Sampling past landscapes

Methodological inquiries into the bias problems of recording archaeological surface assemblages

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CHAPTER 3.

BREAKAGE, BIAS AND THE ARCHAEOLOGICAL SURFACE RECORD: ASSESSING THE QUANTIFICATION PROBLEM IN ARCHAEOLOGICAL FIELD SURVEY

ABSTRACT

In the practice of archaeological field survey there is a manifest importance for densities, that is, an abundance of artifacts, often relying on simple counts of objects. However, a well-known issue is variable breakage of pottery that can cause biases in quantitative analysis. Although such issues are generally acknowledged, a direct assessment of breakage and the resulting biases is lacking in research. This paper explores the pros and cons of quantification methods of surface collections in terms of counts or weights, and demonstrates the importance of analyzing weight and breakage as part of an integrative approach.

3.1 INTRODUCTION

In the practice of archaeological field survey, archaeological artifacts lying on the surface are mapped and collected to study them as material correlates of past human activity. Whereas exact strategies depend on theoretical and practical considerations, the aim of archaeological field survey is generally to arrive at a representative sample of the find distributions present in the area. Common is the so-called regional 'offsite' survey where the archaeological record is considered as a spatial continuum and its variable densities are recorded (Foley 1981) by consistently line walking all, or a subsample of, the visible terrain. This usually entails a systematic collection of finds resulting in a sample collection that can then be compared with other archaeological and historical sources (Alcock et al. 1994, Banning 2002, Barker 1995, Bintliff

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The study in this chapter was executed in the context of the Tappino Area Archaeological Project (TAAP), coordinated by Tese D. Stek, University of Groningen and the Koninklijk Nederlands Instituut in Rome. The TAAP is organized by Tese D. Stek, Rogier A.A. Kalkers, Jesús García Sánchez and Jitte Waagen. The work at Colle S. Martino was conducted with permission of the Soprintendenza Archeologia, Belle Arti e Paesaggio del Molise. The fieldwork would have been impossible without the support provided by the Comune di Jelsi, especially Michele Fratino, and our collaborative research centre in Molise: the Centro Didattico di Studi Archeologici di Jelsi (CeDISA), and the team itself: at Colle S. Martino consisting of Tese D. Stek, Rogier A.A. Kalkers, Lennart Kruijer, Koos Mol and Max Caspers in the field, and Lisa Götz, Lucia Lecce, Sheila Cherubini, Filippo Salamone, Jacqueline Splinter and Rogier Kalkers for finds analyses. Particular thanks to readers of the first drafts of this study, that is, Vladimir V. Stissi, Tese D. Stek, Jesús García Sánchez and Rogier A.A. Kalkers.
Two of the most important parameters for analyses of these find distributions are the relative densities of collected fragments and assemblage composition, that is, relative densities of specific artifacts in proportion to each other: what finds do occur, how often, how abundant in relation to other finds and how much more than in other places? The typical products of an archaeological field survey, after a usually elaborate study of potential biases (cf. Given 2004, van Leusen 2002: 4.1–4.20), are therefore geographical information system (GIS) maps displaying relative densities, either absolute or proportional, of all or subsets of the collected finds. These can be total densities, or densities of specific ceramic ware classes (wares), finds that share a chronological phase, shape or functional interpretation, or statistical properties such as richness (variation in wares) and evenness (proportional distribution of different wares). Focal points are often called ‘sites’ or what can more technically be described as concentrations of surface finds likely related to human activity in the past at that location, often associated with buried archaeological strata. However variably designated, for example, abnormal densities above background, points of interest or concentrations, ‘sites’ are generally defined by shifts in proportional abundance of finds in combination with other variables, such as the specific composition of ware types, shapes or functional interpretation of surface samples, as well as an assessment of post-depositional processes and geomorphological context. So-called offsite finds are displayed as general trends in their distribution, again represented as fluctuating densities of find types in various compositions (Bevan and Conolly 2004, Bintliff et al. 1999, Dunnell and Dancey 1983, Foley 1981, Gallant 1986, Waagen 2014). Such maps are either used for direct interpretation, or considered as heuristic tools in an integrated analysis together with other archaeological and historical sources such as excavation data, geomorphological studies, textual evidence, etc. Assessments of these maps are then characterized by spatial pattern analyses, which aim to link the variability in surface distributions recovered by the survey to human activity in the past.

