled and true age of the CRS (red) and *Plum* (blue). This panel shows how *Plum* provides a small offset in almost every scenario with both models improving their offset as more information is available. Middle panel B) shows the 95% confidence intervals. It is clear, from this panel, than the uncertainty provided by *Plum* is a lot bigger for low percentage of information and it constantly improves as more data is available, whereas the length of the intervals provided by the CRS appear to stay constant regardless of the available information. Bottom panel C) shows the normalized offsets, presenting the distance between the modelled age and the true age normalized divided by the standard deviation (in the case of *Plum*, the length of the 95% interval divided by 4). This panel presents a worrying situation where the CRS model’s calculated standard deviation (on average) is incapable of of capturing the true age. On the other hand, *Plum*’s credible intervals almost always capture the true age even when little information is available.

Figure [4](#fig:accpre) show results similar to those presented by . The classical model (CRS) at first appears to provide a similar results (similar offsets) to the Bayesian alternative (*Plum*), but at higher estimated precision (if we only look at the length of the 95% interval). These results can be misleading if we do not consider both the effects of both the offset and length of the interval together. To have a more realistic representation of how the models capture the true age-depth models relationship, we should observe the normalized offset. This variable shows the degree the average models contain the truth within their uncertainty intervals (normalized to one standard deviation). Any model with a normalized offset larger than two (two standard deviations) is incapable of capturing the true ages within its uncertainty intervals. This means that, while the CRS estimates smaller uncertainties, it does so at the cost of its accuracy. It also appears that the length of the 95% interval and offset are not affected by how much information is provided to the CRS model.

On the other hand, *Plum* seems to provide increasingly accurate results as more information is added to the model. This again coincides with the results outlined by . When we observe the regular offset (not normalized), we find that textitPlum provides a smaller offset in comparison to the CRS model; this in combination with slightly larger modelled uncertainties results in more consistently accurate age-depth models which are capable of capturing the true values within their uncertainty intervals. This result supports the claim that *Plum* provides more realistic uncertainties than those of the CRS. Another important statistic to take into account is that 87.86% (4686/5333) of *Plum*’s runs remain within the 2 standard deviations, opposed to 7.48% (399/5333) for the CRS model. Furthermore, only 0.54% (29/5333) of the CRS model runs remain under the 1 standard deviation, which is the most commonly reported interval when reporting CRS results. We can also observe a clear structure in the way that *Plum* increases its accuracy and precision to obtained a better chronology as more information is available, whereas the CRS model does not appears to learn from additional data.

These results are presented for the overall chronology (the mean offset, interval and normalized offset of the overall chronology). In order to evaluate whether certain models are better predicting ages at certain section of the sediment cores, we have to look at the normalized offset of every depth.



Normalized offset of every sampling sample at every depth for the three simulated scenarios - CRS age estimates at samples depths and *Plum*’s age estimates at 1 cm intervals. Dots go from lowest information percentage samples (few dated depths; red) to high percentage samples (nearly completely dated cores; purple). The CRS’s normalized offset shows no structure at any particular depth regardless of the available information. This means that the model can provide a reasonable chronology with low levels of information or a very inaccurate age estimate with high levels of information at any given depth resulting in a distrustful age-depth model. On the other hand, *Plum* demostrares a systematic improvement in its age estimates as more data is available. This results assures that a Bayesian approach would consistently provide more reliable results.

Figure [5](#fig:depths) shows the normalized accuracy of every simulation according to depth for both models. *Plum* shows a clear learning structure which depends on the information available to the model. The information percentage appears to be irrelevant to the accuracy of the CRS model, contrary to the results obtained by *Plum*. It is important to note that the inaccuracies of the CRS model are not exclusive to any particular sections of the chronology; this is most likely driven by the small uncertainties estimated by the CRS model.See below for a discussion of how *Plum* behaved in sedimentation simulation 2.

# Discussion and Conclusions

These results clearly show the biases associated with the CRS model. discussed this point and states that the bias is the product of the use of a logarithmic function for the age-depth model. It is also evident from these results that the CRS model’s uncertainty estimates are not sufficient to capture the true age-depth function. This is an important point given the fact that it is common practice among the Pb dating community to report credible intervals to one single deviation, instead of the 95% confidence intervals, which have become the norm in most other chronological reconstructions. Other confidence intervals can be calculated for this model but the fact that these intervals are even smaller than the ones obtained by error propagation is of concern.

