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Single-grain OSL chronologies for the Still Bay and Howieson's Poort industries and the transition between them: Further analyses and statistical modelling

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Response to copy editing

We have looked at and accepted all changes to the document suggested by the copy editor and have specifically done the following changes:

1. provided a corresponding author and e-mail address
2. shortened the abstract to be 313 words
3. swapped on all occasions the Jacobs et al. (2008c) and Jacobs et al. (2008a) citations so that a comes before c
4. changed the Table captions and additions to the footnotes
5. removed the captions from the figures and provide these as Tiff files.

These are two occasions where we chose not to make changes. Both these relate to comments in the main text where the copy editor requested minimal overlap between the text and figure captions. Although there is some overlap, we still feel that there are good reasons for using the wording etc. in the main text to draw out the details of the figures and prefer to leave it as it is.
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Keywords

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1. Introduction

The Still Bay (SB) and Howieson’s Poort (HP) are two widely distributed Middle Stone Age (MSA) industries in southern Africa. Still Bay and HP artefacts have been recovered from sites spread across two million square kilometres of South Africa, Lesotho and Namibia. They exhibit a range of early technological and behavioural innovations, including the complex processing of ochre, the engraving of ochres and ostrich eggshells, the creation of decorative shell beads, and the production of multi-component tools that were hafted using compound adhesives (e.g., Henshilwood et al., 2002, 2004, 2014; Wadley et al., 2009; Texier et al., 2010, 2013; Vanhaeren et al., 2013). Other technological innovations also appear to originate at around the same time (e.g., Brown et al., 2009, 2012; Porraz et al., 2013).

In 2008, we proposed a chronology for the SB and HP, based on a systematic dating study of nine sites spanning a variety of climatic and ecological zones in southern Africa (Jacobs et al., 2008a). Age estimates for 54 sediment samples (and the associated artefacts) were obtained by optically stimulated luminescence (OSL) dating of individual grains of quartz sand, combined with statistical modelling to estimate the start and end dates of the SB and HP. Jacobs et al. (2013) re-examined the timing of the SB, based on the inclusion of 10 samples from Blombos Cave that had been analysed with the same instruments and experimental procedures as those used by Jacobs et al. (2008a).

To discern small differences in age – such as the possible existence of a short time gap between the end of the SB and start of the HP – it is important to estimate ages with maximum precision. Resolving such events and their timing is easily obscured by the chronological ‘haze’ arising from different dating methods being applied to different sites and, in the case of OSL dating, from the use of different instruments, calibration standards and procedures for sample preparation, measurement and data analysis (Jacobs and Roberts, 2008; Jacobs et al., 2008a). Our published SB/HP chronology is more precise than most other OSL chronologies, because the same instruments and procedures were used to measure all of the samples. This systematic approach
allowed several of the largest uncertainties attached to OSL ages to be removed when comparing our ages against each other, so that a more precise estimate could be made of the time span of the SB and HP industries and any gap between them (Jacobs and Roberts, 2008, 2009; Jacobs et al., 2008c, 2013). That is, the systematic error terms will be the same for each age, so it will not affect comparisons between estimates; only the random measurement errors, listed as the $\sigma_1$ values in Fig. 2 of Jacobs et al. (2008a), will contribute to the uncertainty in age. However, when comparing our SB/HP chronology with independent ages – including OSL ages obtained in other laboratories and on other instruments – all of the systematic error terms (listed as the $\sigma_2$ values in Fig. 2 of Jacobs et al., 2008a) should be included in the age uncertainty. For readers unfamiliar with the treatment of errors in OSL dating, and in statistics more generally, we refer them to Appendix A in Galbraith and Roberts (2012).

In this paper, we take the opportunity to update our original SB/HP data set using recent updates in dose rate conversion factors and error estimation of beta dose rates (Jacobs and Roberts, 2015), and to then re-run the age model for further comparison. In doing so, we confirm the high likelihood of a gap of several millennia between the SB and HP for those samples reported by us in 2008, and demonstrate that the published single-grain OSL chronologies for these two industries (Jacobs et al., 2008a, 2013) are robust relative to a range of alternative assumptions and modelled ages. We also discuss the two aspects of our original analysis that were commented on by Guérin et al. (2013): (1) the small size of the uncertainties attached to the beta-particle contribution to the environmental dose rate; and (2) the details of implementation of a method proposed by Jacobs et al. (2008b) to adjust the beta dose rate in specific circumstances. The latter adjustment affected a subset of the samples from three of the nine sites reported in Jacobs et al. (2008a). We briefly discuss the first issue below (see section 4), but have addressed it in full in Jacobs and Roberts (2015), together with a new method that is now used in our laboratory to calculate beta dose rate errors when using our GM-25-5 beta counters. Here, we focus on the concept and implementation of the beta dose rate adjustment method of Jacobs et al. (2008b) to explore the effect of different
approaches – including that advocated by Guérin et al. (2013) – on the ages calculated for the SB and HP samples. We also discuss the previously published data set from Pinnacle Point site 5-6 (PP5-6) that contains the HP, as well as evidence of a superficially similar microlithic industry earlier in time (Brown et al., 2012). The chronology of the latter is starting to fill the ‘gap’ between the SB and the HP, and, hence, directly address the question of the timing of the transition between the SB and the HP. This and other studies on lithic technology and chronology in this critical time period are currently stimulating a reassessment of the meaning of the HP (e.g., Mackay et al., 2011a,b, 2014; Porraz et al., 2013; Conard and Porraz, 2015) and the SB (e.g., Archer et al., 2015, 2016) as homogeneous entities at different time scales.

2. Background

2.1 OSL dating and field context

There are two parts to the OSL dating equation; Jacobs and Roberts (2007), Duller (2008), Wintle (2014) and Roberts et al. (2015) provide overviews of OSL dating for non-specialists. The numerator is called the equivalent dose (usually denoted as D_e and expressed in gray, Gy), which represents the radiation energy absorbed by a mineral grain since it was last exposed to sunlight (‘bleached’) or heated to a high temperature. The D_e is estimated from measurements of the OSL signal emitted by individual grains or by single aliquots composed of multiple grains. The burial history of an individual grain is the smallest meaningful unit of analysis in OSL dating, so Jacobs et al. (2008c) obtained D_e estimates for all of the samples from OSL measurements of individual sand-sized grains of quartz.

The denominator in the OSL age equation is the environmental dose rate (Gy per unit time), which consists of four separate contributions: the beta-particle and gamma-ray dose rates from materials surrounding the quartz grains by up to ~3 mm and ~30 cm, respectively, and lesser contributions from cosmic rays and alpha particles (the latter emitted by radioactive inclusions internal to the quartz grains). Many archaeological deposits are heterogeneous in their composition.
and in the spatial distribution of the organic and inorganic constituents (e.g., Goldberg et al., 2009; Miller et al., 2013; Karkanas et al., 2015). Changes also commonly occur over time due to a variety of diagenetic, pedogenic and taphonomic processes. As a result, quartz grains deposited at the same time and situated within a few mm or cm of each other can have very different beta dose rates, depending on their relative proximity to materials of high or low radioactivity. Quartz itself typically has a very low dose rate, as do many secondary carbonates formed in archaeological sediments after deposition. By contrast, minerals such as zircon and potassium feldspar have much higher dose rates, so a quartz grain juxtaposed by a zircon grain will receive a much higher beta dose rate than a contemporaneous quartz grain coated in calcium carbonate, for example.

The differences in beta dose rate to individual quartz grains is difficult to model in a way that is both simple to implement and faithful to the field context of the sample, because each sample poses a unique set of circumstances that no single model can address without knowing the boundary conditions specific to that sample. Ideally, what is required is knowledge of the beta dose rate to each dated grain in its original, undisturbed burial position. Combining spatially resolved measurements of the OSL signal from individual grains with measurements of the dose rate at the single-grain scale of analysis is currently a research priority (e.g., Martin et al., 2015a,b; Roberts et al., 2015). At the present time, however, it is not feasible to model the beta dose rate to individual grains for each and every sample.

