

# Carbon-isotopic analysis of individual pollen grains from C<sub>3</sub> and C<sub>4</sub> grasses using a spooling-wire microcombustion interface

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## Abstract

Pollen grains from grasses using the C<sub>3</sub> and C<sub>4</sub> photosynthetic pathways have distinct ranges of  $\delta^{13}\text{C}$  values that may be used to estimate their relative abundance in paleorecords. We evaluated a spooling-wire microcombustion device interfaced with an isotope-ratio mass spectrometer (SWiM-IRMS) for  $\delta^{13}\text{C}$  analysis of individual grass-pollen grains. Pollen from four C<sub>3</sub> and four C<sub>4</sub> grass species was isolated through micromanipulation and analyzed as single grains suspended in water. A carbon yield greater than the  $2\sigma$  range of the carbon content of blanks containing only water was used to distinguish samples containing pollen (“pollen present”) from those not containing pollen. This criterion resulted in the exclusion of  $\sim 45\%$  of the 946 samples applied to the wire. The average  $\delta^{13}\text{C}$  values ( $\pm 1\sigma$ ) of the remaining samples were  $-26.9\text{‰}$  ( $\pm 6.3\text{‰}$ ) and  $-11.5\text{‰}$  ( $\pm 9.6\text{‰}$ ) for C<sub>3</sub> grasses and C<sub>4</sub> grasses, respectively, after blank-correcting the  $\delta^{13}\text{C}$  data. These results suggest that the SWiM-IRMS system can be used to distinguish C<sub>3</sub> from C<sub>4</sub> grass pollen. The high variability in measured  $\delta^{13}\text{C}$  values is likely caused by a combination of factors. These include natural isotopic variability among individual pollen grains; the relatively poor precision that can be obtained when determining  $\delta^{13}\text{C}$  values of such small samples; and the uncertainty in the magnitude, isotopic composition, and stability of the analytical blank. Nonetheless, high percentages of individual pollen grains were correctly classified as being of either C<sub>3</sub> or C<sub>4</sub> origin. On average, 90% (range = 78–100%) of pollen grains from C<sub>3</sub> grasses had  $\delta^{13}\text{C}$  values more negative than the cutoff threshold of  $-19.2\text{‰}$ ; while 84% (range = 77–90%) of pollen grains from C<sub>4</sub> grasses had  $\delta^{13}\text{C}$  values more positive than  $-19.2\text{‰}$ . Compared with analysis using an elemental analyzer interfaced with an IRMS (EA-IRMS), the number of pollen grains required for  $\delta^{13}\text{C}$ -based evaluation of C<sub>3</sub>/C<sub>4</sub> grass composition is many times lower with the SWiM-IRMS. Additionally,  $\delta^{13}\text{C}$  data from the SWiM-IRMS does not need to be incorporated into a mixing model to derive estimates of the abundance of C<sub>3</sub> and C<sub>4</sub> grass pollen. Carbon-isotopic analysis of individual grass-pollen grains using the SWiM-IRMS system may help improve our understanding of the evolutionary and ecological significance of grass taxa in the paleorecord.

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## 1. INTRODUCTION

Today grass-dominated communities cover about one-third of Earth’s land surface, exert a large influence on glo-

bal biogeochemical cycles, and provide valuable sources of food for humans and herbivores (Jacobs et al., 1999; Saugier and Roy, 2000). Within these and other communities, grasses use two major pathways of carbon fixation in plants: C<sub>3</sub> and C<sub>4</sub> photosynthesis. Based on their distinct physiologies, C<sub>3</sub> and C<sub>4</sub> grasses should respond differently to changes in important environmental variables such as atmospheric CO<sub>2</sub> concentrations, aridity, and temperature

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(Ehleringer, 2005). Elucidating the response of C<sub>3</sub> and C<sub>4</sub> grasses to past changes in such variables is useful for projecting how grass community composition may respond to future environmental changes. It also provides the basis for understanding the factor(s) driving the origin of the C<sub>4</sub> physiology in grasses, which is considered a significant evolutionary achievement (Osborne and Beerling, 2006) in large part because C<sub>4</sub> grasses comprise <~2% of the total number of terrestrial plant species (Sage et al., 1999) but account for ~20% of global net primary production (Lloyd and Farquhar, 1994).

Despite the importance of distinguishing the abundance of C<sub>3</sub> and C<sub>4</sub> grasses in paleorecords, such information cannot be directly derived from pollen assemblages because grass pollen cannot be morphologically resolved beyond the family level (Fægri et al., 1989). Recent studies (e.g., Clark et al., 2001; Huang et al., 2001; Smith and White, 2004; Nelson et al., 2006) have attempted to use new isotopic and morphological proxies to estimate the relative abundances of C<sub>3</sub> and C<sub>4</sub> plants in sediment records. For example, using a micromanipulator to isolate grass-pollen grains that are then analyzed using a modified elemental analyzer interfaced with a standard isotope-ratio mass spectrometer (EA-IRMS), Nelson et al. (2006) showed that differences in δ<sup>13</sup>C values between C<sub>3</sub> and C<sub>4</sub> grasses could be reliably detected in grass-pollen samples containing a minimum of 600 grains. By incorporating δ<sup>13</sup>C data from samples of bulk grass pollen into a simple two end-member mixing model, the relative abundance of C<sub>3</sub> and C<sub>4</sub> grasses in a sample could be estimated. This new ability to assess the relative abundance of C<sub>3</sub> and C<sub>4</sub> grasses is an important step towards improving our understanding of grass taxa in geological records.

However, there are two main drawbacks to utilizing δ<sup>13</sup>C measurements made on bulk samples of grass pollen. First, because it takes ~10 h to manually isolate the 600 fossil grass-pollen grains for a single analysis, the technique greatly limits the number of samples that may be routinely analyzed. The second drawback is that the relative proportion of C<sub>3</sub> and C<sub>4</sub> grasses must be estimated from a mixing model. For example, consider the equation

$$\% C_4 = 100 * (\delta_s - \delta_3) / (\delta_4 - \delta_3),$$

where δ<sub>s</sub> is the δ<sup>13</sup>C value of the 600-grain sample, and δ<sub>3</sub> and δ<sub>4</sub> are the δ<sup>13</sup>C end-members for C<sub>3</sub> and C<sub>4</sub> grasses, respectively. End-member values of δ<sub>3</sub> and δ<sub>4</sub> are usually assumed to be constant (~−27‰ for C<sub>3</sub> and ~−13‰ for C<sub>4</sub>). However, both are known to vary greatly across environmental gradients and taxonomic groups (range = −33‰ to −22‰ for C<sub>3</sub> and −15‰ to −10‰ for C<sub>4</sub>), which means that results from such mixing models have large uncertainties (Cerling, 1999). Using the extreme ranges of δ<sub>3</sub>, a mixed pollen sample having a bulk δ<sup>13</sup>C value of −22‰ could represent 55% abundance of C<sub>4</sub> grasses if δ<sub>3</sub> was −33‰ (and δ<sub>4</sub> was −13‰), or 0% C<sub>4</sub> abundance if δ<sub>3</sub> was −22‰. The end-member variability means that mixing models are not suitable for detecting relatively small fluctuations in C<sub>4</sub> abundance or the presence of C<sub>4</sub> grasses at abundances below ~30–40% of the total grass-pollen grains. Thus EA-IRMS analysis of

bulk samples of grass pollen is not suitable for studying the origin of C<sub>4</sub> grasses in the geological record.