Given the manifest importance of densities, that is, abundance, in the analyses of data generated by archaeological field survey, quantification of finds is pivotal in the interpretative process. Although there has been well-founded criticism on the intensive offsite survey method, the value of the resulting distribution maps and the potential of quantification of the data (cf. Blanton 2001, Fentress 2000, Terrenato 2004), it is still recognized that counting and quantifying in this field are vital for integrative problem-oriented research (Fentress 2000: 50). However, the basic quantification in archaeological field survey often relies on simple counts of objects, which can be problematic. In most arable environments the sheer majority of the collected artifacts consists of broken ceramic sherds, in various degrees of physical deterioration. Various natural and mechanical processes can affect the state of ceramic finds in the subsoil and on the surface (Ammerman 1985, Barton et al. 1999, Taylor 2000, Terrenato 2004, Winther-Jacobsen 2010), one of its properties being fragmentation, also known as breakage or brokenness (Orton et al. 2013: 166–181). Although fragmentation issues are known, and can even reflect useful patterns, variability is introduced by pre-, peri- and post-depositional processes, for example, agricultural activities such as ploughing or levelling. Given this potential for variable states of breakage, relative densities may be affected. Observation of ploughed surfaces suggests higher fragmentation rates of sherds than for unploughed surfaces, attested through observation (Dunnell and Simek 1995: 308) and by statistical size comparison between the two (Palumbo 2015: 86). Variability in ware-dependent fragmentation rates results in similar biases, that is, large pots of wares susceptible to high degrees of fragmentation will likely be overrepresented (Orton et al. 2013: 169, Strack 2011: 24). Whereas such a bias may be overcome by comparing samples only based on a single ware and/or a limited range of vessel sizes, post-depositional biases are not so easy to avoid (Orton et al. 2013: 169). Although these issues are acknowledged by field survey practitioners, and various approaches have been developed (cf. Akkeraz and Collins-Elliot 2017, Coccia and Mattingly 1992, Tol 2012, Winther-Jacobsen 2010), a direct assessment of these quantifiers and the degree to which they can result in biases is often lacking in research and the use of counts is still ubiquitous. This is tricky since such biases potentially affect research outcomes if not dealt with carefully. One may therefore argue that the quantitative fundament of the creation of distribution and site maps is a subject that calls for further examination. This paper explores the pros and cons of quantification methods of surface collections in terms of counts or weights.
Establishing a proxy for abundance, that is, a quantifier, is a topic that has been widely addressed in ceramic studies, and various techniques have been examined (Orton 1975, Orton 1993, Orton 2000, Orton et al. 2013). Most authors stress that it is important for ceramic studies to move beyond mere counts and weights of sherds. There have been many different attempts to find a workable solution to the problem of quantifying ceramic assemblages, for both excavation as well as surface record assemblages (cf. Arcelin and Tuffreau-Libre 1998, Dawson 1971, Egloff 1973, Fulford 1973, Mateo Corredor and Molina Vidal 2016, Orton 1975, Poulin 2013, Py 1991, Strack 2011, Verdan, 2011). An important branch of techniques that have been developed can be grouped under the term ‘estimate of vessels represented’ (EVREP), originally designed for use with closed contexts (Orton 1993: 176). These methods attempt to circumvent the issue of breakage by applying a standardized approach to establish the number of whole vessels that the sherds are likely to represent. Some of the best known of these methods calculate the minimum number of individuals (MNI) and maximum number of individuals (MNA), or an average between the two. These rely on a count of feature sherds, that is, rims, handles, bases or those with other identifiable features, rather than the body sherds, of any given ware. Because their relative number is usually known, these can be counted and attributed to a minimum or a maximum number of pots from which they could stem. An alternative is the aggregate feature count (AFC) which is derived by adding sums of rims, handles and bases. This method assumes equal breakage rates for similar sized and types of vessel, and that all vessels present actually have rims, handles and bases (Strack 2011: 24). The estimated vessel equivalent (EVE) is a non-representational quantifier (e.g. Egloff 1973, Orton 1982: 164–167, Orton 1993: 172). Whereas a non-representational quantifier also strives to arrive at comparable quantities, this is attempted by constructing an abstract representation, that is, there is no claim of reference to actual pots represented. The EVE refers to quantification of a part of a pot for which its proportion to the whole vessel is known, alleviating the problem of variable breakage. Often-used is the rim-EVE, but other EVEs are possible, for example, a weight-EVE is possible in case of highly standardized weights per pot (Baumhoff and Heizer 1959: 309, Raux 1998: 12), a vessel surface-EVE has also been proposed (Byrd and Owens 1997; Hulthén 1974). Whereas the EVE is deemed useful as well as powerful (Orton et al. 2013: 173–174), understanding the abstract mathematical mechanics used to derive a metric that allows statistical analysis, the pie slice, may be challenging. With good reason, its creator aptly calls it “a creature fit for a mathematical zoo” (Orton et al. 2013: 174), and probably the reason it is often not structurally used.