In practice, sediment samples are disaggregated in the laboratory and quartz grains are separated for OSL measurements. The dose rate is determined for the bulk sample (i.e., the sample as collected from the field) and D_e values are obtained from single grains or multi-grain aliquots composed of purified quartz grains. Based on the extent of scatter among these D_e values, and the existence of any patterns in the D_e distribution, some estimate is made of the overall D_e value that corresponds most closely to the last bleaching or heating event. Several well-established statistical models are widely used in OSL dating for combining independent D_e values appropriately, with the choice of model depending on sample context, among other things (Galbraith and Roberts, 2012).
The $D_e$ so obtained is representative of the values for the group of grains in the sample that are thought to be related to the event of interest, and this population-estimate of $D_e$ is then usually divided by the bulk sample (‘sample-average’) dose rate to determine the OSL age. The latter is, thus, not a single-grain age in the strict sense, because the sample-average dose rate is used for all grains in the population, and not the dose rate specific to each grain.

For samples consisting of grains that were well bleached before burial and that have not been affected by post-depositional mixing, the central age model (CAM) is often used to determine the weighted mean $D_e$ for all of the grains (Galbraith et al., 1999); the CAM $D_e$ is divided by the sample-average dose rate to estimate the OSL age. When calculating the standard error on the $D_e$, the CAM includes an allowance for the spread in $D_e$ values over and above their measurement errors. This additional ‘overdispersion’ (OD) is a ubiquitous feature of $D_e$ distributions. For quartz grains that had been fully bleached at the time of deposition and thereafter remained undisturbed, the single-grain $D_e$ distributions are typically overdispersed by 10–20% or more (e.g., Galbraith et al., 2005; Arnold and Roberts, 2009).

The CAM may be sufficient for mineral grains buried in homogeneous radiation environments, because all of the grains will have experienced similar dose rates during the period of burial. It may even be appropriate for samples in which the dose rates to individual grains are markedly different, provided the spread in single-grain $D_e$ values is the same as the distribution of grain-specific dose rates. Suffice to say here that the CAM and sample-average dose rate can provide reliable estimates of age for many samples – indeed, Jacobs et al. (2008a) used this combination for many samples in their SB/HP study. But it is not the only approach that can be used, nor necessarily the best, and this is especially true for grains buried in heterogeneous radiation environments (Galbraith, 2015b).

2.2 Beta dose rate heterogeneity
At many archaeological sites, such as Sibudu Cave, the radiation environment is far from uniform at the scale of beta particles, because of spatial differences in the texture and composition of the deposit over distances of a few millimetres. Microscopic and spectroscopic studies illustrate the complex micromorphology of the Sibudu deposits, including the existence of a variety of microstratigraphic structures associated with burnt material and bedding (e.g., Schiegl et al., 2004; Schiegl and Conard, 2006; Goldberg et al., 2009; Wadley et al., 2011). Most of the deposit is anthropogenic in character, with lesser amounts of geogenic and biogenic material (Goldberg et al., 2009). Given the heterogeneity of the deposit at the scale of an individual sand grain, it is not unreasonable to expect that quartz grains deposited at the same time will experience different beta dose rates after burial and that these grain-to-grain differences will contribute to the spread in single-grain $D_e$ values.

Jacobs et al. (2008c) measured the $D_e$ values for individual grains of quartz extracted from 14 sediment samples collected from the post-HP, late MSA and final MSA deposits at Sibudu and reported three types of $D_e$ distribution: one consistent with a single dose population, another indicative of mixing between the MSA and overlying Iron Age deposits, and a third that they termed ‘scattered’. Most of the samples had scattered $D_e$ distributions, which were distinguished by a small population of grains with much smaller $D_e$ values than the majority (>$82\%$) of grains in the sample. Jacobs et al. (2008c) attributed the minor component to those grains experiencing smaller beta dose rates than the majority, as would be expected if the grains had been surrounded by $>2\ mm$ of radioactively inert material, which occur in the deposits (Schiegl et al., 2004; Schiegl and Conard, 2006; Goldberg et al., 2009). Jacobs et al. (2008b) assumed that dividing the $D_e$ values for these grains (i.e., the minor component of grains with smaller $D_e$ values) by the sample-average dose rate would give rise to age underestimates, and thus devised a procedure to correct for this bias. We note that such $D_e$ distributions and reasons for spread could be interpreted differently and may well change if more data are collected and further archaeological and geological studies are conducted at a site.
The 3-step procedure of Jacobs et al. (2008b) first consisted of fitting the finite mixture model (FMM) to the \( D_e \) distributions to estimate the \( D_e \) value of each population of grains and the relative proportion of grains in each fitted component (Roberts et al., 2000; Galbraith and Roberts, 2012). Then, the beta dose rate of the minor component was calculated by making a simple assumption: that it cannot be less than zero or more than the sample-average beta dose rate. A value mid-way between these two extremes was chosen as a first approximation, and assigned a standard deviation that spanned this full range at 2\( \sigma \). The \( D_e \) component that contains the majority of grains represents the population of interest, so the final step of the procedure was to estimate the beta dose rate for this major component. This was accomplished by making a pro rata adjustment to the sample-average beta dose rate based on the proportion of grains in the major and minor components. So, the sum of the beta dose rates for both components (taking into account the proportion of grains in each) equals the sample-average beta dose rate. For these 13 samples, the ages were thus obtained by dividing the \( D_e \) of the major component by the dose rate specific to that population of grains. (See Jacobs et al. [2008b: section 10.4] and Jacobs et al. [2011, their Table 4] for details of this approach and worked examples.)

At Sibudu, Jacobs et al. (2008b) proposed two reasons, field context and statistical precision, to use the FMM \( D_e \) values and adjusted beta dose rates (henceforth ‘aFMM ages’) to estimate the OSL ages for these samples, rather than the CAM \( D_e \) values and sample-average beta dose rates (henceforth ‘CAM ages’). The CAM ages calculated by Jacobs et al. (2008b) for all of their samples produced a conspicuous outlier among the post-HP samples: the age of SIB9 was much too young for its stratigraphic position, and the finely-resolved stratigraphy ruled out mixing as the cause. The ‘homogeneity test’ (Galbraith, 2003; Galbraith and Roberts, 2012) suggests that the CAM ages (and associated random errors) of all six post-HP Sibudu samples do not come from a single population \( (p = 0.0045) \). This appears to be due to the inclusion of the SIB9 age, which differs by more than 3\( \sigma \) \( (p = 0.0006) \) from the weighted mean CAM age for the other five post-HP samples. Sample SIB9 was collected from stratum P, “an orange–brown sandy-silt with white flecks
of gypsum” and small ashy lenses that probably represent the remains of a hearth (Schiegl and Conard, 2006, their Table 1). Given the observations of gypsum in this and other strata at Sibudu, there were prima facie contextual grounds for expecting that some quartz grains in SIB9 – and, by extension, other Sibudu samples with similarly scattered $D_e$ distributions – would have received much lower beta dose rates than the majority. This, of course, is a very simplified assumption and if we scrutinise the distribution of radioactivity on a microscale more closely we will doubtless see further variation due to a range of different factors. But given the constraints of their knowledge at the time, Jacobs et al. (2008b) applied their dose rate adjustment procedure to these samples and obtained aFMM ages without any significant stratigraphic anomalies. Accordingly, Jacobs et al. (2008b,c) proposed that the aFMM ages were more accurate than the CAM ages. Three of the 14 aFMM ages were also more precise than their CAM counterparts. This outcome can occur because the $D_e$ may be more precisely estimated using the FMM than the CAM, and this increase in precision may outweigh the decrease in precision associated with the adjusted beta dose rate. The balance between these different factors will differ between samples; see Galbraith (2015a,b) for related statistical discussion.