We analyzed the <sup>13</sup>C composition of 946 individual grains of pollen from eight C<sub>3</sub> and C<sub>4</sub> grasses using a spooling-wire microcombustion device interfaced with an IRMS (SWiM-IRMS). Here we evaluate the potential of this technique to overcome the limitations of δ<sup>13</sup>C analysis of bulk grass pollen for estimating the abundance of C<sub>3</sub> and C<sub>4</sub> grass pollen. Specifically, we expected that (1) when analyzing individual grains using the SWiM-IRMS, the total number of grains required to accurately assess C<sub>3</sub> and C<sub>4</sub> grass-pollen abundance in an unknown sample would be significantly lower than that required for analysis using the EA-IRMS, and (2) analysis of single grains would allow direct counts of the number of C<sub>3</sub> and C<sub>4</sub> grass-pollen grains in a sample (using a simple threshold to distinguish δ<sup>13</sup>C values from C<sub>3</sub> and C<sub>4</sub> samples). Analysis of single pollen grains using the SWiM-IRMS would eliminate the need to estimate the abundance of C<sub>3</sub> and C<sub>4</sub> grass pollen using a mixing model.

## 2. MATERIALS AND METHODS

Modern pollen samples were obtained from four C<sub>3</sub> and four C<sub>4</sub> grass species (Table 1). All samples were treated following standard pollen preparation techniques modified to eliminate carbon-containing chemicals (Nelson et al., 2006). Pollen grains treated in this way are primarily composed of chemically resistant sporopollenin, the main constituent of fossil pollen grains (Loader and Hemming, 2000).

All samples were manipulated in nano-pure water stored in a pre-combusted beaker. A steel and glass syringe was used to apply discrete ~0.6-μl drops of this water to the right side of a methanol-cleaned microscope slide, as well as a larger pool (~80 μl) of water to the left side of the slide. For each species, a separate syringe was used to apply pollen samples to the large pool of water on the slide. Grass pollen was detected at 200× magnification, and individual grains were transferred from the pool of water into the ~0.6-μl drops of water using micromanipulation, as previously described (Nelson et al., 2006). A total of 423 individual pollen grains from four C<sub>3</sub> grass species and 523 individual pollen grains from four C<sub>4</sub> grass species (946 total grains) were applied to the SWiM-IRMS system in ~0.6-μl drops of water for determination of their δ<sup>13</sup>C values (Table 1). The syringe (0.460 mm inner diameter) used to apply samples to the wire was rinsed between each sample. For comparison with δ<sup>13</sup>C data from individual grains, δ<sup>13</sup>C values of bulk aliquots of the same pollen samples from each species were analyzed using EA-IRMS, as previously described (Nelson et al., 2006). All pollen-δ<sup>13</sup>C data were corrected for the post-industrial depletion of atmospheric <sup>13</sup>C, as described in Nelson et al. (2006).

The configuration and operation of a SWiM-IRMS system was previously described in detail (Brand and Dobberstein, 1996; Sessions et al., 2005; Eek et al., 2007). The system used here is a modified version built at Harvard University following the original prototype. Modifications include the use of a combustion furnace

Table 1  
Summary data from individual pollen-grain samples

Name	Photosynthetic type	Samples applied to wire	Samples $>2\sigma$ of blank	$\delta^{13}\text{C}$ ( $1\sigma$ range)	Fraction $>$ or $<-19.2\text{‰}$
<i>Andropogon gerardii</i>	C <sub>4</sub>	114	70	-10.5 (9.4)	86
<i>Sorghum halepense</i>	C <sub>4</sub>	143	60	-10.7 (9.3)	90
<i>Sorghum vulgare</i>	C <sub>4</sub>	90	57	-13.5 (6.0)	82
<i>Spartina pectinata</i>	C <sub>4</sub>	176	64	-11.1 (13.8)	77
C <sub>4</sub> total		523	251	-11.5 (9.6), average	84, average
<i>Agropyron repens</i>	C <sub>3</sub>	101	73	-23.1 (7.7)	78
<i>Bromus inermis</i>	C <sub>3</sub>	116	85	-26.8 (5.4)	95
<i>Elymus canadensis</i>	C <sub>3</sub>	101	55	-30.5 (4.7)	100
<i>Festuca elatior</i>	C <sub>3</sub>	105	55	-27.1 (7.3)	85
C <sub>3</sub> total		423	268	-26.9 (6.3), average	90, average
Unknown 1		120	67	-14.1 (8.3)	
Unknown 2		111	64	-24.7 (5.6)	

$\delta^{13}\text{C}$  data are based on samples with Vs yield  $>2\sigma$  value of the “processing” blank for each species.

The fraction  $>-19.2\text{‰}$  was used to distinguish C<sub>4</sub> from C<sub>3</sub>, whereas the fraction  $<-19.2\text{‰}$  was used to distinguish C<sub>3</sub> from C<sub>4</sub>.

of smaller internal volume (0.5 mm i.d.  $\times$  14 cm length), a shorter Nafion drying tube (10 cm), and a ThermoFinnigan DeltaPlus Advantage IRMS. The wire (0.25 mm nickel wire moving at 0.8 cm/s) is passed through a cleaning oven (850 °C) purged with filtered air. A syringe is then used to transfer  $\sim 0.6\ \mu\text{l}$  drops of water (containing pollen) from the microscope slide to the wire. The wire passes through a drying oven (120 °C) before entering a combustion oven (800 °C). Atmosphere is excluded from the combustion furnace by positive pressure of the helium carrier gas (6.5 psi). A portion of the combustion gases flows through a countercurrent Nafion membrane to remove H<sub>2</sub>O, through an open split, and then to the IRMS. CO<sub>2</sub> reference gas is supplied via a Conflo III interface, and samples are analyzed as sets of seven injections spaced at 45-s intervals, flanked by three initial and two final pulses of reference gas. Sample yields (as CO<sub>2</sub> gas) are reported in units of Volts-seconds (Vs, peak area) of the mass 44 ion current. Using combustion yields of leucine injections containing 5 nmol carbon, the approximate conversion is  $\leq 1.4$  Vs/nmol C.

To assess the contribution of background carbon we also analyzed three types of blanks: water from the beaker (called “water” blanks), water from the microscope slide (called “slide-water” blanks), and water from the microscope slide to which single grains of grass pollen were added and then subsequently removed (called “processing” blanks). “Processing” blanks were prepared for each species. These blanks were analyzed at the same time as pollen samples, which is critical in order to account for potential changes in the split ratio. All blanks were applied to the wire in  $\sim 0.6\ \mu\text{l}$  aliquots. Assessing the contribution of carbon from blanks enables us to account for the possibility of pollen grains falling off the wire before reaching the combustion oven or of pollen that produces too little carbon to be distinguished from carbon-containing background (from the water or from the sample processing).

### 3. RESULTS AND INTERPRETATIONS

#### 3.1. Precision versus sample size

To assess the shot-noise limited precision of average-sized grains of grass pollen we analyzed the  $\delta^{13}\text{C}$  composition of two dissolved organic standards, leucine and dextrose, which have  $\delta^{13}\text{C}$  values similar to those of C<sub>3</sub> and C<sub>4</sub> plants, respectively. Results show that the analytical precision of  $\delta^{13}\text{C}$  values from leucine and dextrose is no worse than  $\pm 5\text{‰}$  at sample sizes similar to those of pollen grains (Fig. 1). Thus, the observed scatter in the pollen  $\delta^{13}\text{C}$  data (Section 3.4,  $\delta^{13}\text{C}$  composition of blank-corrected samples) cannot be caused solely by analytical imprecision or sample handling blanks.