It is important to note that all these quantification solutions perform well in assemblages with a high level of completeness (percentage of the pot being present in the assemblage) and a low level of brokenness, and in most cases require an individual treatment of every single sherd. The obvious problem here for archaeological field survey is the generally worn physical condition of the collected finds, which are most often highly fragmented and incomplete. Feature sherds typically make up a small size of the sample; in the ceramic body collected by the Tappino Area Archaeological Project (TAAP), Molise (central–southern Italy), the percentage of such sherds hovers around 10% (Stek and Waagen 2022). This renders approaches based on feature sherds not very easily applicable, if not impossible, at least for Mediterranean archaeological field survey assemblages (Winther-Jacobsen 2010: 50). To be able to account for all finds, and have sizeable and well–distributed samples, for many research purposes it is imperative to work with the large numbers of non-feature sherds (Strack 2011: 24). Furthermore, the sheer size of the collected samples often inhibits individual assessment of the objects. Using a broad range of quantifiers side by side such as the proposed standard by the Seville protocol (Adroher Auroux et al. 2016) is nigh impossible to implement. Although in Greece individual assessment (e.g. Bintliff et al. 2017, Krijnen et al. accepted/in press) seems to be more often part of the research tradition than in other Mediterranean areas, the above problems make weighing and counting the batches of sherds per ware or feature type the modus operandi.

3.2 A REVIEW OF THE ARGUMENTS

In the scope of archaeological field survey where the bulk of the finds are badly preserved, the question is to what degree is counting or weighing sherds the least biased quantity estimator. The rationale for the comparison is that weight is not affected by breakage and therefore does not suffer from that bias when comparing based on ware types (Orton et al. 2013: 169). If breakage is constant, a choice for using counts
can well be defended for archaeological field survey. It is a straightforward measure that relates to what is picked up in the field and connected to common thresholds for, for example, site identification. Using weight, on the other hand, introduces difficulties on its own, which will be addressed further below.

Previous research indicates preference of weight above counts (e.g. Carrete et al. 1995, Millet 1991, Millett 2000, Orton and Tyers 1993). This is empirically demonstrated by strong correlations between various measures (e.g., numbers, weights, rim counts), so whichever measure was chosen, the trends in the abundance estimates remain similar. Therefore, the choice of a good working parameter can be purely based on ease of implementation; weights of sample batches are by far the fastest to record (Millet 1991). However, details of the tests are omitted, and they are likely not representative of all ware types and contexts, for example, based on wheel-made pottery from a single courtyard (Millett 1979), limiting the degree to which the results can be generalized. Yet others look at the general correlation between the relative abundance of wares per period as expressed in counts or weights and draw similar conclusions (Slane 2003: 325–326). Although they indeed attest such a correlation, they do not specify statistical significance, but more importantly, only very general trends are tested. The issue for the current problem is that taking batches of finds and correlating numbers and weights invariably introduces a smoothing effect; working with total or average weights, variation on an individual sherd basis is masked.

