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Variation matters: Expanding the scope of experimental archaeology using the Perception-Process-Product conceptual framework

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Abstract:	This paper presents the Perception-Process-Product ('Triple P') framework that aims to expand the scope of experimental archaeology. The Triple P framework emphasizes multi-level variation and interactions across the levels of perception, process, and product to provide a more grounded and richer explanation of the past archaeological record. It consists of three principles: 1) acknowledging the inherent trade-off between control and generalizability in naturalistic research design; 2) encouraging collaborative projects that involve geographically diverse and non-traditional research participants such as hobbyists and novices; 3) adopting a workflow that normalizes the collection and curation of ethological and ethnographic data in experimental projects.

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Abstract: This paper presents the Perception-Process-Product (‘Triple P’) framework that aims to expand the scope of experimental archaeology. The Triple P framework emphasizes multi-level variation and interactions across the levels of perception, process, and product to provide a more grounded and richer explanation of the past archaeological record. It consists of three principles: 1) acknowledging the inherent trade-off between control and generalizability in naturalistic research design; 2) encouraging collaborative projects that involve geographically diverse and non-traditional research participants such as hobbyists and novices; 3) adopting a workflow that normalizes the collection and curation of ethological and ethnographic data in experimental projects.

Introduction

This paper presents the Perception-Process-Product (hereafter ‘Triple P’) conceptual framework to expand the scope of experimental archaeology. The field has long tended to adopt the principle of Occam’s razor (e.g., Domínguez-Rodrigo, 2008; Reeves et al., 2009), whether explicitly or implicitly, and center around the reverse engineering of a past technology in a minimal or least-effort manner while ignoring the rich contextual information it affords. Nevertheless, Occam’s razor can be insufficient to infer the preferences of ‘irrational’ agents possessing incomplete information (Mindermann & Armstrong, 2018) in tool design and use, and the two conditions described here provide a better approximation of past humans displaying extensive cultural variation as opposed to the assumption of omniscient *Homo economicus* for most anthropologists (Henrich et al., 2001). In the evolution of technology, it is rather common that opaque causal perception and its resulting tendency of over-imitation (e.g., Sylvia’s Recipe, Gergely & Csibra, 2006) together with path dependence (e.g., QUERTY keyboard design, Kafaee et al., 2022) can lead to the widespread and long-lasting reproduction of technological solutions that are neither optimal in functional efficiency nor minimal in manufacture complexity.

Built upon the *Homo economicus* critique and early works in behavioural archaeology (Schiffer, 2010), here I propose the Triple P framework, which aims to a) amplify the expression of variation in experimental replicas (product) and their associated behavioural channels (process) as well as sensory experiences (perception) by experiments in diverse

contexts and **b)** better identify the complex interacting relationships across these three levels of variations in real-world conditions. To accomplish these two objectives, I advocate the following three principles as integral components of the Triple P framework, which requires **1)** acknowledging the inherent trade-off between control and generalizability in naturalistic research design and **2)** encouraging collaborative projects that involve geographically diverse and non-traditional research participants such as hobbyists and novices. These two principles are developed to advocate a pluralistic approach to the explanation of complex variation, which has received more attention from evolutionary anthropology ([Antón & Kuzawa, 2017](#)) to cognitive science ([Barrett, 2020](#)), instead of treating the optimization-based research agenda as a panacea. The second principle particularly allows researchers to develop research questions that are also meaningful to descendant communities through respectful conversation and collaboration ([Montgomery & Fryer, 2023](#)). The Triple P framework also **3)** adopts a workflow that normalizes the collection and curation of ethological and ethnographic data in experimental projects. It is acknowledged that strategies of data collection and analysis of a given experimental project should be primarily derived from the research question, but the awareness of the rich toolkit available can sometimes inspire researchers to ask questions that are bold and transformative ([Schmidt & Marwick, 2020](#)). Here I will leverage the extensive corpus in experimental designs and inferences revolving around stone artefacts to clarify its meaning and demonstrate the necessity and potential of this framework.

What good is naturalistic experimentation?