#### 3.2. Peak areas of samples

The peak areas, corresponding to combustion yield of CO<sub>2</sub> gas, do not differ between “water” and “slide-water” blanks (Fig. 2). However, the peak areas of “processing” blanks are significantly greater than “water” blanks (Fig. 2), suggesting the addition of water-soluble carbon to the sample during chemical treatment. This excess carbon amounts to  $\sim 0.1$  Vs ( $\leq 0.07$  nmol carbon) per grain (Fig. 2). Because no carbon-containing chemicals were used during treatment, the origin of the excess carbon is not immediately clear. There are three possible sources. First, the soluble component derives from traces of carbon removed from the inner part of the grains (intine) during treatment (Nelson et al., 2006) that remained with the sample through the treatment process. Second, airborne particles in the laboratory contribute the excess carbon. Third, soluble material from the surfaces of the plastic tubes in which the samples were treated was released during treatment and remained with the sample through the treatment process. The two latter possibilities imply that the  $\delta^{13}\text{C}$  value of “processing” blanks would be uniform between

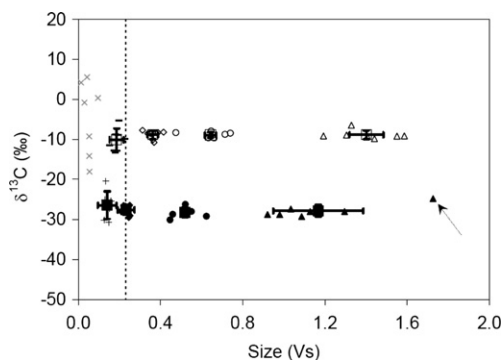


Fig. 1. Isotopic data for organic standards. Hollow symbols, dextrose ( $\delta^{13}\text{C}_{\text{true}} = -9.8\text{‰}$ ); solid symbols, leucine ( $\delta^{13}\text{C}_{\text{true}} = -29.8\text{‰}$ ). Data are shown for samples of nominal mass 1.25 nmol C (triangles), 0.625 nmol C (circles), 0.312 nmol C (diamonds), and 0.156 nmol C (+ and - symbols);  $n = 7$  replicates for each standard.  $\text{CO}_2$  area data are in Volt-second units, Vs. The means and sample standard deviations are shown for each set. One of the 1.25-nmol C leucine samples appears to be an outlier (arrow). The dashed line at  $\sim 0.2$  Vs indicates the approximate  $2\sigma$  threshold used to distinguish the presence or absence of pollen in samples. Isotopic values for pure water blanks (gray 'x' symbols) also are shown.

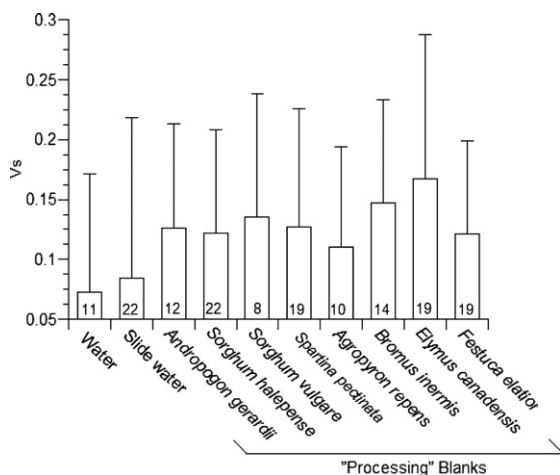


Fig. 2.  $\text{CO}_2$  area data of blanks. Means and  $2\sigma$  ranges are shown. The numbers inside the bars indicate the number of samples analyzed. “Water” indicates water applied directly to the spooling-wire. “Slide water” indicates water applied to a microscope slide and then to the spooling-wire. For the blanks of each species (“processing” blanks) a pollen grain was dispensed and then removed from water on a microscope slide and 0.6  $\mu\text{l}$  of the water was subsequently applied to the spooling-wire. The “water” and “slide water” blanks are not statistically different ( $p = 0.30$ ). The blanks of each species are statistically different from the “water” ( $p < 0.05$ , with the exception of *Agropyron repens*, for which  $p = 0.09$ ).

$\text{C}_3$  and  $\text{C}_4$  samples, whereas the first possibility implies that the blanks would differ between  $\text{C}_3$  and  $\text{C}_4$  samples. The isotope data presented below (Section 3.3) suggest that the latter possibilities are the most likely explanations.

Regardless of the precise origin of the carbon in the “processing” blanks, their peak areas were used to determine the presence or absence of a pollen grain in a sample. Of the 946 individual pollen-grain samples applied to the moving wire, 303 (32%) and 427 (45%) have total peak areas within  $\pm 1\sigma$  and  $\pm 2\sigma$ , respectively, of the peak areas of “processing” blanks. For each species we defined a sample as “pollen present” if its Vs yield was greater than the  $2\sigma$  value of the “processing” blank and “pollen absent” if its Vs yield was less than or equal to the  $2\sigma$  value of this blank. The more conservative  $\pm 2\sigma$  criterion results in better separation of  $\delta^{13}\text{C}$  values from  $\text{C}_3$  and  $\text{C}_4$  grains than the  $\pm 1\sigma$  range. The improvement in separation of  $\delta^{13}\text{C}$  values from  $\text{C}_3$  and  $\text{C}_4$  grains when using an even more conservative  $\pm 3\sigma$  criterion is minimal (improvement of  $\pm 2\%$ ), but the percentage of eliminated samples (an additional 12%) is much greater at a  $\pm 3\sigma$  cutoff. In addition, the significant positive relationship between mean grain diameter and peak areas ( $n = 8$ ,  $r = 0.73$ ,  $p = 0.038$ ) means that setting the “pollen present” criterion too high (e.g.,  $\pm 3\sigma$  or higher) would weight paleo-reconstructions toward species that produce larger grains. Thus the  $\pm 2\sigma$  criterion appears optimal. Finally, the sizes of the “processing” blanks for each species are not significantly different from the average size of the blank for all species combined (average  $p$ -value = 0.36, all  $p$ -values  $> 0.02$ ). Thus the difference is minimal between using the individual  $\pm 2\sigma$  ranges of the “processing” blanks for each species and using the average  $\pm 2\sigma$  range of all of the species as the threshold for “pollen present.”

A total of 268  $\text{C}_3$  and 251  $\text{C}_4$  samples (63 and 48% of the total number of  $\text{C}_3$  and  $\text{C}_4$  grains applied, respectively) have peak areas exceeding their respective  $\pm 2\sigma$  thresholds (Fig. 2). It is not surprising that nearly half (45%) of the individual analyses have to be excluded. The most likely explanation is that pollen grains fall off the moving wire in transit before reaching the combustion oven. Natural variability in carbon content of individual grains may also cause the peak areas of some combusted grains to be genuinely smaller than the  $\pm 2\sigma$  threshold and thus indistinguishable from blanks. We also saw no evidence for grains being unintentionally carried-over in the syringe (resulting in no grains in some samples and more than one grain in others). However, because it is impossible to discriminate between these options, all Vs data smaller than the  $\pm 2\sigma$  cutoffs were excluded.

### 3.3. Blank corrections

The  $\delta^{13}\text{C}$  value of the blank could not be measured directly because its low carbon content led to poor precision (Fig. 1), and the cause of the poor precision (e.g., true variability versus instrumental variability) could not be assessed. However, we were able to determine the average isotopic composition of the “processing” blanks by comparing the  $\delta^{13}\text{C}$  values of bulk pollen aliquots from EA-IRMS with the average values of individual pollen grains obtained from SWiM-IRMS. All of the  $\text{C}_3$  species fall near a 1:1 line, whereas  $\delta^{13}\text{C}$  values for three out of four  $\text{C}_4$  species fall below this line (Fig. 3, hollow symbols). The one  $\text{C}_4$