Some studies suggest that breakage rates do show averaging trends. The concept lies at the basis of the modulus of rupture (MR) (Molina Vidal 1997, Mateo Corredor and Molina Vidal 2016), which is an estimate for the average size of a sherd breaking off from a specific vessel. This modulus is then a corrector for breakage that can be applied to counteract differential breakage rates between wares and pot types. However, the assumption of a random breakage process, which is the basis for the MR coefficient, is unwarranted for survey archaeology due to variable post-depositional histories. Another argument for treating counts and numbers as roughly displaying the same trends is that whatever biases are there, they are likely to be consistent (Winther-Jacobsen 2010: 49), an argument that, again, is invalid for archaeological field survey, where conditions of pottery preservation can be different for adjacent fields.

A final argument that prefers counts above weights is the perception that the statistical probability of picking up two sherds from a single pot is negligible, and thus one could use counts as a sort of \textit{MNI} (V. V. Stissi, personal communication). However, in cases of ploughed up deposits such as in a site context, this assertion appears hard to justify. Since buried contexts can show a high degree of completeness, it is statistically very possible that sherds originally did belong to the same pot, or to a larger surviving fragment. Admittedly, such connections between sherds, that is, ‘sherd families’ (Orton et al. 2013: 172), are difficult to detect due to the physical wear of break lines and, again, the usual abundance of finds. One usually does not invest in comparing all possible cases of fitting sherds, though where it has been, such cases have been signaled (Tol 2012: 237–238). Also, as mentioned above, there is statistical evidence pointing towards breakage of surface finds due to ploughing (Palumbo 2015: 86).

\textbf{3.2.2 Weight Issues}

From the above, there is clearly no solid ground for assuming a constant relation between number and weight on a general level, or an uncomplicated argument to prefer counts over weights. The degree to which breakage is a factor influencing the quantitative basis of find densities is still largely one open for investigation. To recount, if breakage is very variable, weight is likely the less biased quantifier (Orton et al. 2013: 169), but weight brings its own specific limitations and potential biases.

The most obvious limitation is that large vessels, at least those with thick walls, will on average feature heavier sherds in comparison with small vessels, rendering in-sample comparisons of proportions skewed. Working with weight means that the individual ware classes must be considered in all comparisons (Orton et al. 2013: 169, Millett 2000: 54). Solutions have been proposed, such as adjusted weight, surface correction and volume displacement (Hinton 1977, Hulthén 1974), but these are practically cumbersome. Average vessel weight, leading to a weight-EVE, would probably overcome most of those problems (Rice 1987: 292); however, a lot of data would be necessary to make this work, and there is the issue of specific weight (Mateo Corredor and Molina Vidal 2016), as mentioned below.
Similarly, a large range of vessel sizes within one ware will influence abundance estimates between samples, that is, if one sample contains sherds of small pots and another of large pots of a single ware class, quantity estimates based on weight will be skewed towards the latter. This is difficult to avoid and very dependent on the range of vessel sizes within one ware. Since wares are often based on fabrics with similar physical properties, such size ranges will not always be extreme, but certainly can be. Other potential biases may be caused by variability of weight within wares for similar vessels. After all, if sherd weight is not a robust proxy for sherd volume, there are similar issues as with using numbers as measures of abundance. Such may be caused by variable specific weight due to differences in the chemical composition of fabrics within a single ware, and because of processes such as overfiring and weathering. Since empirical research into the matter is lacking, albeit some explorations (Kinnunen 2020), little is known about potential impact and solutions. A final issue concerning weights is loss of moisture due to evaporation. Ceramic finds can be assumed to be fully saturated with moisture at the time of collection, after which it will gradually reduce depending on the procedures of finds processing and storage. This has been demonstrated to lead to differences of 10–15% (Slane 2003: 324), though this bias may be mitigated by timing the moment of measurement.