Except for SIB9, there are negligible differences between the aFMM and CAM ages for the Sibudu samples dated by Jacobs et al. (2008c). Figure 1a shows the CAM and aFMM ages measured for 13 post-HP, late and final MSA samples from Sibudu (the age of SIB12 was determined using the CAM only). Four systematic errors are not included for comparison, because they are the same for both (aFMM and CAM) data sets. These are the errors associated with the laboratory reference standard and the grain-size attenuation factors used to estimate the beta dose rates (see Appendix A in Jacobs and Roberts, 2015), the systematic error on the cosmic-ray dose rates, and the allowance of 2% for possible bias in the calibration of the laboratory beta source used for $D_e$ determination.

The same data are shown in Figure 1b as the difference between each pair of log age estimates plotted against the mean log age of the pair (after Oldham, 1962). We have used the
natural logarithms of the ages, following Galbraith and Roberts (2012), because the size of the standard errors increases in proportion to the ages. It is difficult to calculate the standard error of a difference in log age estimates from the individual standard errors in Fig. 1a because the aFMM and CAM ages are both obtained from the same data, so their estimation errors are correlated, but to an unknown degree. However, the standard deviation of the differences gives an estimate of this individual standard error, assuming it is approximately the same for each pair of ages. The grey band in Figure 1b shows 2.18 of these standard deviations above and below zero, where 2.18 is the 97.5 percentage point of the t-distribution with 12 degrees of freedom (equivalent to 1.96 for a normal distribution), so that any point within this band has a 95% confidence interval for the true difference that includes zero. The ratios of the aFMM/CAM ages are consistent with unity for all samples, except for SIB9 which lies away from the 1:1 line (Fig. 1a) and outside the grey band (Fig. 1b).

No model should be applied ‘blind’ to any sample, without contextual support for its use (Galbraith et al., 1999, 2005; Galbraith and Roberts, 2012). It is for this reason that we carry out intensive fieldwork alongside archaeologists, geoarchaeologists and, where possible, with those doing micromorphology and related studies, to understand better not only the site formation processes at play at each site, but also the finer scale differences between the sedimentary and archaeological characteristics of different units and layers. In the absence of a specific reason to use the FMM or some other model (such as the minimum or maximum age models; Galbraith and Roberts, 2012), and taking into account knowledge of site and sample context, the CAM Dc divided by the sample-average dose rate may give the most reliable estimate of age. Such was the case with some or all of the OSL samples from Apollo 11, Klasies River, Melikane, Ntloana Tsoana, Rose Cottage Cave, Sehonghong and Sibudu dated by Jacobs et al. (2008a), and for seven (of 10) samples associated with the SB at Blombos (Jacobs et al., 2013). For samples with mixed and/or scattered Dc distributions, there is nonetheless merit in also exploring other options (Galbraith, 2015a,b).
3. Methods

3.1 Dose rate determination

In this study, we have recalculated the dose rates for the SB and HP samples, incorporating:
(1) the updates to dose rate conversion factors reported in 2011 (Guérin et al., 2011); (2) the error estimation procedure for beta dose rates published by Jacobs and Roberts (2015); and (3) the uncertainty associated with the water content of each sample, to which we assigned a minimum relative standard error of ± 25%. These changes affect our estimates of the beta dose rate and, in some cases, also the gamma dose rate (with, in general, a slight increase in dose rate and, hence, decrease in age). For some of the samples, we also re-measured the beta dose rates using the GM-25-5 beta counters and used the average of the new and old measurements, where consistent. We found an inconsistent result for only one of the samples (DRS6) and used the new results (which we confirmed by replication) to calculate the revised age reported in Jacobs and Roberts (2015).

Our revised method of error estimation for beta dose rates was formulated with statistician Rex Galbraith and is described in full in Appendix A of Jacobs and Roberts (2015), together with a worked example for one of the samples from Diepkloof Rockshelter. The procedure is based on an analysis of variance between replicate measurements that is applicable to all samples measured using our GM-25-5 beta counters; we have always counted at least three replicates of each sample to check for reproducibility. Our error estimation procedure takes into account the counting errors associated with day-to-day fluctuations in the behaviour of the equipment, the differences in counting efficiencies of the five detectors, the reproducibility among the three replicates measured for each sample, and the counting statistics of the replicates, the laboratory standard and the background blank. The total error on the beta dose rate also includes systematic errors for the measurement imprecision of the laboratory reference standard (~1.8%), the uncertainties associated with the dose rate conversion factors (assumed to be ~2%) and a ~2% uncertainty on the correction required for grain-size attenuation of beta particles (Murray and Olley, 2002). There may well be
other sources of uncertainty that can be included or their estimation improved. The purpose of publishing our method is to show exactly what we include and how we calculate the errors associated with our beta dose rates. We encourage others to do likewise for transparency and to enable further improvements to be made to error estimation.

3.2 Chronological modelling

We have fitted the same statistical model as that used by Jacobs et al. (2008a, 2013) to the revised ages to estimate new start and end ages for the SB and HP, the duration of the HP period, and the length of the gap between the SB and HP. We tested three different scenarios to examine the sensitivity of the modelled ages to a range of alternative assumptions:

Estimate A uses all the updated ages for all of the samples, with the aFMM ages used instead of the CAM ages for those samples where the scatter in D_e values is thought to be due to dose rate heterogeneity; these are indicated with asterisks in Supplementary Online Material (SOM) Table S1. For sample DRS6, the FMM age is used in preference to the CAM age (see Jacobs and Roberts, 2015 for discussion of this sample), but the model outcome is not sensitive to this choice. The individual ages are listed in SOM Table S1. This estimate represents the most direct comparison between the revised ages and the ages reported by Jacobs et al. (2008a, 2013).

Estimate B uses the same samples as A, but CAM ages with sample-average beta dose rates are used for the 10 samples (three from Blombos and seven from Sibudu) for which aFMM ages are used in A. The individual ages are listed in Table 1 and SOM Table S1; these now represent our preferred age estimates. This estimate is also the analysis closest to that favoured by Guérin et al. (2013).

Estimate C is the same as B, except that three of the Diepkloof samples (DRS11, 13 and 14) are omitted due to the currently inexplicable differences in age for the middle part of this MSA sequence (Tribolo et al., 2013; Feathers, 2015; Jacobs and Roberts, 2015).
As in Jacobs et al. (2008a, 2013), we calculated standard errors $\sigma_1$ and $\sigma_2$ (Table 1 and SOM Table S1). These exclude and include, respectively, systematic errors that are common to each of our age estimates and, hence, do not affect comparisons among estimates. The $\sigma_1$ values exclude the four systematic errors noted earlier (i.e., those associated with the laboratory beta source and GM-25-5 beta counter reference standard, the grain-size attenuation factors and the cosmic-ray dose rates). As the gamma dose rates were measured using a variety of methods at the different sites, the associated uncertainties are included in $\sigma_1$. Uncertainties in water content are also treated as random errors and, thus, included in $\sigma_1$.

4. Results and preliminary discussion

In this section, we present the results of these new analyses, starting with an assessment of the accuracy and precision of the beta dose rates estimated using our GM-25-5 beta counters and the procedures described in Jacobs and Roberts (2015). This is followed by a discussion of the merits of this approach in the light of the remarks made by Guérin et al. (2013). We then present the results of applying the statistical model to Estimates A, B and C to ascertain the chronology of the SB and HP industries, and discuss the strength of the evidence for a time gap between them.