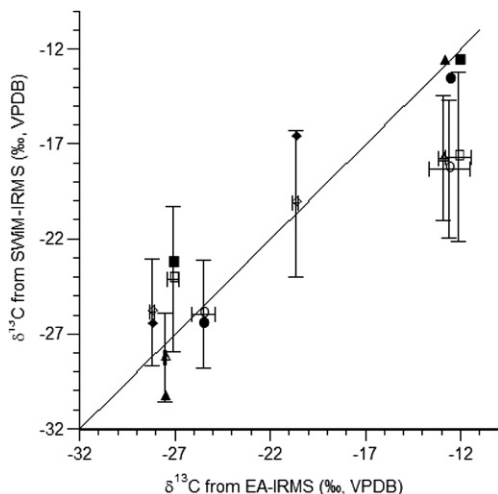


Fig. 3. Comparison of mean  $\delta^{13}\text{C}$  data from  $\text{C}_3$  and  $\text{C}_4$  grass pollen obtained using EA-IRMS and SWiM-IRMS. Hollow symbols represent raw data; closed symbols represent blank-corrected data as described in Section 3.3. Error bars ( $\pm 1\sigma$ ) are displayed only on hollow symbols.  $\text{C}_3$  species (on left): square, *Agropyron repens*; circle, *Bromus inermis*; triangle, *Elymus canadensis*; diamond, *Festuca elatior*.  $\text{C}_4$  species (on right): square, *Andropogon gerardii*; circle, *Sorghum halepense*; triangle, *Sorghum vulgare*; diamond, *Spartina pectinata*. The solid line represents a 1:1 relationship between  $\delta^{13}\text{C}$  EA-IRMS and SWiM-IRMS.

species falling on the 1:1 line (*Spartina pectinata*) has a much more negative  $\delta^{13}\text{C}$  value ( $-20.7\text{‰}$ , based on EA-IRMS) than is typical of  $\text{C}_4$  species. These results suggest that if the size of the “processing” blank is similar among all samples (which it appears to be, based on Fig. 2, and as stated above in Section 3.2), the greater negative bias on the  $\text{C}_4$  data means that the  $\delta^{13}\text{C}$  value of the blank is likely close to the typical isotopic composition of  $\text{C}_3$  species. Thus we suggest that the greater peak areas of “processed” blanks than of “water” blanks results from traces of water-soluble carbon from the plastic tubes in which the samples were treated or traces of carbon from airborne particles.

To estimate the mass and  $^{13}\text{C}$  composition of the blank we used the equation,

$$\delta_{\text{SWiM}} = \delta_{\text{ea}} + (M_{\text{b}}/M_{\text{SWiM}})(\delta_{\text{b}} - \delta_{\text{ea}}),$$

where  $\delta_{\text{SWiM}}$  and  $M_{\text{SWiM}}$  denote the average  $\delta^{13}\text{C}$  and Vs values, respectively, for a species obtained using SWiM-IRMS,  $\delta_{\text{ea}}$  equals the average  $\delta^{13}\text{C}$  value obtained using EA-IRMS, and  $M_{\text{b}}$  and  $\delta_{\text{b}}$  signify the average Vs and  $\delta^{13}\text{C}$  value of the blank. We solved for  $M_{\text{b}}$  and  $\delta_{\text{b}}$  iteratively by minimizing the sum of the absolute difference between  $\delta_{\text{SWiM}}$  predicted using the above equation and the average  $\delta^{13}\text{C}$  value measured for each species using the SWiM-IRMS. This process (excluding the anomalous *S. pectinata* data) yields values of 0.16 Vs and  $-25.3\text{‰}$  for  $M_{\text{b}}$  and  $\delta_{\text{b}}$ , respectively. The value of  $M_{\text{b}}$  calculated this way (0.16 Vs) is very similar to the average mass of all of the “processing” blanks (0.14 Vs, Fig. 2) as determined by direct measurements. When the SWiM-IRMS data are corrected using these calculated values of  $M_{\text{b}}$  and  $\delta_{\text{b}}$ , they

showed a much closer fit (slope of regression improves from 0.57 to 0.95) with the EA-IRMS data (Fig. 3, solid symbols). This result suggests that the interpretation that  $M_{\text{b}}$  and  $\delta_{\text{b}}$  represent a generally uniform addition of exogenous carbon is reasonable; however, we acknowledge that this average  $\delta^{13}\text{C}$  value of the “processing” blank could reflect a wide range of true variability. The precise source of the excess carbon is uncertain, although it appears more likely to be from plasticware than airborne particles, because if airborne particles were the sole component of the blank our “slide-water” and “processing” blanks should have peak areas of similar size, in contrast to our results (Fig. 2). If the *S. pectinata* data are included in the above calculations,  $M_{\text{b}}$  and  $\delta_{\text{b}}$  are 0.16 Vs and  $-24.9\text{‰}$ , respectively. Using  $-24.9\text{‰}$  rather than  $-25.3\text{‰}$  for  $\delta_{\text{b}}$  results in a difference of only 0.16‰ in  $\delta_{\text{SWiM}}$ ; thus the effect of excluding *S. pectinata* is minimal. We blank-corrected all of the SWiM-IRMS data using values of 0.16 Vs and  $-25.3\text{‰}$  for  $M_{\text{b}}$  and  $\delta_{\text{b}}$ , respectively.

#### 3.4. $\delta^{13}\text{C}$ composition of blank-corrected samples

The mean  $\delta^{13}\text{C}$  values of the blank-corrected individual pollen grains obtained using the SWiM-IRMS system fall within the expected ranges for  $\text{C}_3$  and  $\text{C}_4$  plants for each species (Fig. 4). Thus the SWiM-IRMS may be used to distinguish populations of individual  $\text{C}_3$  and  $\text{C}_4$  pollen grains. However, there is large variation around these means, and many individual data points exceed the ranges expected for  $\delta^{13}\text{C}$  values of  $\text{C}_3$  and  $\text{C}_4$  plants, in both positive and negative directions. For example, although the mean  $\delta^{13}\text{C}$  value for *Sorghum halepense* ( $-10.7\text{‰}$ ) is within the typical range of  $\delta^{13}\text{C}$  values expected for  $\text{C}_4$  plants, both the standard deviation ( $\pm 9.3\text{‰}$ ) and the absolute range of values measured ( $-34.8\text{‰}$  to  $19.7\text{‰}$ ; Fig. 4) exceed the  $-10$  to  $-15\text{‰}$  natural range. The breadth of the scatter in  $\delta^{13}\text{C}$  values at a given peak size also varies among species (Fig. 4), showing that some species have more natural or analytical variability than others.

Sporopollenin-rich organic matter could be expected to deviate from the bulk plant isotopic composition by 5–7‰ in the negative direction, because of the predominance of lipidic carbon in the biosynthesis of sporopollenin (Hayes, 2001; Shaw and Yeadon, 1964). However, this difference is not a large enough effect to explain the occasional observations of  $\text{C}_4$  samples with  $\delta^{13}\text{C}$  values  $< -22\text{‰}$ . Similarly, there is no (known) concomitant evidence for significant isotopic enrichment that could result in  $\delta^{13}\text{C}$  values more positive than  $\sim -8\text{‰}$ . A possible explanation for either of these sets of outliers would require intracellular partitioning of carbon to be governed by branched pathways, allowing some pools of biosynthetic intermediates to become isotopically very heavy, or very light (Hayes, 2001). Although such variation in intracellular allocation of carbon may explain some of the excess isotopic variability observed here, it is unlikely to account fully for the wide range of values observed in the data.