To conclude, where some of the limitations may be overcome, that is, by looking at wares separately, others, such as vessel size ranges and specific weight variability within wares, are more difficult to avoid and must be explicitly assessed. To examine the behavior of count and weight proxies in more detail, this paper continues with an empirical statistical treatment of data from a case study.

3.3 A CASE IN POINT

In the context of the Tappino Area Archaeological Project (TAAP), coordinated by Dr. Tesse D. Stek, University of Groningen and the Koninklijk Nederlands Instituut in Rome, a large-scale methodological test was designed for the investigation of a find complex interpreted as a large rural site, located on Colle S. Martino in the Tappino area (fig. 3.1) (Stek 2018, Stek and Waagen 2022). The site was systematically examined by applying transect survey, where five people systematically collect all finds in a 2 m swathe spaced 10 m apart in units of roughly 50 x 50 m, that is, collecting a 20% sample by line walking (Pelgrom and Stek 2010, Stek and Pelgrom 2005, Waagen 2014).

The relatively high densities (often more than 5/m²), the size of the scatter (about 4 ha), the variable visibility and the spatial patterning of the artifacts made the case ideal for testing the efficiency of the so-called point sampling (PS) technique in relation to transect sampling (TS). PS is a very intensive sampling method in which a relatively small area is cleaned of its vegetation and very thoroughly examined for archaeological artifacts, in our case a circle with a 1 m diameter (fig. 3.2).

Similar techniques, such as shovel-testing or test-pitting, have been experimented with quite a few times (Chadwick and Evans 2000, Kintigh 1988, Krakker et al. 1983, Lightfoot 1989). Although these are comparable with PS in their spatial precision and aims for bias mitigation, PS is not an excavation: it is a surface examination and is relatively effort-efficient as opposed to digging pits. The method carries great potential to map find distributions with high precision avoiding common visibility and observation issues (Stek and Waagen 2022). In order to test the PS method in comparison with regular TS, a grid for PS was laid out over the site, partly overlapping the already surveyed fields (fig. 2). This enabled assessment in various visibility circumstances and densities because they were also placed in fields considered to be outside of the site scatter boundaries. The sample collection resulted in a dataset consisting of 794 PS partly in non-visible terrain, partly overlapping 25 TS and a total of 9255 collected pottery fragments.

Clearly, the methodological study into PS required a good proxy for relative abundance for comparing differences in densities. Moreover, breakage is more prominent because of the increased intensity of the PS method. As opposed to TS, PS allow a very effective alternative for collection in a small area (Stek and Waagen 2022). Consequently, the closer one looks to the ground surface, the smaller the pieces of ceramics that will be recovered. Surface sherds range from big to small, and more intensive sampling techniques, that is, time spent on collection, distance between the observer and the ground, etc., will target more effectively the smaller size ranges. Thus, one should be very careful when comparing counts of different sample types, as the smaller pieces may represent bits broken off larger pieces. Furthermore, the degree of breakage on a
site such as this can be a source of evidence for the state of surface assemblages, and possible relation with buried deposits. Finally, it is known that when completeness is low, it is more likely for pots with a high level of fragmentation to be included in samples (Orton et al. 2013: 169). Especially, in case of the reduced sample area of PS, this could be of notable influence. All these factors render the study of breakage essential for the Colle S. Martino case, and thus formed an opportunity to engage with the problem.
3.3.1 Counts, Weights and Statistics