4.1 Beta-counting accuracy and precision

The accuracy of our GM-25-5 beta counters and sample preparation and presentation procedures has been tested and confirmed in several ways, using the samples from Diepkloof (see Jacobs and Roberts, 2015). We have measured samples as both loose powder and as powder pressed into pellets that are bound together with a few drops of polyvinyl acetate (PVA), with and without delay times between preparation and measurement (to allow for equilibration of the $^{238}$U and $^{232}$Th decay chains). For the 13 samples from Diepkloof, the beta dose rates obtained for the loose sediment (no delay time) are, on average, ~2.4% higher than the corresponding results for the pressed pellets (with a one month delay time; ratio of 1.024 ± 0.017), but the results are randomly
spread around a ratio of unity and are not systematically biased (Jacobs and Roberts, 2015: Fig. 8a). We have also measured the U, Th and K contents of the same samples using other techniques and obtained consistent results. For the Diepkloof samples, we compared the beta dose rates obtained for the pressed pellets (with a one month delay time) using our GM-25-5 beta counters with measurements of U, Th and K from: (1) inductively-coupled plasma mass spectrometry (ICP-MS) and optical emission spectroscopy (ICP-OES); and (2) thick-source alpha counting and X-ray fluorescence (XRF) spectroscopy. Average ratios consistent with unity (0.996 ± 0.015 and 0.997 ± 0.015, respectively) were obtained for these sets of data, again with individual ratios spread randomly around the 1:1 line (Jacobs and Roberts, 2015, their Fig. 8b,c). The ICP-MS, ICP-OES and XRF measurements were made in commercial laboratories, so they provide independent validation of the accuracy of our GM-25-5 beta-counting measurements.

Jacobs et al. (2008a,b,c, 2011, 2012, 2013) also used GM-25-5 beta counting to determine the beta dose rates for the majority of their samples. Guérin et al. (2013) considered the reported errors to be too small and instead suggested that a beta dose rate error of ~7% may be more appropriate, based on their analysis of measurements made by Ankjærgaard and Murray (2007). We note that the beta dose rates reported in six of the seven papers referred to in SOM Table S1 of Guérin et al. (2013) actually have errors that are similar to a value of 7%: 5.2–7.4% (El Harhourha 2, Morocco), 5.6–8.3% (El Mnasra, Morocco), 5.1–7.0% (Contrebandiers, Morocco), 5.1–6.6% (Mumba, Tanzania), 5.8–7.3% (Blombos, South Africa), 5.2–8.5% (Pinnacle Point site 13B, South Africa) and 2.6–9.9% for the post-HP, late and final MSA samples from Sibudu, South Africa. Guérin et al. (2013) do not acknowledge this fact, but imply instead that the beta dose rate errors associated with all of the ages are too small. Only in Jacobs et al. (2008a) are the beta dose rate errors for the SB and HP samples much smaller than 7%, ranging between 1.1% and 3.8%, and we concur with Guérin et al. (2013) that these errors are underestimated. We re-examined the beta dose rate errors published in Jacobs et al. (2008a) and found that we had not fully propagated all errors for these samples. This oversight is rectified in the current study. Our revised estimates of the total
error (random plus systematic) on the beta dose rates range from 4.9 to 7.8% for the SB and HP samples in this study; the random errors alone range between 2.0 and 6.8%, which are larger than those reported by Jacobs et al. (2008a). Since we have undertaken an evaluation of the specific sources of uncertainties in our beta counting measurements (see Jacobs and Roberts, 2015), and because conclusions obtained from experiments using a particular set of settings or circumstances are strictly valid only for those specific conditions (Galbraith and Roberts, 2012), we use our revised estimates in preference to the 7% value suggested by Guérin et al. (2013).

4.2 Updating the SB–HP chronology

The updated dose rate data and ages are presented in Table 1; these ages supersede all previously reported ages for these samples. The modelled estimates for the start and end ages of the SB and HP are listed in Table 2, while those for the length of the HP period and the gap between the SB and HP are given in Table 3; the original results of Jacobs et al. (2008a) are also provided in both tables, for comparison. The confidence intervals for the start and end ages of the SB and HP include all random and systematic errors (i.e., \( \sigma_2 \)), so these ages can be compared directly with other independent chronologies. To estimate the confidence intervals for the durations in Table 3, all systematic errors have been omitted (i.e., \( \sigma_1 \)).

4.2.1 Start and end ages for the SB and HP and the length of the HP

Start and end ages for both the SB and HP are very similar for Estimates A, B and C (Table 2) and agree closely with the original estimates (Jacobs et al., 2008a). The length of the HP has increased under all three scenarios, from the original estimate of 5.3 ka to 7.7 ka (Estimate A), 8.4 ka (Estimate B) and 8.6 ka (Estimate C) (Table 3). Based on our preferred ages (Estimate B), the insensitivity of the modelled estimates to these different scenarios, and the numerous updates and new measurements we have made, we remain confident in our age estimates for the HP and SB (Tables 1 and 2). We also note the good agreement between our ages for the SB and HP and those obtained using other methods and by
other laboratories (e.g., Miller et al., 1999; Grün et al., 2001; Vogel, 2001; Tribolo et al., 2005, 2006; Valladas et al., 2005; Cochrane, 2009; Högberg and Larsson, 2011; Henshilwood et al., 2014; Murray-Wallace et al., 2015).

We caution against conflating these small changes in estimated ages for the SB and HP with issues associated with the chronology for Diepkloof. In Estimate C, we deliberately excluded three samples collected from the middle part of the Diepkloof sequence, because of the unaccountable differences currently observed in three separate studies. We note, however, that our age estimates for the SB and HP at Diepkloof are consistent with those obtained at other sites, and that their exclusion from the model has little effect on the calculated start and end ages for the HP and SB or on the estimated length of the HP. We show below (see crosses in Fig. 2) that the chronological differences in this part of the Diepkloof sequence are not due to the application of the beta dose rate adjustment method (as was suggested by Guérin et al., 2013), but are likely related to the measurement of potassium (\(^{40}\)K) in these deposits (Jacobs and Roberts, 2015). Other aspects of the discrepant chronology for the middle part of the Diepkloof sequence have previously been addressed by Jacobs and Roberts (2015).

4.2.2 The gap between the SB and the HP Jacobs et al. (2008a) identified a gap between the SB and HP (\(p = 0.008\)) and estimated its length to be 6.7 ka, with a 95% confidence interval of 2.7–9.3 ka (Table 3). The gap durations for Estimates A, B and C are also shown in Table 3, together with the 95% confidence intervals and \(p\)-values. The latter are based on the likelihood ratio test of the null hypothesis that there is no gap between the end of the SB and the start of the HP. The length of the gap for Estimates A, B and C are all slightly shorter than estimated originally. The 95% confidence interval for Estimate A includes zero, so those data are consistent (at the conventional 5% significance level) with the null hypothesis that there is no gap; and also, of course, with the hypothesis that there is a gap of up to 8.3 ka. The \(p\)-value of 0.059 (i.e., slightly greater than 0.05) suggests only weak evidence in favour of a gap. Estimates B and C show clear statistical evidence
in favour of a gap of ~5 ka, formally supported by $p$-values of 0.014 and 0.033, respectively. As noted above, our preferred ages are used in Estimate B, which is also the scenario closest to that favoured by Guérin et al. (2013).

The general conclusions of Jacobs et al. (2008a), thus, remain true under two of the three scenarios (Estimates B and C), both with regards to the general timing of the start and end of the SB and the HP, and the likely gap between them of several millennia. Our results are not an artefact of how we analytically process or statistically model our data.

5. Discussion

In this section, we discuss the main concerns raised by Guérin et al. (2013) about the beta dose rate adjustment procedure used for a subset of the samples dated by Jacobs et al. (2008a). We have demonstrated above that the modelled chronology for the SB and HP is not sensitive to the inclusion or exclusion of the ages obtained using this procedure, but it is important nonetheless to address some of the statements made by Guérin et al. (2013). We conclude by reassessing the existence of a gap between the SB and HP in view of archaeological developments that have taken place since our original study (Jacobs et al., 2008a), and we highlight the importance of temporal resolution (scales of analysis) when interpreting OSL (and other) chronologies in the context of stratigraphic and archaeological changes.