Previous work on model comparison has shown the problems with the variability of Pb dating results from different users, even when the same or similar models are applied to a single data set. In our study, user input was reduced to the minimum in an effort to show the potential effects that different percentage of information have on the resulting chronology. The results of this experiment showed that the CRS model can provide extremely different results with data originating from the same data set, even while the effect of user input is mitigated. Figure [4](#fig:accpre) showed that the CRS model appears not to learn from using more data. This explains why over the years, many authors have insisted in the use of other dating techniques to validate the chronologies provided by the CRS model , a point which these results also highly encourage.

On the other hand, *Plum* shows a consistently accurate result by capturing the true values within the 95% credible intervals for most of the simulated sampling strategies. It is important to note that one of the big advantages of *Plum* is its increase in accuracy and precision as more data becomes available. In the case of scenario 2, we observe that *Plum* appears to behaves worse as more data is available, which would be of concern if we did not take into consideration that this sedimentation simulation was extremely unusual in the real world and also if users would not double check the resulting chronologies. If more information is available about the core or the sediment, such as Pb influx rates, prior sedimentationrates or independent dating marker, this information can easily be implemented as prior information in *Plum*, and this will result in a much better chronology. It is also important to note that even when *Plum* performs less well in this specific case, it is providing a much better chronology in upper first 10-15 cm of the core than that of the chaotic CRS age-depth model model.

In conclusion, the use of CRS can only be recommended if users may validate their chronology with external dates as the Cs time markers or other time markers. If not additional information is available for this validation, the use of *Plum*, the Bayesian approach, is still valid, specially if provide it with as as many Pb measurements as possible (at least 60% of the available information of the core). Even so, additional dating information may also be formally included in *Plum* to improve the resulting chronology .

# Supplementary Material

Data for each simulation and code used is hosted at: https://github.com/maquinolopez/Paper\_Simulations

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Label | Depth | Density | 210Pb | sd(210Pb) | Thickness | 226Ra | sd(226Ra) |
|  | (cm) | () | (Bq/kg) |  | (cm) | (Bq/kg) |  |
| Sim01-01 | 1 | 0.10009 | 63.50103 | 2.85755 | 1 | 23.8045 | 1.125 |
| Sim01-02 | 2 | 0.10064 | 80.08738 | 3.60393 | 1 | 23.2924 | 1.125 |
| Sim01-03 | 3 | 0.10173 | 98.32806 | 4.42476 | 1 | 23.434 | 1.125 |
| Sim01-04 | 4 | 0.10334 | 125.45705 | 5.64557 | 1 | 26.0873 | 1.125 |
| Sim01-05 | 5 | 0.10547 | 141.27971 | 6.35759 | 1 | 22.8041 | 1.125 |
| Sim01-06 | 6 | 0.10809 | 130.27571 | 5.86241 | 1 | 23.4333 | 1.125 |
| Sim01-07 | 7 | 0.11116 | 134.04051 | 6.03182 | 1 | 25.6156 | 1.125 |
| Sim01-08 | 8 | 0.11466 | 129.69245 | 5.83616 | 1 | 26.1371 | 1.125 |
| Sim01-09 | 9 | 0.11855 | 134.93655 | 6.07214 | 1 | 25.4813 | 1.125 |
| Sim01-10 | 10 | 0.12278 | 109.39886 | 4.92295 | 1 | 25.8877 | 1.125 |
| Sim01-11 | 11 | 0.12731 | 110.68133 | 4.98066 | 1 | 24.4414 | 1.125 |
| Sim01-12 | 12 | 0.13209 | 102.38094 | 4.60714 | 1 | 24.9053 | 1.125 |
| Sim01-13 | 13 | 0.13706 | 75.80895 | 3.4114 | 1 | 22.9151 | 1.125 |
| Sim01-14 | 14 | 0.14218 | 77.60406 | 3.49218 | 1 | 24.4808 | 1.125 |
| Sim01-15 | 15 | 0.14738 | 68.4401 | 3.0798 | 1 | 24.9343 | 1.125 |
| Sim01-16 | 16 | 0.15262 | 60.72037 | 2.73242 | 1 | 25.2659 | 1.125 |
| Sim01-17 | 17 | 0.15782 | 50.28147 | 2.26267 | 1 | 22.961 | 1.125 |
| Sim01-18 | 18 | 0.16294 | 44.24641 | 1.99109 | 1 | 22.9139 | 1.125 |
| Sim01-19 | 19 | 0.16791 | 39.85997 | 1.7937 | 1 | 28.3774 | 1.125 |
| Sim01-20 | 20 | 0.17269 | 38.40823 | 1.72837 | 1 | 23.5379 | 1.125 |
| Sim01-21 | 21 | 0.17722 | 32.75922 | 1.47416 | 1 | 25.4363 | 1.125 |
| Sim01-22 | 22 | 0.18145 | 28.02545 | 1.26115 | 1 | 24.8995 | 1.125 |
| Sim01-23 | 23 | 0.18534 | 27.8749 | 1.25437 | 1 | 22.6783 | 1.125 |
| Sim01-24 | 24 | 0.18884 | 30.74797 | 1.38366 | 1 | 24.8575 | 1.125 |
| Sim01-25 | 25 | 0.19191 | 28.36187 | 1.27628 | 1 | 24.8724 | 1.125 |
| Sim01-26 | 26 | 0.19453 | 27.24535 | 1.22604 | 1 | 24.3778 | 1.125 |
| Sim01-27 | 27 | 0.19666 | 23.59236 | 1.06166 | 1 | 24.7209 | 1.125 |
| Sim01-28 | 28 | 0.19827 | 25.74855 | 1.15868 | 1 | 24.6615 | 1.125 |
| Sim01-29 | 29 | 0.19936 | 25.05368 | 1.12742 | 1 | 24.7199 | 1.125 |
| Sim01-30 | 30 | 0.19991 | 25.0065 | 1.12529 | 1 | 24.4937 | 1.125 |