To examine breakage, the two most abundant wares, coarse wares (CW) and plain wares (PW) were examined to optimize statistical power. These wares constitute the main classes of wheel-made table and (light) utility ware in ancient Samnium. The bulk of these finds dates from the Archaic to Late Roman period, and in the TAAP dataset most of them are found on sites with 4th and 3rd centuries BC black gloss (BG) pottery. The division between CW and PW is made on the presence or absence of larger inclusions in the clay matrix (fig. 3). Although finer data would have been obtained by analyzing finds of narrower chronological and morphological frames, such data unfortunately, except for a small subset of feature sherds, are simply not available. Apart from further identification of finds and groups being a question of effort expenditure, oftentimes the sherds themselves, being heavily corroded, do not allow for much further specification. Nevertheless, there are several indications that the larger share of these finds actually fit a narrower chronological frame, that is, about 60% of all CW and PW finds in the dataset have been collected at 20 sites, of which at least 14 can be dated to the Hellenistic/Samnite phase based on BG pottery. Datable PW feature sherds follow a similar pattern. There are very few CW and PW finds that correlate with Iron Age or Archaic find distributions, and there is otherwise no evidence to suggest a very dissimilar chronological association of CW and PW offsite finds in the research area. It is a reasonable assumption that the majority of the finds under study are from the Hellenistic/Samnite period.

In the following assessment the weight issues mentioned above will be considered. Weight has been established in weighing batches of finds per ware per sample. The issue of variable moisture is not very influential, as all finds have been weighed on the day of collection after washing. As they have come out of the field fully saturated, there is no great deal of moisture flux at this point. Although the effect of variable specific weight will be mitigated due to the far majority of the finds being of local or regional production, and all being wheel-made, this must be accepted as a potential source of some variability. When it comes to vessel size range per ware, these can create noise as well. The question is of course whether such noise will have a large effect. Very small or very large vessels of these wares may occur, but will likely be proportionally rare. Nevertheless, its effect must be estimated and therefore will be addressed further below.

3.3.2 Correlations

A first assessment is testing any correlation between number and weight to see whether or not the two show a similar trend. Taking all the CW from the full set of PS aggregated per TS, as well as the TS themselves, and comparing counts and weights with linear regression, there are strong and statistically significant correlations, respectively $r^2 = .95, p = .000$ and $r^2 = .96, p = .000$, against an $\alpha = .05$ significance level. For the PW in the PS/TS and the TS there is only a similar good match for the latter, that is, $r^2 = .92, p = .000$, but just a moderate correlation, still significant, for the PS/TS $r^2 = .66, p = .004$ (fig. 4) against an $\alpha = .05$ significance level. Regression analysis assesses the degree to which variation in one variable behaves similarly to variation in another, so in the latter case the variation in counts of the sherds for 66% 'explains' the variation in weight of the sherds, which suggests that 34% of variation in weight is not related to differences in counts.

The 34% deviation is clearly attributable to two PS collections: one in TS 2322, a high-density area, and one in 2343, a low-density area. Removing them as 'outliers' would certainly result in a stronger correlation, but there is no evident reason to manipulate the data in this way. Being both collections of 13 PW sherds, but one with a total of 22 g and the other with a total of 117 g, one may wonder whether this is because of different vessel sizes or possibly variable because breakage due to post-depositional processes. Although the finds study did not clarify the issue, striking is a possible bowl fragment of 33 g in unit 2322, which appears rather large. Both possibilities may potentially provide information on a possible difference between the nature of the assemblages, for example, site or offsite. In this way, studying breakage is provoking questions about the surface assemblages and may bring up new ideas about the site and its finds. More generally and importantly, however, it should be noted that the strong correlations all relate to the aggregations, that is, at a level of grouping of finds that in itself already smoothens variability by taking the mean sherd weight. On the level of individual PS, the same PW finds show an actually very weak correlation, and still statistically significant, that is, $r^2 = .14, p = .000$. Therefore, on a general level the correlation appears strong, but this correlation fails zooming in on individual samples, and variability becomes evident.
Figure 3.3. Examples of coarse ware (left) and plain ware (right) pottery from the Tappino area clearly showing variable fragmentation (top); and 1, coarse ware (CW) cooking pot; 2. CW bowl; 3. CW rim; 4, plain ware (PW) base; and 5, CW base (bottom).