We note that almost all of the examples in Guérin et al. (2013), including the only figure, are unrelated to the timing of the SB and HP or of the transition between them – and, hence, to the title of their commentary. Instead, they concentrate on the post-HP, late and final MSA at Sibudu and on four sites in East and North Africa: Mumba Rockshelter in Tanzania (Gliganic et al., 2012a,b) and three caves in Morocco (La Grotte des Contrebandiers, El Harhoura 2 and El Mnasra; Jacobs et al., 2011, 2012). Interested readers are encouraged to read our original papers, and the supplementary information sections, for detailed accounts of the procedures used and the data so obtained in these studies.
5.1 Beta dose rate adjustments

5.1.1 “All models are wrong but some are useful” This is a famous quote by the statistician George Box (Box, 1979: 202) and reflects a truism: models are a simplification of the real world to a greater or lesser extent, but some models can provide useful approximations and illuminating insights. In the context of OSL dating, a dose rate model is only useful if it bears some (close) relation to the depositional environment of the samples of interest and if the sample was collected and measured in an appropriate manner.

The beta dose rate adjustment method of Jacobs et al. (2008b) was predicated on two observations: (1) the scattered $D_e$ distributions of some of the Sibudu samples, which could not be explained by incomplete bleaching of the grains before burial or mixing after burial; and (2) the presence of low-radioactivity minerals (gypsum and calcite) in the deposit. This model involved some simplifying assumptions, such as estimating the number of single-grain populations in any particular $D_e$ distribution to choose between applications of the CAM or (a)FMM. On the basis of two equations, Guérin et al. (2013: 315) claimed that the model proposed by Jacobs et al. (2008b) was “flawed and unnecessary” and would “introduce additional uncertainties”, concluding that the average $D_e$ and dose rate would give the best age estimate for well-bleached samples with beta dose rate heterogeneity.

Galbraith (2015a) explained in detail how the assumption underpinning the two equations of Guérin et al. (2013) is not tenable and not true, in general, for measured quantities such as $D_e$ estimates, and showed that their conclusion does not follow from their two equations – whether or not they hold in practice. We agree with Galbraith (2015a) that the initial assumption underlying the equations of Guérin et al. (2013) – that the only source of dispersion in single-grain doses is the dose rate – was incorrectly attributed to Jacobs et al. (2008b). On the contrary, Jacobs et al. (2008b) listed numerous factors that contribute to the dispersion in single-grain $D_e$ values, including intrinsic factors related to the OSL characteristics and behaviour of individual grains, and extrinsic
factors related to the depositional and post-depositional histories of the sample. They referred to differences in the beta dose received by individual grains in their burial environment and the inadequate bleaching of some grains before burial, as well as the potential for post-depositional movement of grains, and also stated that some dispersion in single-grain D_e values may arise from non-identical field and laboratory conditions and natural variability (following Galbraith et al., 2005).

5.1.2 Age model selection Some of the other statements of Guérin et al. (2013: 314) are also unfounded. They claimed that Jacobs et al. (2008a) choose between the CAM and FMM on the basis of a “potentially subjective 20% limit” for D_e OD. We have never used an OD value of 20% as the sole criterion for choosing between the CAM and any other age model, but we acknowledge that our wording may have been ambiguous if read in isolation from any of our other papers. Instead, we routinely take several steps to assess the possible cause(s) of the observed distribution of D_e values, as we have described in previous publications (e.g., Jacobs and Roberts, 2007, 2015; Jacobs et al., 2008c, 2011, 2013). First, we calculate the OD value for each sample using the CAM. Overdispersion values of up to 20% or more have been frequently observed for many quartz samples from around the world (e.g., Olley et al., 2004; Arnold and Roberts, 2009). So, when OD values of ~20% or greater are obtained, we give special consideration to any other features of the D_e distribution, but do not automatically apply the FMM or minimum age model (MAM). Second, we make a visual appraisal of the distribution pattern of D_e values in a radial plot (Galbraith, 1988; Galbraith and Roberts, 2012) to assess whether it comprises a single population of values or multiple, discrete components, or if it displays some other feature(s) of interest. Third, we use our knowledge of the site and sample contexts to ask questions about the known depositional and post-depositional processes that could give rise to the observed D_e distribution. These different threads of evidence are then brought together to develop a logical scenario to explain the D_e distribution for any particular sample and to help us select the most appropriate statistical model for age estimation. We note that different practitioners may interpret the same datasets differently. No interpretation is
flawless, but must be explained so that the readers of a paper can make up their own minds and to enable data sets to be re-examined in the light of new observations and associated archaeological and geological information.

When the FMM is selected, we always use statistical diagnostics, namely the Bayes Information Criterion (BIC) and the maximum log likelihood (llik), based on the recommendations of Galbraith (2005). This permits us to decide: (1) whether a $D_e$ distribution is made up of more than one discrete $D_e$ component; and (2) if more than one discrete component can be statistically supported, then the BIC and llik are used to resolve, for a range of tested alternatives, the number of $D_e$ components and OD value at which the model is optimised. Jacobs et al. (2008c) reported OD values of 7% for grains that were bleached and then given a uniform beta dose in the laboratory (for a dose recovery test on 1000 single grains) and 17% for grains from a natural sample (SIB12) that consisted of a single $D_e$ component. These represent ‘best case’ outcomes for laboratory- and field-dosed samples composed of well-bleached grains. Jacobs et al. (2008c) fitted the FMM to the $D_e$ distributions of the other Sibudu samples using a conservative range of OD values (10–20%), and noted that very similar $D_e$ estimates were obtained for all OD values within this range. We did not fit the FMM with OD values of >20% because optimised values of llik and BIC were obtained for the model at values within the 10–20% range. The OD value at which the model was optimised was reported in Jacobs et al. (2008b,c), alongside a worked example, and in Jacobs et al. (2008a) for each sample where the FMM was used. A wider range of values may well be necessary for samples from other sites, but this needs to be assessed on a case-by-case basis. Therefore, contrary to the above claim of Guérin et al. (2013), we maintain that our implementation of the FMM is statistically robust and suitable for our samples.

Guérin et al. (2013: 314) also remarked that the intrusive grains in the three mixed Iron Age/MSA samples were discarded “without explicit justification”. This is not true. Jacobs et al. (2008b,c) stated that those grains were discarded because their $D_e$ values were too small to be explained by low beta dose rates and yielded ages consistent with the Iron Age. Jacobs et al.
(2008b) also provided additional evidence of post-depositional mixing for these three samples, based on the extent of $D_e$ overdispersion among aliquots composed of ~50 grains. Clearly, it would be nonsensical to retain Iron Age grains in data sets used to determine the burial ages of MSA deposits.

5.1.3 Dose rate simulations The beta dose rate adjustment method of Jacobs et al. (2008b) was also challenged by Guérin et al. (2013) on the basis that bimodal $D_e$ distributions were not generated in the numerical (Monte Carlo) simulations of Nathan et al. (2003). However, the imposed constraints of these simulations make them poor approximations of many real archaeological deposits – in particular, the assumed random distribution of radioactively inert or comparatively high radioactivity spheres in an otherwise homogeneous matrix. No allowance was made for spatial heterogeneity in beta dose rate associated with common features such as hearths, organic matter and the precipitation of secondary minerals around grains and in the voids between them. Non-randomly distributed features at the microscopic scale, such as laminated layers, have been comprehensively documented at Sibudu (Goldberg et al., 2009; Wadley et al., 2011), but they were not incorporated in the simulations. Indeed, Nathan et al. (2003: 312) recognised that while they were unable to simulate bimodal distributions, grains surrounded completely by carbonate material “may be an exception to this.” Likewise, Cunningham et al. (2012) conducted an experimental and Monte Carlo simulation of beta dose rate heterogeneity among laboratory-dosed quartz grains, and found that the simulation underestimated the spread in measured single-grain $D_e$ values. Various possible causes of the discrepancy were considered, including those that “lie in the parts of the real world that are not included in the model” (Cunningham et al., 2012: 1067).