The wide range of scatter in the data must have a significant analytical component. Individual grains of grass pollen contain very small amounts of carbon, and thus some of

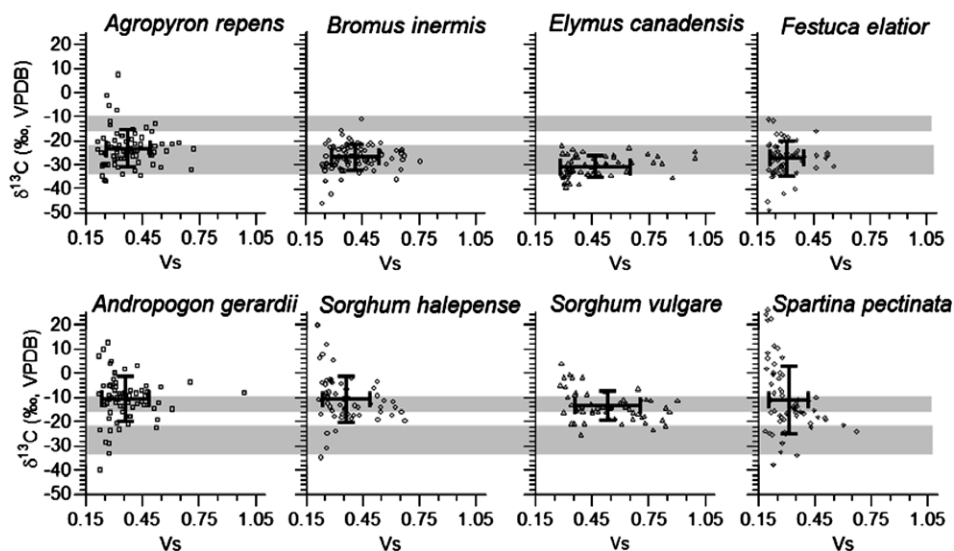


Fig. 4.  $\delta^{13}\text{C}$  and  $V_s$  data from individual grains of pollen from  $\text{C}_3$  (top row) and  $\text{C}_4$  (bottom row) grasses. Only samples exceeding the  $2\sigma$  range of the mean  $V_s$  value of blanks are shown. The mean value for each species is shown with  $1\sigma$  error bars. The shaded boxes encompass the typical ranges (Bender, 1971; Cerling, 1999) of  $\delta^{13}\text{C}$  values for  $\text{C}_3$  and  $\text{C}_4$  plants.

the variation in  $\delta^{13}\text{C}$  values may be caused by relatively poor precision and accuracy of the measurements made with the SWiM-IRMS. However, the average grain of treated grass pollen is  $\sim 0.38 V_s$ , and the isotopic precision that can be obtained for replicate injections of leucine and dextrose in this size range is no worse than  $\pm 5\%$  (Fig. 1), suggesting that poor precision is not the only cause of variation in  $\delta^{13}\text{C}$  values. The variation in final reported values also incorporates a large amount of uncertainty in both the isotopic composition of—and more importantly, the isotopic variability of—the analytical “processing” blank. Calculating  $\delta^{13}\text{C}_b$  by the methods above yields only the average value but not the absolute range of  $\delta^{13}\text{C}_b$ . Together, these factors would be expected to most greatly affect the  $\delta^{13}\text{C}$  values of the samples with the smallest reported peak areas. The observed range of variation in  $\delta^{13}\text{C}$  values (Fig. 4) is indeed greatest for samples with  $V_s$  values below  $\sim 0.3$ , and the uncertainty in the blank correction has the largest impact on these samples. The fact that none of the species trend toward an average  $\delta_{\text{blank}}$  value (or a species-specific  $\delta_{\text{blank}}$  value) at small sample sizes suggests that  $\delta^{13}\text{C}_b$  may have greater variability than the natural  $\delta^{13}\text{C}$  variability of the pollen grains. However, we cannot currently test this hypothesis because along with variability in the “processing blank,” analytical imprecision and true natural variation may also influence the wide range of scatter in the data.

The  $\delta^{13}\text{C}$  values of individual pollen grains from  $\text{C}_3$  and  $\text{C}_4$  grasses obtained using the SWiM-IRMS system exhibit large variations and overlap with one another. Therefore, it is not feasible to distinguish individual grains of  $\text{C}_3$  and  $\text{C}_4$  grass pollen by simply classifying the data into the typical ranges of  $\text{C}_3$  and  $\text{C}_4$  plants (Fig. 4,  $-33$  to  $-22\%$  and  $-15$  to  $-10\%$ , respectively). The SWiM-IRMS method cannot at present be used to count single occurrences of “a  $\text{C}_3$  grain” or “a  $\text{C}_4$  grain,” because even the largest grains (e.g., those with a  $\text{CO}_2$  yield  $>0.75 V_s$ ) occasionally

produced  $\delta^{13}\text{C}$  values outside of their typical ranges and could be misclassified (Fig. 4). However, the vast majority of  $\delta^{13}\text{C}$  values from  $\text{C}_3$  grasses is more negative than the typical range for  $\text{C}_4$  grasses; and conversely, most  $\delta^{13}\text{C}$  values from  $\text{C}_4$  grasses are more positive than a typical  $\text{C}_3$  range. In an approximately equal distribution of data points (268  $\text{C}_3$  and 251  $\text{C}_4$ ), the data are clearly bimodal (Fig. 5). Because the data are normally distributed we attempted to use a Gaussian mixture model (<http://cobweb.ecn.purdue.edu/~bouman/software/cluster/>) to fit distributions to the  $\text{C}_3$  and  $\text{C}_4$  populations, deconvolute the populations, and directly determine the proportion of the data in each distribution. When  $\text{C}_4$  abundance was  $<50\%$  the mixture model classified the proportion of  $\text{C}_4$  grains well. However, the mixture model did a poor job of classifying the proportion of  $\text{C}_4$  grains when their abundance was  $>50\%$ , likely because the  $\text{C}_4$  data are noisier than the  $\text{C}_3$  data. Thus we decided to distinguish populations of  $\text{C}_3$  and  $\text{C}_4$  grass pollen based on their distribution around a  $\delta^{13}\text{C}$  threshold value.

We explored three ways for establishing this threshold value. We first used the mid-point between the average  $\delta^{13}\text{C}$  values of the EA-IRMS pollen data (excluding *S. pectinata* data), which yields a threshold value of  $-19.7\%$  (Fig. 3). However, the greater variation associated with the SWiM-IRMS data suggests that EA-IRMS data from bulk samples may not be appropriate for establishing the threshold. This is especially true if the SWiM-IRMS imparts a systematic offset to the data (accuracy issue) for which we currently have no means of assessing or correcting. We next binned the SWiM-IRMS data by increments of  $1.0\%$  and chose a threshold that minimized the sum of  $\text{C}_3$  bins with  $\delta^{13}\text{C}$  values more positive than the threshold and  $\text{C}_4$  bins with  $\delta^{13}\text{C}$  values more negative than the threshold. This approach produces a threshold value of  $-20.0\%$ . However, this value may be biased by individual species

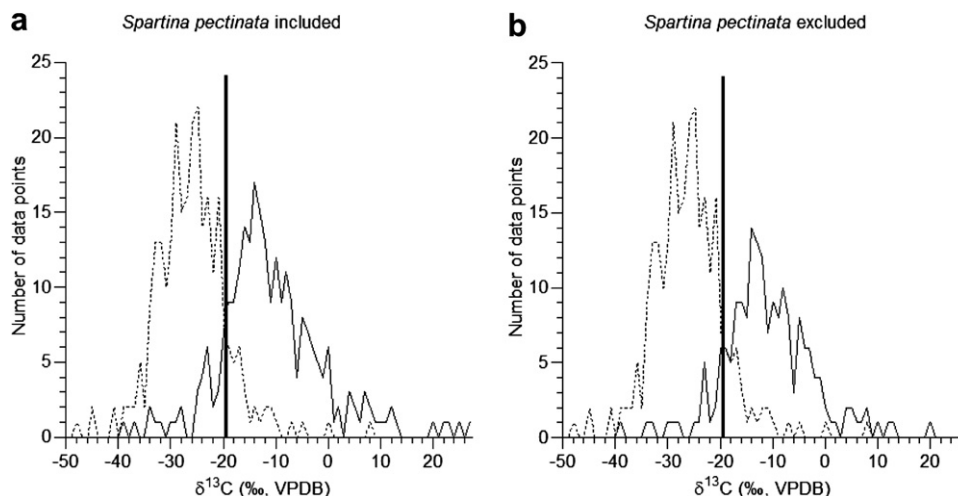


Fig. 5. Histograms of data presented in Fig. 4. The dashed lines represent data from  $C_3$  grasses and the solid lines data from  $C_4$  grasses. The thick lines at  $-19.2\text{‰}$  signify the threshold distinguishing  $C_3$  from  $C_4$  grass pollen, as discussed in the text.