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Label | Depth | Density | 210Pb | sd(210Pb) | Thickness | 226Ra | sd(226Ra) |
|  | (cm) | () | (Bq/kg) |  | (cm) | (Bq/kg) |  |
| Sim02-01 | 1 | 0.1001 | 909.3928 | 40.9227 | 1 | 8.9761 | 0.45 |
| Sim02-02 | 2 | 0.1006 | 683.9989 | 30.7799 | 1 | 10.0607 | 0.45 |
| Sim02-03 | 3 | 0.1017 | 453.0503 | 20.3873 | 1 | 9.8701 | 0.45 |
| Sim02-04 | 4 | 0.1033 | 310.7897 | 13.9855 | 1 | 10.37 | 0.45 |
| Sim02-05 | 5 | 0.1055 | 218.0058 | 9.8103 | 1 | 10.0418 | 0.45 |
| Sim02-06 | 6 | 0.1081 | 158.6974 | 7.1414 | 1 | 10.104 | 0.45 |
| Sim02-07 | 7 | 0.1112 | 113.9062 | 5.1258 | 1 | 10.2049 | 0.45 |
| Sim02-08 | 8 | 0.1147 | 75.5493 | 3.3997 | 1 | 9.334 | 0.45 |
| Sim02-09 | 9 | 0.1185 | 56.6252 | 2.5481 | 1 | 10.5145 | 0.45 |
| Sim02-10 | 10 | 0.1228 | 44.1595 | 1.9872 | 1 | 9.8677 | 0.45 |
| Sim02-11 | 11 | 0.1273 | 34.7448 | 1.5635 | 1 | 9.7694 | 0.45 |
| Sim02-12 | 12 | 0.1321 | 25.384 | 1.1423 | 1 | 10.5134 | 0.45 |
| Sim02-13 | 13 | 0.1371 | 24.0007 | 1.08 | 1 | 10.4589 | 0.45 |
| Sim02-14 | 14 | 0.1422 | 21.3643 | 1 | 1 | 9.9504 | 0.45 |
| Sim02-15 | 15 | 0.1474 | 17.7932 | 1 | 1 | 10.5135 | 0.45 |
| Sim02-16 | 16 | 0.1526 | 15.0416 | 1 | 1 | 10.3362 | 0.45 |
| Sim02-17 | 17 | 0.1578 | 14.2937 | 1 | 1 | 10.5131 | 0.45 |
| Sim02-18 | 18 | 0.1629 | 12.3844 | 1 | 1 | 10.368 | 0.45 |
| Sim02-19 | 19 | 0.1679 | 12.6023 | 1 | 1 | 10.5297 | 0.45 |
| Sim02-20 | 20 | 0.1727 | 11.9329 | 1 | 1 | 10.0924 | 0.45 |
| Sim02-21 | 21 | 0.1772 | 9.301 | 1 | 1 | 10.118 | 0.45 |
| Sim02-22 | 22 | 0.1815 | 10.7777 | 1 | 1 | 10.249 | 0.45 |
| Sim02-23 | 23 | 0.1853 | 12.9491 | 1 | 1 | 10.134 | 0.45 |
| Sim02-24 | 24 | 0.1888 | 10.6571 | 1 | 1 | 10.1151 | 0.45 |
| Sim02-25 | 25 | 0.1919 | 9.6297 | 1 | 1 | 9.6608 | 0.45 |
| Sim02-26 | 26 | 0.1945 | 8.4331 | 1 | 1 | 8.7821 | 0.45 |
| Sim02-27 | 27 | 0.1967 | 10.4921 | 1 | 1 | 9.8995 | 0.45 |
| Sim02-28 | 28 | 0.1983 | 11.135 | 1 | 1 | 9.2481 | 0.45 |
| Sim02-29 | 29 | 0.1994 | 10.109 | 1 | 1 | 10.4398 | 0.45 |
| Sim02-30 | 30 | 0.1999 | 9.5404 | 1 | 1 | 10.1114 | 0.45 |