The trade-off between causal inference (aka ‘internal validity’) and generalization (aka ‘external validity’) forms a central issue in experimental design across different disciplines ([Eren et al., 2016](#); [Roe & Just, 2009](#): 1266-1267). Even in fields known for their development of rigorous and well-controlled experimental methods such as cognitive psychology and neuroscience, researchers have started to use naturalistic stimuli more frequently and advocate a paradigm shift to semi-controlled experiments due to the generalizability crisis, namely the prevailing mismatch between phenomenon of interest and measured variables in psychological science ([Nastase et al., 2020](#); [Shamay-Tsoory &](#)

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3 59 [Mendelsohn, 2019; Sonkusare et al., 2019; Yarkoni, 2022](#)). In contrast, the past decades
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5 60 have witnessed experimental archaeology's growing research interests focusing on the
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7 61 causal mechanism at the behavioural level in the explanation of material culture variation
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9 62 ([Eren et al., 2016; Lin et al., 2018](#)). In the context of stone artefact replication, one typical
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11 63 research design emphasizing causality over generalizability is the use of knapping
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13 64 machines/robots ([Li et al., 2022; Pflöging et al., 2019](#)), which has helped map out the
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15 65 physical constraints of stone artefact manufacture and use through the identification of
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17 66 causal relationships between input (force, exterior platform angle, platform depth, etc.) and
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19 67 outcome variables (flake size, flake shape, wear formation, etc.). All variables of interest in
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21 68 this setting are relatively easy to measure, quantify, and control, but this type of design can
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23 69 be insufficient in inferring how context-generic principles interact in a particular context as
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25 70 reflected in real-world conditions. Similarly, standardized artificial materials like bricks
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27 71 ([Lombao et al., 2017](#)) or foam blocks ([Schillinger et al., 2016](#)) have been used to
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29 72 standardize materials and/or reduce learning demands in experimental studies focusing on
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31 73 the transmission of lithic technologies, with implications for the generalizability of results
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33 74 ([Liu et al., 2023](#)). In real-world knapping, each rock has a different shape and often
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35 75 different physical properties such as inner cracks and inclusions, and this heterogeneity
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37 76 itself represents a critical variable in cultural transmission and skill development ([Proffitt
38 et al., 2022](#)).

37
38 78 On the other hand, naturalistic experiments pay more attention to how experimental
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40 79 insights can be generalized to archaeological samples by incorporating authentic materials
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42 80 and plausible social settings with a certain degree of compromised control ([Outram, 2008](#)).
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44 81 Back to the cases of cultural transmission experiments, a naturalistic experiment would
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46 82 involve the use of natural rocks with varied morphology instead of standardized artificial
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48 83 materials as well as human demonstrators instead of videos of knapping instruction,
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50 84 despite the fact that the latter will remain consistent across individuals. Unlike controlled
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52 85 experiments, variation could be easily observed in naturalistic experiments by design. This
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54 86 feature is crucial and cannot be simply replaced by ethnographic records, because many
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56 87 palaeolithic technological components do not have analogues in contemporary non-
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58 88 industrial societies (e.g., [Arthur, 2018; Stout, 2002](#)). While uncontrolled variation has

traditionally been viewed as highly problematic, statistical techniques for developing causal inference from observational data, of the kind produced by naturalistic experiments, have also been greatly boosted in epidemiology and economics in recent years (Cunningham, 2021; Hernan & Robins, 2023). Naturalistic experiment can serve a heuristic role in hypothesis generation, aligning with the perspective of Lin et al. (2018: 680-681), who proposed that the interaction between naturalistic and controlled experiment “operates in a cyclical form of induction and deduction.”

Many places, many voices

Contemporary practices in experimental archaeology, as manifested by the fact that a majority of scholarly publications are produced as results of experiments conducted by a single knapper with a dual identity of researcher (Whittaker, 2004), tend to be restrained by the cognitive bias known as the ‘curse of knowledge’ or ‘curse of expertise’. This psychological term originally refers to the phenomenon that it is extremely challenging for experts to ignore the information that is held by them but not others, particularly novices (Hinds, 1999), but it has further implications for the sample representativeness in experimental archaeology. When the knapping expertise is gradually formed through multiple years of observations and trial-and-error learning, an expert knapper develops some specific ways of strategic planning, motor habits (and their associated impacts on anatomical forms like wrist and elbow), preferences of percussor and raw material types, as well as familiarity of various techniques that become unforgettable (Moore, 2020: 654). The existence of this cognitive bias is not inherently bad, and these many years of experience should be appreciated and celebrated by experimental archaeologists. However, what is problematic is that the results of replication experiments conducted by these experienced practitioners, often in settings of single knapper, has been constantly framed as generalization regarding the evolution of technology and cognition that masks a vast range of technological diversity.