that are represented more frequently in the data set. Therefore, we also calculated the mid-point ( $-19.2\text{‰}$ ) between the average  $\delta^{13}\text{C}$  values for the  $C_3$  ( $-26.9\text{‰}$ ) and  $C_4$  ( $-11.5\text{‰}$ ) grass-pollen samples from the SWiM-IRMS (Table 1). This approach is more comparable to the approach using the EA-IRMS data, and unlike the binning method, the result of  $-19.2\text{‰}$  partially removes the effect of unequal numbers of grains for all species. Regardless, the three approaches yielded similar threshold values ( $-19.7\text{‰}$ ,  $-20.0\text{‰}$ , and  $-19.2\text{‰}$ ); thus they would not lead to significantly different estimates of the abundance of  $C_3$  and  $C_4$  grass pollen. We chose to retain  $-19.2\text{‰}$  as the final threshold value because it results in a slightly more conservative estimate of the proportion of  $C_4$  grass pollen, which is critical when  $C_4$  grass pollen is at low abundance ( $< \sim 30\%$ ). The threshold value of  $-19.2\text{‰}$  remains the same whether the mean value of *S. pectinata* ( $-11.1\text{‰}$ , Table 1) is included (Fig. 5a) or excluded (Fig. 5b) from the average  $\delta^{13}\text{C}$  value of individual grains of  $C_4$  grass pollen.

### 3.5. Samples with unknown $C_3$ and $C_4$ proportions

To estimate the number of individual pollen-grain measurements that would be necessary to accurately estimate the abundance of  $C_3$  and  $C_4$  grass pollen in an unknown sample we developed subroutines in Matlab (The MathWorks, Inc.) to resample and count data points. The master dataset used for the calculations included the 268  $C_3$  and 251  $C_4$  samples classified as “pollen present.” From the master dataset a series of new datasets (dataset<sub>new</sub>) were created. Each dataset<sub>new</sub> contained specified percentages of  $C_3$  and  $C_4$  pollen grains (e.g., 0%, 10%, ..., 100%). The maximum possible number of grains was included in each dataset<sub>new</sub>. For example, the dataset<sub>new</sub> with  $< 48.4\%$  ( $= 251/[251 + 268]$ )  $C_4$  grains contained all 268 of the  $C_3$  data points; the dataset<sub>new</sub> with  $> 48.4\%$   $C_4$  contained all 251 of the  $C_4$  data points. From each dataset<sub>new</sub> 50, 100, or 150 grains were randomly selected. The abundances of  $C_3$  and  $C_4$  grains in each dataset<sub>new</sub> were then predicted

using the threshold value of  $-19.2\text{‰}$ . The accuracy of predictions (e.g.,  $\pm 10\%$  means that the predicted distribution reproduced the known distribution to within  $\pm 10\%$ ) was then evaluated by comparison with the true proportions of  $C_3$  and  $C_4$  grass pollen in each dataset<sub>new</sub>. A subroutine was created to repeat the evaluation process 250 times for each of the 30 datasets<sub>new</sub> that were randomly created for each proportion of  $C_3$  and  $C_4$  grass pollen from the master dataset.

The results from this resampling exercise show that the proportion of correctly classified trials increased as a function of the number of pollen grains included in a sample (Fig. 6). For example, in a sample of 50 individual grains in which the true abundance of  $C_4$  grass pollen is 50%, the counted abundance of  $C_4$  grass pollen would be expected to fall within  $\pm 10\%$  of the true abundance with only  $1\sigma$  (68%) confidence. But if the number of grains in the sample is increased to 100, the counted abundance would be expected to fall within  $\pm 10\%$  of the true abundance with  $2\sigma$  confidence (or 95% of the time). This shows the power of analyzing greater numbers of pollen grains within a given sample. Conversely, the proportion (e.g., 68% or 95%) of correctly classified trials is larger if the acceptable range of error (e.g.,  $\pm 10\%$ ) is larger (Fig. 6). In the example above, the counted abundance of  $C_4$  grass pollen would be within  $\pm 10\%$  of the true abundance with  $1\sigma$  (68%) confidence; but that same 50-grain sample would be classified to within  $\pm 15\%$  of the true abundance with  $2\sigma$  (95%) confidence.

The most stringent classification is achieved by measuring a 150-grain sample, which under certain conditions can be classified to within  $\pm 10\%$  of its “true”  $C_3/C_4$  distribution if the sample falls in the range of  $\sim 30\text{--}60\%$  total  $C_4$  abundance (Fig. 6b). The accuracy of classification is also better at low, rather than high, relative abundance of  $C_4$  grass pollen, regardless of the number of grains analyzed (Fig. 6). For example, in a sample of 50 individual grains the counted abundance of  $C_4$  grass pollen would be within  $\pm 20\%$  of the true abundance at the 95% ( $2\sigma$ ) confidence

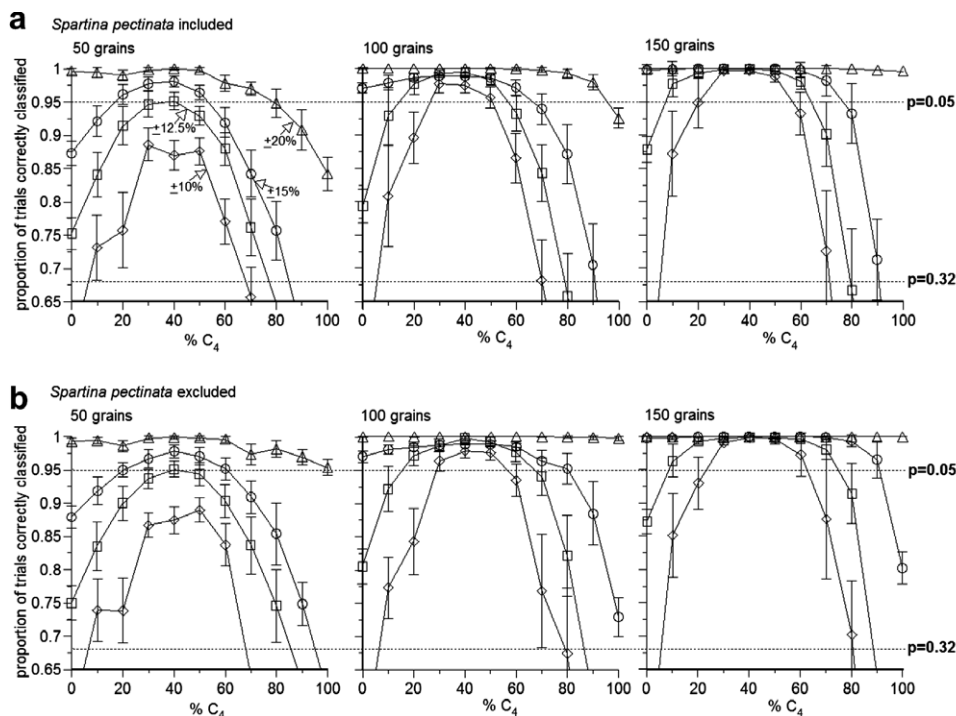


Fig. 6. Proportion of randomized trials in which  $C_4$  pollen abundance of a sample dataset was correctly identified. Each dataset contained 50, 100, or 150 grains randomly selected from the master dataset, as described in Section 3.5. The accuracy of estimated  $C_4$  abundance in relation to the true  $C_4$  proportion of each dataset is shown with different symbols: diamonds, squares, circles, and triangles represent accuracy of  $\pm 10\%$ ,  $\pm 12.5\%$ ,  $\pm 15\%$ , and  $\pm 20\%$ , respectively. Lines for  $1\sigma$  and  $2\sigma$  confidence limits of the classifications are shown.

level if the true abundance of  $C_4$  grass pollen is 0%, but correspondingly only at the 68% ( $1\sigma$ ) confidence level if the true abundance is 100%. This effect is not as large if the data from *S. pectinata* are excluded (Fig. 6b).