Our opinion is that generalised models of beta microdosimetry are not likely to prove useful (sensu Box, 1979) in the absence of specific information about the burial environment. That is, there is no a priori basis to suppose that they will provide a realistic approximation of the beta dose rate for any particular sample. A ‘useful model’ requires knowledge of the field context of the
samples under investigation to ensure that any spatial variability in the beta dose rate is taken into account— that is, the model must allow for all things that are known to be true, rather than be developed in isolation from physical reality. There is merit in measuring the spatial distribution of radioactivity in the samples of interest and then constructing a numerical model based on those data. The beta dose rate adjustment method of Jacobs et al. (2008b) attempted to incorporate knowledge of field context for samples that are known, or are thought, to contain a proportion of low dose rate material. More sophisticated models, coupled with Monte Carlo simulations (e.g., Martin et al., 2015a,b), will offer advantages if the field context supports their application and if the necessary data are available to implement them in practice. Accomplishing the latter will be non-trivial, however, if the burial environment is complex (see Roberts et al., 2015).

The fact that the Sibudu samples with scattered D_e distributions could be optimally fitted by only two FMM components is, of course, a simplification of reality. For all of the non-single D_e distributions, Jacobs et al. (2008c: 1800) noted the common occurrence of “a small number of grains with intermediate values”, but the vast majority of grains in the scattered D_e distributions were captured by just two components, thus representing a useful approximation. Monte Carlo simulations are also an approximation of reality and all relevant parameters have not yet been included in simulations of natural samples. This shortcoming should be rectified in future simulations, so that models are able to explain the diversity of D_e distributions observed in real samples. Likewise, interpretation of complex D_e distributions should ideally be accompanied by supporting micromorphological observations and chemical analyses of the related sediments (e.g., Jankowski et al., 2015).

5.1.4 Age comparisons Guérin et al. (2013) contend that the aFMM ages of Jacobs et al. (2008b) are less accurate than the CAM ages and, on p. 315, they say that the CAM and aFMM ages are significantly different “for 13 out of 14 samples at Sibudu.” In fact, the numbers are the reverse: only one sample (SIB9) out of the 14 has an aFMM age that differs significantly from its CAM age.
(see Fig. 1a,b). Of course, the CAM age of SIB9 identified as an outlier by Jacobs et al. (2008b) provided the catalyst for their beta dose rate adjustment method. None of these samples is from the HP or SB, so any differences between the two methods will have very little effect on the calculated chronology provided in Jacobs et al. (2008a). The same applies to the North African data set of Jacobs et al. (2012) that Guérin et al. (2013) also examined. Thirty-two OSL samples were dated from two sites in Morocco (El Harhoura 2 and El Mnasra) and 20 of these samples had scattered $D_e$ distributions. As noted earlier, it is difficult to calculate a valid standard error for the difference or ratio of the aFMM and CAM ages from their individual standard errors, because of the unknown correlation between their estimation errors. A simple valid comparison can be made using a standard paired t-test and confidence interval applied to the logs of the ratios. For these 20 samples, the mean of the log ratios is 0.021, corresponding to an aFMM/CAM ratio of 1.022 (95% confidence interval from 1.008 to 1.035). So, on average, the aFMM ages are about 2% higher than the CAM ages. The individual log ratios vary between −0.052 and +0.064 (equivalent to ratios of between 0.949 and 1.066), with a standard deviation of 0.028. The latter figure can be interpreted as an approximate average relative standard error of 2.8% applicable to a single aFMM/CAM age ratio. These differences are small, given that the relative standard errors of the aFMM and CAM ages are typically ~5% and vary between 3.9% and 8.7% (excluding the systematic errors).

In Figure 2, we scrutinise more closely those samples relevant to the title of the comment by Guérin et al. (2013) – that is, samples associated with the HP and the SB. The beta dose rate adjustment was not applied to all samples; it was used to determine the OSL ages for eight of the 13 samples from Diepkloof (shown as crosses in Fig. 2; Jacobs et al., 2008a), seven of eight additional samples from Sibudu (shown as filled circles in Fig. 2; Jacobs et al., 2008a), and three of 10 samples from Blombos (shown as open triangles in Fig. 2; Jacobs et al., 2013). The open circles in Figure 2 are for the same Sibudu post-HP, late and final MSA ages shown in Fig. 1a,b. Figure 2 plots the difference in log ages against the mean log age for each sample ($n = 31$); scales of ages and ratios are also shown. We calculated an overall mean difference in log ages for all 31 samples
of 0.0233, with a standard deviation of 0.0412. This mean difference converts to an aFMM/CAM age ratio of 1.024 (95% confidence interval from 1.008 to 1.039), so the aFMM ages are, on average, ~2.4% larger than the CAM ages. The standard deviation of 0.0412 is an estimate of the standard error of an individual difference and the grey band in Figure 2 is calculated as $2.04 \times 0.0412 (= 0.084)$ around zero, where 2.04 is the 97.5 percentage point of the t-distribution with 30 degrees of freedom (corresponding to 1.96 for a normal distribution). A point falling within this band is individually consistent (within estimation error) with zero difference (age ratio = 1). The band captures all of the points, except SIB9. The degree of concordance between each pair of log age estimates is unsurprising, as the aFMM ages are based on the $D_c$ of the major component, which contains >83% of the grains for all but two (DRS6 and DRS9) of the 31 samples.

Guérin et al. (2013) suggest that the adjustment procedure of Jacobs et al. (2008b) could be improved by including an ‘optimisation’ feature that forces the ages of the two FMM components to be equivalent. This approach is sound in principle, but there are statistical arguments for and against the use of optimisation models (Galbraith, 2015b). If implemented, optimisation would need to account for the differing sizes of the standard errors of the two $D_c$ components (often the minor component has a relative error several-fold larger than that of the major component – see, for example, Jacobs et al., 2012, their Tables 1 and 2) and the errors in the dose rate that are common to both components. Guérin et al. (2013) iterated their calculations until the age estimates for the two components were indistinguishable, but they do not say if they took the respective errors into account.

The effect of optimising the dose rate model is, anyway, very small: for the samples from El Harhoura 2 and El Mnasra, Guérin et al. (2013) obtained FMM ages that were, on average, ~1.5% lower after optimisation, which would draw them even closer to the CAM ages. We obtain a very similar result by adjusting the assumed beta dose rate of the minor $D_c$ component until the mean ages of both FMM components match exactly. The mean of the log ratios is 0.013, corresponding to a ratio of 1.013 (95% confidence interval from 1.010 to 1.016), so the pre-optimisation ages are, on
average, higher than the post-optimisation ages by ~1.3%. The individual log ratios are all positive and vary between 0.005 and 0.031 (equivalent to ratios of between 1.005 and 1.031), with a standard deviation of 0.006. Again, these differences are very small compared with a typical relative standard error of a single aFMM age estimate of between 4.7% and 8.7% (excluding the systematic errors). It is also worth noting that not optimising the dose rate model is not necessarily invalid – that is, you can still obtain a valid estimate from the dominant component alone (Galbraith, 2015b).