Figure 3.4. Regression analyses of plain ware (PW) pottery, point sampling (PS) and transect sampling (TS) combined.
3.3.3 Variability

Whereas correlations allow assessment of trends, it is imperative to have a finer grip on the detail, for which common descriptive statistics can be applied, such as the coefficient of variation (CV). The CV is a descriptive metric that aims to provide information on the spread of a batch of measurements regardless of the sample mean ($\bar{x}$). Where the standard deviation ($s$) expresses an absolute measurement relative to $\bar{x}$, which is difficult to assess without $\bar{x}$ itself, the CV gives the relative dispersion by dividing $s$ by $\bar{x}$, arriving at a ratio, that is, the spread expressed as portion of the mean. This is best illustrated by an example: a 5 g $s$ with $\bar{x} = 10$ g is a lot of dispersion, whereas the 5 g $s$ with $\bar{x} = 100$ g is proportionally little dispersion; whereas $s$ is the same, the CV of the former is 50% and that of the latter is 5%, expressing the difference of variation respective to $\bar{x}$. For assessing spread it is very useful to consider the CV alongside $s$ and $\bar{x}$.

The CV for the PW finds from the PS is 42.72%, for the TS is 69.09%, and the average is 55.90%, which is considerable. To provide a comprehensive impression, this can be translated to the actual mean weight deviation. A total of 1 $s$ from the mean, that is, notionally 66% of the finds, comprises a range of 2.55–6.35 g ($\bar{x} = 4.45$ g, $s = 1.9$ g). In other words, for two-thirds of the finds, the heaviest pieces are 2.5x the weight of the least heavy sherds. For PW found in the TS, the variation soars to 69%, within 1 $s$ ranging from 1.89 to 10.29 g ($\bar{x} = 6.09$ g, $s = 4.2$ g). This points to a considerable weight variation in PW sherds, and potentially indicative of breakage effects.

A difficulty with assessing the potential effect of vessel size range is that there is limited information on that range, and that establishing it based on archaeological field survey finds is nigh impossible. A potential proxy is the distribution of diameter estimates taken from feature sherds. These show quite some variability, and that establishing it based on archaeological field survey finds is nigh impossible. A potential difficulty with assessing the potential effect of vessel size range is that there is limited information on that range, and that the PS collection method gives up smaller finds because of a difference in intensity, obviously due to the ground being examined from a much shorter distance and for a longer time. These sherds might, at least for a part, be more fragmented parts of vessels, and thus counting numbers of sherds potentially overrepresents the total quantity of pottery collected. It is worth mentioning that this conclusion echoes the results from a seeding experiment, referring to the so-called size effect (Odell and Cowan 1987).

Surely, this is a very small sample of a single shape, and it demonstrates the rather obvious: pots break in variably sized sherds independent of vessel size. Nevertheless, it eventually corroborates the notion that weight variability is partly related to breakage and not just to vessel size range. Although it may be hard to disentangle the actual cause for weight variability, all the presented analysis point towards an effect of breakage reflected in weight variability.

The CV weight difference between the PW samples in PS and TS show less variation in the PS. Since regression analysis showed that for PW in PS, an increase in count per PS/TS does only moderately cause an increase in weight, it is evident that PS collect the smaller pieces in the surface assemblage more effectively. The CV demonstrates here that the spread of weight per sherd for PS/TS is much smaller than that for the TS, corroborating the same pattern. The PS sample method gives up smaller finds because of a difference in intensity, obviously due to the ground being examined from a much shorter distance and for a longer time. These sherds might, at least for a part, be more fragmented parts of vessels, and thus counting numbers of sherds potentially overrepresents the total quantity of pottery collected. It is worth mentioning that this conclusion echoes the results from a seeding experiment, referring to the so-called size effect (Odell and Cowan 1987).
The conclusion must then be that in general, there are good reasons to believe that weight variability is at least partly a result of breakage differences. This means that it is reasonable to consider that counts of sherds are a more biased estimation of densities than weight, and that in the case of PS, but more generally in case of more intensive sampling, biases can be expected to be stronger.