To further evaluate our ability to count the number of  $C_3$  and  $C_4$  grass-pollen grains in a sample with an unknown  $C_3$  and  $C_4$  grass proportion, we created and then analyzed two “unknown” samples. The first unknown sample contained 20% pollen from *F. elatior* ( $C_3$ ) and 80% pollen from *S. halepense* ( $C_4$ ) and the second contained 80% pollen from *F. elatior* and 20% pollen from *S. halepense*, but these proportions were unknown to the analyst (a “blind” trial). Of the 67 data points with Vs values greater than the  $2\sigma$  range of blanks from the first unknown (Table 1), 70% are classified as from a  $C_4$  grass, and of the 64 data points from the second unknown, 13% are classified as from a  $C_4$  grass. These estimated  $C_4$  proportions are within the  $\pm 1\sigma$  ranges of the true abundance of  $C_4$  grass pollen in both unknowns, consistent with what would be predicted by our resampling method. The accuracy of the counted abundance of  $C_4$  pollen in the unknowns decreases by 1% if a  $3\sigma$  criterion is used instead of the  $2\sigma$  criterion.

#### 4. DISCUSSION

Our results demonstrate the potential for SWiM-IRMS analysis to help distinguish  $C_3$  from  $C_4$  grass pollen in paleo-records. The number of grains required for  $\delta^{13}\text{C}$ -based evaluation of the composition of  $C_3$  and  $C_4$  grass-pollen grains is much lower with SWiM-IRMS than with

EA-IRMS. A single  $\delta^{13}\text{C}$  analysis of grass pollen using EA-IRMS requires  $\sim 600$  grains, whereas estimates of  $C_3$  and  $C_4$  grass-pollen abundance can be obtained using SWiM-IRMS with as few as 50 samples, each representing a single pollen grain. The one drawback to using the SWiM-IRMS is that, on average,  $\sim 45\%$  of samples applied to the spooling-wire did not yield adequate carbon above the  $\pm 2\sigma$  threshold of “processing” blanks (see Section 3.2). Thus if 50 data points are required to address a specific research question,  $\sim 91$  grains must be applied to the wire. Nonetheless, the time required to obtain reliable  $\delta^{13}\text{C}$  data from pollen grains is still substantially lower with SWiM-IRMS than with EA-IRMS analysis. If  $\sim 60$  grains can be isolated from a sediment sample per hour (Nelson et al., 2006), using the SWiM-IRMS results in a reduction of  $\sim 8$  h of isolation time. Thus the entire process of sample preparation and analysis remains substantially faster with SWiM-IRMS than with EA-IRMS.

In addition to reducing the sample size requirements, the SWiM-IRMS technique also avoids the assumption of end-member constancy that is required when using a two end-member mixing model to decouple the fractions of  $C_3$  and  $C_4$  pollen contained in a set of 600-grain samples (e.g., Nelson et al., 2006). Because of potential variation in end-member  $\delta^{13}\text{C}$  values, changes in the abundance of  $C_4$  grass pollen derived from a mixing model may reflect changes in the abundance of  $C_4$  grass pollen or simply changes in the  $\delta^{13}\text{C}$  value of the  $C_3$  and/or  $C_4$  end-member. Thus detecting small changes in the abundance of  $C_4$  grass pollen is difficult based on bulk samples of grass pollen,



particularly when the overall fraction of C<sub>4</sub> grains is low. In contrast, fluctuations in the fraction of C<sub>4</sub> grass pollen can be detected with higher confidence by counting the number of individual grains using the SWiM-IRMS technique. For example, in a sample of 100 individual grains in which the true abundance of C<sub>4</sub> grass pollen is 20%, the counted abundance of C<sub>4</sub> grass pollen is within 10% of the true abundance at the 68% (1 $\sigma$ ) confidence level, or within 12.5% of the true abundance with 95% (2 $\sigma$ ) confidence (Fig. 6). Thus there is 95% (2 $\sigma$ ) certainty that such a sample contains at least 8% C<sub>4</sub> grass pollen. This example also illustrates another advantage of our new technique over calculating the abundance of C<sub>4</sub> grass pollen using a mixing model: the ability to quantify the uncertainty in the estimated abundance of C<sub>4</sub> grass pollen.

There are numerous potential applications for  $\delta^{13}\text{C}$  analysis of grass pollen using SWiM-IRMS. Recent  $\delta^{13}\text{C}$  evidence from carbonate and organic matter in paleosols suggests that C<sub>4</sub> plants were present in low abundance (<~30%) in North America by 21 Ma (Fox and Koch, 2003), which is broadly consistent with molecular clock evidence for the appearance of C<sub>4</sub> grasses between 25 and 32 Ma (Gaut and Doebley, 1997). However, interpretation of  $\delta^{13}\text{C}$  data from such materials is often difficult because of the potential for a varying C<sub>3</sub> end-member and the possibility that the  $^{13}\text{C}$  content of the carbonates and/or bulk organic matter has been altered after deposition. In addition, previous studies may be confounded by contributions to the substrate being analyzed (e.g., carbonates and organic matter) from non-grass taxa, which predominantly use C<sub>3</sub> photosynthesis. Our method illustrates the potential for using  $\delta^{13}\text{C}$  analysis of individual grains of grass pollen to determine the timing of C<sub>4</sub>-grass evolution with greater confidence than was previously available. Analysis of  $\delta^{13}\text{C}$  values of individual grains of grass pollen could also be used to study C<sub>3</sub> and C<sub>4</sub> grass dynamics after the well-characterized rise to dominance of C<sub>4</sub> plants during the late-Miocene (Cerling et al., 1993). For example, uncertainties in the distribution of tall-, mid- and, short-grass prairie communities in the North American Great Plains during the late-Quaternary could be resolved through knowledge of the abundance of C<sub>3</sub> and C<sub>4</sub> grasses. Finally,  $\delta^{13}\text{C}$  analysis of individual grains of pollen could be used to estimate variability in pollen  $\delta^{13}\text{C}$  and provide information about the biosynthesis of pollen in different species.

One challenge to determining the numbers of C<sub>3</sub> and C<sub>4</sub> grass-pollen grains in an unknown sample is how to account for variation in the threshold value (-19.2‰) used to distinguish C<sub>3</sub> from C<sub>4</sub> data. Such variation may result from changes in the  $^{13}\text{C}$  content and/or partial pressure of atmospheric carbon dioxide. However, our method of distinguishing C<sub>3</sub> and C<sub>4</sub> grass pollen using the SWiM-IRMS should be more robust than the EA-IRMS method (Nelson et al., 2006) against a shift in the threshold value. The bimodal distribution of data points (Fig. 5) means that such a shift should be detected, assuming that at least one representative sample in any given period of the geological record contains an approximately even abundance of C<sub>3</sub> and C<sub>4</sub> grass pollen. If the distribution of individual SWiM-IRMS data points were at all bimodal, the data

could be used to revise the threshold value to one more appropriate to the geologic interval of interest. For time periods when an approximately bimodal distribution is unavailable, the threshold value may be adjusted based on other independent proxy data, such as estimates of the concentration (Pagani et al., 2005) and  $^{13}\text{C}$  composition (Zachos et al., 2001) of atmospheric carbon dioxide.