The same conclusions hold true for other samples in which the major \( D_e \) component contains the vast majority of grains. Galbraith (2015b), however, elaborates on the statistical issues relating to dose rate heterogeneity, taking issue with the beta dose rate adjustment method of Jacobs et al. (2008b) and with several of the claims made on p. 315 of Guérin et al. (2013). More mathematically efficient and statistically rigorous approaches should, therefore, be used to refine the beta dose rate adjustment method of Jacobs et al. (2008b), and we welcome further efforts to this end. We acknowledge that our beta dose rate adjustment method has several shortcomings, so until direct measurements of beta dose rate at the single-grain scale, combined with improved models, are available to deal with issues of dose rate heterogeneity, our preferred approach at present is to use the CAM estimate of \( D_e \) and the average beta dose rate, rather than the aFMM estimate of \( D_e \) and the adjusted beta dose rate. What ultimately matters to archaeologists is to what extent the ages depend significantly on the choice of model – and we have demonstrated in this study that the answer is clearly ‘very little’ for these samples.

It is important to note that Figure 2 does not include two-thirds of the OSL ages presented in Jacobs et al. (2008a, 2013), because no beta dose rate adjustment was applied to 36 of the 54 samples – CAM or FMM ages were reported instead. The latter procedure was applied when two \( D_e \) components were identified with the FMM, but the minor dose component contained \( D_e \) values that were too small to be explained by beta dose heterogeneity (i.e., they represent younger intrusive grains), consisted of \( D_e \) values larger than the weighted mean \( D_e \) value of the major component.
(e.g., DRS10), or where our field observations did not support an interpretation consistent with beta microdosimetry as the main cause of spread in $D_e$ values (e.g., KKH samples). Based on the underlying assumptions of the adjustment method of Jacobs et al. (2008b), the beta dose rate correction can only be applied to a minor component composed of $D_e$ values smaller than the main component. We interpret minor components with larger $D_e$ values as representing intrusive older grains, but acknowledge that they may instead, or as well, represent a small proportion of grains exposed to higher beta dose rates than the sample average. Guérin et al. (2013: 316) err, therefore, in suggesting that Jacobs et al. (2008a) applied the beta dose rate adjustment model to Diepkloof sample DRS10, “resulting in an age significantly smaller than would otherwise be obtained.” The adjustment method was not applied to this sample. Guérin et al. (2013: 315–316 and Fig. 1 caption) also presume that we incorrectly applied this method to Sibudu samples SIB2 and SIB7, by calculating the ages from the $D_e$ values for the minor component; in fact, the ages of both samples were calculated using the procedure described by Jacobs et al. (2008b).

5.2 Revisiting the gap

Returning to the question of a time gap between the end of the SB and the start of the HP, Jacobs et al. (2008a: 19 in supplementary material) explicitly noted the possibility “that there was activity in this period and that samples from this time were simply not obtained”. So, have any further studies in the intervening decade started to close or fill this gap? Work by one of us (Z.J.) at Pinnacle Point site 5-6 (PP5-6) on the south coast of South Africa shows clear evidence for ‘filling the gap’. PP5-6 is a site that has a long, high-resolution depositional sequence that records major changes in stone artefact technology that have been argued to represent significant shifts in behaviour (Brown et al., 2009, 2012) – including features described by some as akin to Early and Intermediate HP at Diepkloof (e.g., Porraz et al., 2013; Conard and Porraz, 2015). Porraz et al. (2013) informally relate the SGS, OBS1 and SADBS stratigraphic aggregates at PP5-6 to, respectively, the Intermediate HP, MSA Jack and Early HP at Diepkloof. Site PP5-6 is exceptional
because it has sufficient resolution to allow natural sedimentary processes to be related directly to 
behavioural and cultural changes through a critical period of the MSA, aiding the unpacking of finer 
details of what constitutes and drives the different anthropogenic markers (Karkanas et al., 2015). 
The different types of sedimentary deposit preserved at PP5-6 also provide an ideal test of how well 
OSL dating can resolve the timing between clearly different stratigraphic aggregates, sub-
aggregates and cultural phases.

Eleven ages have been published previously for this part of the PP5-6 sequence, which 
contains the HP and earlier microlithic industries of the SADBS (Brown et al., 2009, 2012). We 
have updated those ages here, so that they are consistent with the other ages presented in this study 
(Table 1). The updated ages are shown in Figure 3, together with the end date of the SB and start 
date of the HP (stippled lines) and the gap between them based on Estimate B (grey band). The 
different lithic phases follow a clear chronological succession, ending with the ‘classic’ HP in 
DBCS and OBS2, which has ages consistent with those for the HP in this study. The sequence 
includes the earlier phases of microlithic technology, including the SADBS with ages of between 66 
± 3 and 73 ± 3 ka (at 1σ), thus filling the gap between the SB and the HP. In the absence of the SB 
at PP5-6, the SADBS microlithic assemblage overlaps in age with the SB at the nearby site of 
Blombos, which can also be reconciled with the cultural sequence and overlap of bifacial points and 
HP backed artefacts at Diepkloof (Porraz et al., 2013). Filling the gap in this way reflects the greater 
resolution provided by site PP5-6, in particular, and by the systematic application of single-grain 
OSL dating to improve our ability to compare and integrate different data sets; it is not due to a 
change in methodology or our analytical or modelling procedures. Eliminating the gap between the 
SB and HP obviously has implications for how we define the HP and its duration.

We expect that fine-grained OSL chronologies can be obtained for different cultural phases 
at sites that are well-resolved stratigraphically. Taphonomic changes, however, commonly result in 
compression of the stratigraphy and time-averaging of anthropogenic contents, so an average 
depositional age for the sediments is the best that can be achieved using OSL dating. This restricts
the level of detail that can be discerned at sites that are less well preserved and resolved in space and time. Accordingly, our emphasis is now on dating sites such as PP5-6 at finer stratigraphic resolution, so that we can utilise the strengths of statistical models, such as the Bayesian models used widely in \(^{14}\)C dating (e.g., Bronk Ramsey, 2009), to improve the precision of chronological sequences and interrogate cultural processes in more detail. Sites such as PP5-6 and Blombos will allow the HP and SB to be examined at higher temporal resolution, while other sites (e.g., Diepkoof and Sibudu) can provide insights into the chronological relationships of these and other lithic industries – and any spatial patterns – across southern Africa.

6. Conclusions

The modelled start and end ages for the SB and HP, the duration of the HP, and the length of the ‘gap’ between the SB and the HP all have confidence intervals that span several millennia. Owing to the number of separate measurements made to obtain estimates of the \(D_e\) and dose rate to calculate an OSL age, and the other uncertainties inherent to the method, it is not practicable to reduce the total error associated with individual or modelled ages to less than ~3%. At 60 ka, this represents 1800 years at 1\(\sigma\) or about 62 human generations, assuming that each generation is ~29 years (Fenner, 2005). This is much coarser than the decadal or centennial level resolution (corresponding to 1–3 generations) required to truly understand the operation of cultural processes, such as information flow, the impact of demographic changes on resident populations, or the effects of short-term environmental variations on resource use. It is fundamentally important, therefore, to be aware of these different scales of analysis when interpreting all types of data, including technological and stratigraphic changes through time.

The collection of more detailed data at higher stratigraphic resolution (e.g., Goldberg et al., 2009; Miller et al., 2013; Karkanas et al., 2015) and more comprehensive technological analyses are leading to changes in our understanding of the meaning of the HP and SB, and of the MSA more generally. This is evident from the finer details of diachronic changes within single assemblages.
that have emerged from analyses of the HP at, for example, Klasies River (Wurz, 2002; Villa et al., 2010), Rose Cottage Cave (Soriano et al., 2007), PP5-6 (Brown et al., 2009, 2012); Klein Kliphuis (Mackay, 2010, 2011a), Diepkloof (Mackay, 2010, 2011b; Porraz et al., 2013), Klipdrift Rockshelter (Henshilwood et al., 2014) and Sibudu (Soriano et al., 2015). The same applies also to analyses of the SB (Archer et al., 2016) at Hollow Rock Shelter (Högberg and Larsson, 2011), Diepkloof (Porraz et al., 2013), Blombos (Archer et al., 2015; Soriano et al., 2015) and Sibudu (Soriano et al., 2015). These studies are improving the resolution of the questions we can ask of the HP and SB, and about the meaning of these labels to describe assemblages. Of increasing importance is the concept of scales of analysis and the implications of a more finely resolved understanding of the effects of taphonomic processes on our interpretation of chronological and other data used to reconstruct the human past.