### 3.4. Effects on the Spatial Distributions

Whereas an elaborate treatment of the use of weights and counts for the study of Colle S. Martino is not the scope of this paper (Stek and Waagen 2022), a few examples can be given to show that the biases created by using simple counts are not trivial. Plotting weights and counts per PS on the map, the variation is immediately visible (fig. 6).

Evidently, where there is a general spatial correlation between the site boundaries and weight densities in TS 2320 and 2322, there are substantial deviations as well with the majority of the high-weight densities falling outside of the site boundaries (fig. 6). Given the hypothesis that bigger sherds are often related to ploughed up deposits, and those finds will be subsequently spread and further fragmented, this observation could point to site haloes instead of the location of buried remains.

Using either numbers or weights of PW finds in the TS smoothens or sharpens apparent variability in density classified by a natural breaks algorithm (fig. 6). In this particular case, this corroborates the observation made above, but more generally, it is evident that this can give up differential spatial patterning in find distribution maps.

Translating back to total find densities in the TS, in 2320 there is a numeric density of 1.3 sherds/m² with a weight density of 37 g/m², where in 2322 there is a numeric density of .9 sherds/m² with a weight density of 16 g/m². Therefore, where calculated in counts, there is a 1.3/.9 = 1.4 factor difference (FD) (Pettegrew 2014) in density, when calculated in weights there is a 37/16 = 2.3 FD in density. The degree to which relative densities scale up or down for counts and weights is quite different. As a final example, for our comparison between sample types, the PS gave up almost three times as many PW sherds when counting, but only two times as many PW sherds when weighing, evidently the result of picking up smaller finds. Clearly such differences can affect interpretations, as the information of the surface distributions noticeably changes looking at either numbers or weights.

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**Figure 3.5. Diameter distribution for plain ware (PW) feature sherds (left) and recorded metrics for point sampling (PS) bowl shapes (right).**

<table>
<thead>
<tr>
<th>FORM</th>
<th>FRAGMENT</th>
<th>WEIGHT</th>
<th>THICKNESS</th>
<th>DIAMETER</th>
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<td>0.6</td>
<td>12</td>
</tr>
<tr>
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<td>0.7</td>
<td>18</td>
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<tr>
<td>BOWL</td>
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<td>2</td>
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<tr>
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<td>10</td>
<td>0.6</td>
<td>-</td>
</tr>
<tr>
<td>BOWL</td>
<td>r</td>
<td>4</td>
<td>0.5</td>
<td>16</td>
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</table>
3.4 Conclusions

This elaborate assessment has demonstrated the importance of a careful treatment of breakage and the possible biases as a result of it for archaeological field survey. With a weight variation that is partly the result of breakage as variable as suggested in the case study, weight should be taken into account as an estimator for abundance. Weight has its specific problems, and there is the issue of variation caused by vessel size range. To get a good grip on quantitative patterns, it is imperative to assess counts and breakage next to weight, and study spatial patterning in them. The integrative assessment of these basic quantifiers for surface assemblages is imperative because it may also provide information about states of conservation and formation in various places. Furthermore, a more intensive sampling technique results in more and smaller pieces of pottery is a warning for assessing numeric abundance in research designs where various sampling methods are combined.

As said, with weight there are problems such as the aforementioned vessel size range and specific weight, which probably generate noise, or worse, introduce notable error in weight densities. Although such variability can be great due to the broad chronological frame of the coarse- and plain wares, that potential bias is mitigated due to the actual finds under study largely dating to the 4th to 3rd centuries BC. Nevertheless, it is sensible to assess the variability they can exhibit and consider those as error margins on the weight densities. A way forward here, and something really needed, is to design empirical studies targeting vessel size ranges and specific weight for wares to better understand potential influence. However, it is important
for examinations such as these to be able to assess the finds at the level of the individual object. Since this takes a lot of effort, a statistical subsampling design should be applied. Alternatives for weight such as volume are practically cumbersome, though possibly the fast development of three-dimensional recording can tip the balance in favour of such an approach in the near future.