Improved isotopic estimates of C<sub>4</sub>-grass abundance in the geological record can help elucidate potential global (e.g., atmospheric CO<sub>2</sub> concentrations) or local (e.g., aridity, fire) factors favoring the emergence, expansion, and fluctuations of C<sub>4</sub> grasses (e.g., Cerling et al., 1994; Cerling et al., 1997; Pagani et al., 1999; Pagani et al., 2005; Osborne and Beerling, 2006). For example, the origin of C<sub>4</sub> plants has been hypothesized to have occurred in response to declining atmospheric CO<sub>2</sub> concentrations during the Oligocene (Pagani et al., 2005). Yet, the timing of the origin of C<sub>4</sub> plants is equivocal (Osborne and Beerling, 2006). An evaluation of factors controlling the origin of C<sub>4</sub> plants is best done by considering only one plant life-form (e.g., grasses) (Ehleringer, 2005), and our new method is promising for that purpose. Previous studies have also linked the well-documented expansion of C<sub>4</sub> dominated communities during the late-Miocene to declining atmospheric CO<sub>2</sub> concentrations (e.g., Cerling et al., 1997). The lack of evidence to support this linkage (Pagani et al., 1999) has led others to suggest an important role of fire in the late-Miocene expansion of C<sub>4</sub> dominated ecosystems (Keeley and Rundel, 2005). The fire hypothesis involves the replacement of C<sub>3</sub> woodlands by C<sub>4</sub> grasslands as the result of dry fire-seasons. However, it remains unclear whether fire specifically favored C<sub>4</sub> grasslands *per se* rather than just grasslands, or whether C<sub>4</sub> grasslands replaced C<sub>3</sub> grasslands instead of C<sub>3</sub> woodlands. Analysis of  $\delta^{13}\text{C}$  values of individual grains of grass pollen would help resolve these and other uncertainties about the response of C<sub>3</sub> and C<sub>4</sub> grasses to environmental changes in the geologic record.

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#### REFERENCES

- Bender M. M. (1971) Variations in the  $^{13}\text{C}/^{12}\text{C}$  ratios of plants in relation to the pathway of photosynthetic carbon dioxide fixation. *Phytochemistry* **10**, 1239–1244.
- Brand W. A. and Dobberstein P. (1996) Isotope-ratio-monitoring liquid chromatography mass spectrometry (IRM-LCMS): first

- results from a moving wire interface system. *Isotop. Environ. Health Stud.* **32**, 275–283.
- Cerling T. E. (1999) Paleorecords of C<sub>4</sub> plants and ecosystems. In *C<sub>4</sub> Plant Biology* (eds. R. F. Sage and R. K. Monson). Academic Press, pp. 445–469.
- Cerling T. E., Harris J. M., MacFadden B. J., Leakey M. G., Quade J., Eisenmann V. and Ehleringer J. R. (1997) Global vegetation change through the Miocene/Pliocene boundary. *Nature* **389**, 153–158.
- Cerling T. E., Quade J. and Wang Y. (1994) Expansion and emergence of C<sub>4</sub> plants. *Nature* **371**, 112.
- Cerling T. E., Wang Y. and Quade J. (1993) Expansion of C<sub>4</sub> ecosystems as an indicator of global ecological change in the late Miocene. *Nature* **361**, 344–345.
- Clark J. S., Grimm E. C., Lynch J. and Mueller P. G. (2001) Effects of Holocene climate change on the C<sub>4</sub> grassland/woodland boundary in the Northern Plains, USA. *Ecology* **82**, 620–636.
- Eek K. M., Sessions A. L. and Lies D. P. (2007) Carbon-isotopic analysis of microbial cells sorted by flow cytometry. *Geobiology* **5**, 85–95.
- Ehleringer J. R. (2005) The influence of atmospheric CO<sub>2</sub>, temperature, and water on the abundance of C<sub>3</sub>/C<sub>4</sub> taxa. In *A History of Atmospheric CO<sub>2</sub> and Its Effects on Plants, Animals, and Ecosystems* (eds. J. R. Ehleringer, T. E. Cerling and M. D. Dearing). Springer, pp. 214–231.
- Fægri K., Iversen J., Kaland P. E. and Krzywinski K. (1989) *Textbook of Pollen Analysis*. Wiley.
- Fox D. L. and Koch P. L. (2003) Tertiary history of C<sub>4</sub> biomass in the Great Plains, USA. *Geology* **31**, 809–812.
- Gaut B. S. and Doebley J. F. (1997) DNA sequence evidence for the segmental allotetraploid origin of maize. *Proc. Natl. Acad. Sci. USA* **94**, 6809–6814.
- Hayes J. M. (2001) Fractionation of the isotopes of carbon and hydrogen in biosynthetic processes. In *Stable Isotope Geochemistry, Reviews in Mineralogy and Geochemistry*, vol. 4 (eds. J. W. Valley and D. R. Cole). Mineralogical Society of America, pp. 225–278.
- Huang Y., Street-Perrott F. A., Metcalfe S. E., Brenner M., Moreland M. and Freeman K. H. (2001) Climate change as the dominant control on glacial–interglacial variations in C<sub>3</sub> and C<sub>4</sub> plant abundance. *Science* **293**, 1647–1651.
- Jacobs B. F., Kingston J. D. and Jacobs L. L. (1999) The origin of grass-dominated ecosystems. *Ann. Mo. Bot. Gard.* **86**, 590–643.
- Keeley J. E. and Rundel P. W. (2005) Fire and the Miocene expansion of C<sub>4</sub> grasslands. *Ecol. Lett.* **8**, 683–690.
- Lloyd J. and Farquhar G. D. (1994) <sup>13</sup>C discrimination during CO<sub>2</sub> assimilation by the terrestrial biosphere. *Oecologia* **99**, 201–215.
- Loader N. J. and Hemming D. L. (2000) Preparation of pollen for stable carbon isotope analyses. *Chem. Geol.* **165**, 339–344.
- Nelson D. M., Hu F. S. and Michener R. H. (2006) Stable carbon isotope composition of Poaceae pollen: an assessment for reconstructing C<sub>3</sub> and C<sub>4</sub> grass abundance. *The Holocene* **16**, 819–825.
- Osborne C. P. and Beerling D. J. (2006) Nature's green revolution: the remarkable evolutionary rise of C<sub>4</sub> plants. *Philos. Trans. R. Soc. B* **361**, 173–194.
- Pagani M., Freeman K. H. and Arthur M. A. (1999) Late Miocene atmospheric CO<sub>2</sub> concentrations and the expansion of C<sub>4</sub> grasses. *Science* **285**, 876–879.
- Pagani M., Zachos J. C., Freeman K. H., Tipler B. and Bohaty S. (2005) Marked decline in atmospheric carbon dioxide concentrations during the Paleogene. *Science* **309**, 600–603.
- Sage R. F., Meirong L. and Monson R. K. (1999) The taxonomic distribution of C<sub>4</sub> photosynthesis. In *C<sub>4</sub> Plant Biology* (eds. R. F. Sage and R. K. Monson). Academic Press, pp. 551–584.
- Saugier B. and Roy J. (2000) Estimations of global terrestrial productivity: converging towards a single number. In *Global Terrestrial Productivity: Past, Present, and Future* (eds. J. Roy, B. Saugier and H. A. Mooney). Academic Press, p. 528.
- Sessions A. L., Sylva S. P. and Hayes J. M. (2005) Moving-wire device for carbon isotopic analyses of nanogram quantities of nonvolatile organic carbon. *Anal. Chem.* **77**, 6519–6527.
- Shaw G. and Yeadon A. (1964) Chemical studies on the constitution of some pollen and spore membranes. *Grana Palynologica* **5**, 247–252.
- Smith F. A. and White J. W. C. (2004) Modern calibration of phytolith carbon isotope signatures for C<sub>3</sub>/C<sub>4</sub> palegrassland reconstruction. *Palaeogeogr. Palaeoclimatol. Palaeoecol.* **207**, 277–304.
- Zachos J., Pagani M., Sloan L., Thomas E. and Billups K. (2001) Trends, rhythms, and aberrations in global climate 65 Ma to present. *Science* **292**, 686–693.

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