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Label | Depth | Density | 210Pb | sd(210Pb) | Thickness | 226Ra | sd(226Ra) |
|  | (cm) | () | (Bq/kg) |  | (cm) | (Bq/kg) |  |
| Sim03-01 | 1 | 0.1001 | 6384.1354 | 287.2861 | 1 | 15.8007 | 0.675 |
| Sim03-02 | 2 | 0.1006 | 3550.0809 | 159.7536 | 1 | 14.5245 | 0.675 |
| Sim03-03 | 3 | 0.1017 | 1954.5702 | 87.9557 | 1 | 15.6527 | 0.675 |
| Sim03-04 | 4 | 0.1033 | 1183.8917 | 53.2751 | 1 | 14.5175 | 0.675 |
| Sim03-05 | 5 | 0.1055 | 760.2132 | 34.2096 | 1 | 14.9242 | 0.675 |
| Sim03-06 | 6 | 0.1081 | 360.2553 | 16.2115 | 1 | 14.801 | 0.675 |
| Sim03-07 | 7 | 0.1112 | 212.9402 | 9.5823 | 1 | 14.8738 | 0.675 |
| Sim03-08 | 8 | 0.1147 | 104.2684 | 4.6921 | 1 | 14.9028 | 0.675 |
| Sim03-09 | 9 | 0.1185 | 44.3849 | 1.9973 | 1 | 15.0768 | 0.675 |
| Sim03-10 | 10 | 0.1228 | 18.6447 | 1 | 1 | 15.3764 | 0.675 |
| Sim03-11 | 11 | 0.1273 | 23.2778 | 1.0475 | 1 | 14.6231 | 0.675 |
| Sim03-12 | 12 | 0.1321 | 53.1587 | 2.3921 | 1 | 15.1629 | 0.675 |
| Sim03-13 | 13 | 0.1371 | 97.363 | 4.3813 | 1 | 14.3047 | 0.675 |
| Sim03-14 | 14 | 0.1422 | 116.9788 | 5.264 | 1 | 14.0261 | 0.675 |
| Sim03-15 | 15 | 0.1474 | 153.2901 | 6.8981 | 1 | 15.9723 | 0.675 |
| Sim03-16 | 16 | 0.1526 | 151.8496 | 6.8332 | 1 | 14.7579 | 0.675 |
| Sim03-17 | 17 | 0.1578 | 136.3609 | 6.1362 | 1 | 16.114 | 0.675 |
| Sim03-18 | 18 | 0.1629 | 107.2736 | 4.8273 | 1 | 15.4595 | 0.675 |
| Sim03-19 | 19 | 0.1679 | 76.8966 | 3.4603 | 1 | 15.9439 | 0.675 |
| Sim03-20 | 20 | 0.1727 | 48.9213 | 2.2015 | 1 | 14.6235 | 0.675 |
| Sim03-21 | 21 | 0.1772 | 40.4439 | 1.82 | 1 | 14.6716 | 0.675 |
| Sim03-22 | 22 | 0.1815 | 26.5638 | 1.1954 | 1 | 16.2541 | 0.675 |
| Sim03-23 | 23 | 0.1853 | 21.714 | 1 | 1 | 14.4826 | 0.675 |
| Sim03-24 | 24 | 0.1888 | 17.6428 | 1 | 1 | 15.5109 | 0.675 |
| Sim03-25 | 25 | 0.1919 | 17.3533 | 1 | 1 | 13.6898 | 0.675 |
| Sim03-26 | 26 | 0.1945 | 17.4211 | 1 | 1 | 14.4684 | 0.675 |
| Sim03-27 | 27 | 0.1967 | 16.4246 | 1 | 1 | 15.3889 | 0.675 |
| Sim03-28 | 28 | 0.1983 | 12.4828 | 1 | 1 | 15.0698 | 0.675 |
| Sim03-29 | 29 | 0.1994 | 13.5514 | 1 | 1 | 15.2346 | 0.675 |
| Sim03-30 | 30 | 0.1999 | 14.3145 | 1 | 1 | 14.7846 | 0.675 |

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