Acknowledgements

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References


Mackay, A., 2011b. Potentially stylistic differences between backed artefacts from two nearby sites occupied ~60,000 years before present in South Africa. J. Anthropol. Archaeol. 30, 235–245.


**Figure captions**

Figure 1: a) Comparison of aFMM and CAM (see definitions in text) ages calculated for the same samples from Sibudu; the error bars are σ1 in the notation used here. SIB12 is not shown since only a CAM age was calculated for this sample. This figure shows the same data as Figure 1 of Guérin et al. (2013), but using the dose rate errors as published by Jacobs et al. (2008a), rather than the 7% applied to all beta dose rates by Guérin et al. (2013; see further discussion in our main text). b) Corresponding plot of the difference between each pair of log age estimates (on the vertical axis) against the mean log age estimate of the pair (on the horizontal axis). The grey band is calculated as 2.18 estimated standard deviations above and below zero (based on 12 degrees of freedom). So, for a point within this band, the 95% confidence interval for the true difference includes zero (see text).

Figure 2: Comparison of OSL age estimates obtained by two methods (aFMM and CAM; see definitions in text) for 31 samples from four groups. The difference in log age is plotted against the mean log age for each sample. Scales of ages and their ratios are also shown. The grey band denotes ± 2.04 standard deviations about zero for individual differences (with 30 degrees of freedom), so each point within this band has a 95% confidence interval for the true difference that includes zero.

Figure 3: Published ages for the Pinnacle Point site 5-6 (PP5-6) deposits that contain ‘classic’ HP (DBCS) and other microlithic industries (OBS1 and SADBS). The error bars (σ2) denote the 95% confidence interval for each age. The stippled lines are the end and start ages of the Still Bay (SB) and Howieson’s Poort (HP), respectively, and the grey band marks the gap between the SB and HP based on Estimate B in Tables 2 and 3.
Table 1: Dose rate data, $D_s$ values and OSL ages (in ka) for all samples in this study.

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<th>Beta Dose Rate (±1σ)</th>
<th>CAM Site</th>
<th>FMM Site</th>
</tr>
</thead>
<tbody>
<tr>
<td>SEH1</td>
<td>10 ± 2</td>
<td>1.74 ± 0.09</td>
<td>CAM</td>
<td>10 ± 2</td>
</tr>
<tr>
<td>SEH2</td>
<td>6 ± 2</td>
<td>1.44 ± 0.07</td>
<td>CAM</td>
<td>10 ± 2</td>
</tr>
<tr>
<td>SEH3</td>
<td>6 ± 2</td>
<td>1.42 ± 0.07</td>
<td>CAM</td>
<td>10 ± 2</td>
</tr>
</tbody>
</table>

### Pinnacle Point site 5-6 (PP5-6)

<table>
<thead>
<tr>
<th>Site</th>
<th>Age (kya)</th>
<th>Beta Dose Rate (±1σ)</th>
<th>CAM Site</th>
<th>FMM Site</th>
</tr>
</thead>
<tbody>
<tr>
<td>46784</td>
<td>8 ± 2</td>
<td>0.89 ± 0.04</td>
<td>CAM</td>
<td>10 ± 2</td>
</tr>
<tr>
<td>46785</td>
<td>8 ± 2</td>
<td>0.80 ± 0.04</td>
<td>CAM</td>
<td>10 ± 2</td>
</tr>
</tbody>
</table>

\(^a\)Ages listed here are the preferred ages included in Estimate B (except PP5-6; see text for details). CAM and FMM defined in text.

\(^b\)Beta dose rates are all based on the sample-average beta dose rate obtained from GM-25-5 beta counting; see Jacobs and Roberts (2015) for details.

\(^c\)The FMM was used only when samples showed evidence of mixing (see Jacobs et al., 2008a; Jacobs and Roberts, 2015).

\(^d\)Uncertainties are given at 1σ (standard error of the mean). The ±2 values listed here were derived by combining, in quadrature, all known and estimated sources of random and systematic error, whereas the ±1 values exclude the systematic errors (see text).
The dose rate data for these two samples were transposed in Jacobs et al. (2008a) and are shown correctly here.
Table 2: Maximum likelihood estimates and 95% confidence intervals for the start and end ages (in ka) of the Howieson’s Poort (HP) and Still Bay (SB) periods.

<table>
<thead>
<tr>
<th></th>
<th>End of HP</th>
<th>Start of HP</th>
<th>End of SB</th>
<th>Start of SB</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Age</td>
<td>95% CI</td>
<td>Age</td>
<td>95% CI</td>
</tr>
<tr>
<td>Original</td>
<td>59.5</td>
<td>56.5–62.7</td>
<td>64.8</td>
<td>61.6–68.2</td>
</tr>
<tr>
<td>Estimate A</td>
<td>58.7</td>
<td>55.9–62.5</td>
<td>66.5</td>
<td>62.0–70.9</td>
</tr>
<tr>
<td>Estimate B</td>
<td>58.3</td>
<td>54.5–62.0</td>
<td>66.6</td>
<td>62.2–71.1</td>
</tr>
<tr>
<td>Estimate C</td>
<td>58.2</td>
<td>54.4–61.9</td>
<td>66.8</td>
<td>62.2–71.3</td>
</tr>
</tbody>
</table>

The ‘Original’ estimates are those published in Jacobs et al. (2008a). The three sets of ‘Estimates’ (A, B and C) are explained in the text. In addition to measurement and estimation errors, the confidence intervals for Estimates A, B and C include a relative systematic error of 3%, rather than 2% in the Original estimates. Note that the confidence intervals for different start and end ages are highly dependent and cannot be compared together.
Table 3: Maximum likelihood estimates and 95% confidence intervals for the durations (in ka) of the Howieson’s Poort (HP) period and of the gap between the end of the Still Bay (SB) and start of the HP

<table>
<thead>
<tr>
<th></th>
<th>Duration of HP period</th>
<th>Duration of gap between SB and HP</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Estimate</td>
<td>95% CI</td>
<td>Estimate</td>
</tr>
<tr>
<td>Original</td>
<td>5.3</td>
<td>2.0–8.3</td>
<td>6.7</td>
</tr>
<tr>
<td>Estimate A</td>
<td>7.7</td>
<td>4.9–10.5</td>
<td>4.1</td>
</tr>
<tr>
<td>Estimate B</td>
<td>8.4</td>
<td>5.6–11.1</td>
<td>5.3</td>
</tr>
<tr>
<td>Estimate C</td>
<td>8.6</td>
<td>5.7–11.6</td>
<td>4.8</td>
</tr>
</tbody>
</table>

aBecause we are estimating differences between ages, systematic errors common to all estimates do not contribute to the uncertainties in these estimates. The final column gives p-values for testing the null hypothesis that there is no gap between the SB and HP; the smaller the p-value, the stronger is the evidence in favour of a gap.
Figure 2

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A scatter plot showing the relationship between age (ka) and mean log age. The plot includes data points for Sibudu 1 (n=13), Sibudu 2 (n=7), Diepkloof (n=8), and Blombos (n=3). The mean difference is given as 0.0233 ± 0.0074.
Supplementary Material

Click here to download Supplementary Material: Supplementary Online Material Jacobs and Roberts.docx