Modern flintknapping techniques, as a research subject and a scientific method, originated from hobbyists’ individualistic trials of reverse engineering during the 19th century (Coles,

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3 117 1979; Flenniken, 1984; Johnson, 1978; Whittaker, 1994: 54-61). Hobbyist knappers
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5 118 represent a huge repertoire of technological knowledge that does not fully overlap with
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7 119 what is acquired by academic knappers. They tend to generate ideas that may appear to be
8
9 120 counter-intuitive at first glance for academics. One such example is the utility of obtuse
10
11 121 edge angle as demonstrated by Don Crabtree (1977), a mostly self-educated flintknapper
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13 122 yet one of the most important figures in experimental archaeology. In his experiment,
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15 123 Crabtree demonstrated the excellent performance of blade dorsal ridge on tasks like
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17 124 shaving and cutting hard materials, challenging the traditional perspective on producing
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19 125 sharp lateral edges as the sole purpose of stone toolmaking and shedding light on future
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21 126 functional reconstruction through the use-wear analysis. It is rather unfortunate that
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23 127 collaborations between academics and hobbyists are less common than expected due to
24
25 128 their complicated and uneasy relationships as detailed in Whittaker's (2004) ethnography.
26
27 129 Likewise, novices' lack of expertise also helps to mitigate the 'curse of knowledge' bias that
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29 130 may hinder expert knappers. Their involvement can potentially lead to the discovery of
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31 131 alternative methods, techniques, and interpretations that may have been overlooked by
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33 132 experts.

34
35 133 Emphasizing variation at its core, the Triple P conceptual framework recognizes that
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37 134 experimental archaeology can greatly benefit from diverse perspectives (Pargeter et al.,
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39 135 2023: 164) and thereby inherently adopts a collaborative mode of knowledge production,
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41 136 which has been recently advocated in experimental studies (Liu & Stout, 2023; Ranhorn et
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43 137 al., 2020) and museum collection studies (Timbrell, 2023) of stone artefacts. Furthermore,
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45 138 the Triple P framework acknowledges that communities living in specific geographical
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47 139 areas possess unique insights and understanding of their cultural heritage. This emphasis
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49 140 on team efforts and inclusivity allows for a more complete understanding of the non-
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51 141 utilitarian or unexpected aspects of raw material selection (Arthur, 2021), pre-treatment
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53 142 (Maloney & Street, 2020), production (Griffin et al., 2013), and use (Martellotta et al., 2022)
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55 143 across different regions. Through ethical collaborations with those knapping practitioners
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57 144 in non-industrial societies in the research process, the framework allows their voices to be
58
59 145 heard and their contributions to be acknowledged. This not only enhances the quality of
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146 research outcomes but also fosters a sense of ownership and pride within these

communities, strengthening the connection between archaeological research and the people it directly affects (Montgomery & Fryer, 2023).

The Triple P framework in action

The implementation of the Triple P framework involves the collection of process-level (ethological) and perception-level (ethnographic) data (**Fig. 1**), which is critical to address equifinality and multifinality (Nami, 2010; Premo, 2010), two daunting challenges in archaeological inference. Equifinality refers to situations in which a similar state or consequence can be achieved through multiple different paths, while multifinality emerges when a similar process can lead to multiple ends. While we cannot fully solve these two problems and accurately reconstruct the past behavioural processes simply based on materials remains, context-rich experiments involving the collection of ethological and ethnographic data can help us better document an enlarged range of possible combinations of variation and draw a more informed inference (Reynolds, 1999). The importance of specifying and documenting the context information of both the experiment and the phenomenon of interest has also been recently highlighted in psychological sciences (Holleman et al., 2020).

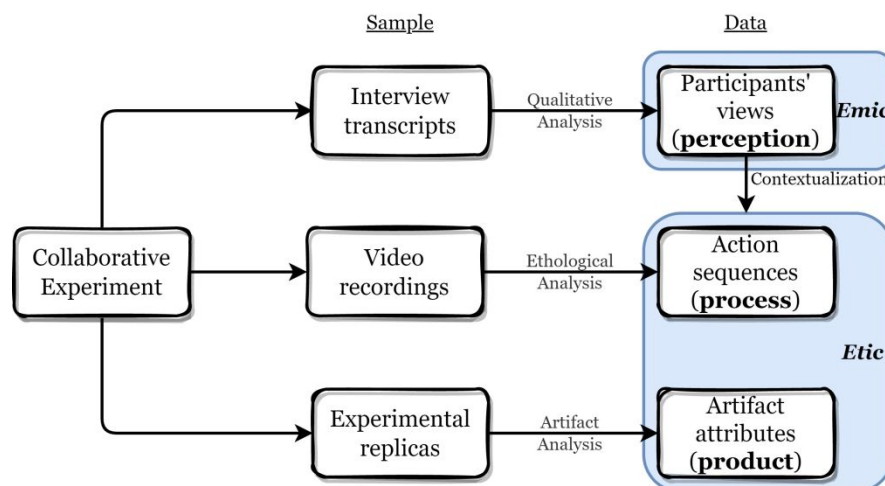


Fig.1. A schematic diagram demonstrating how to operationalize the Perception-Process-Product conceptual framework.

Product-level data

Traditionally speaking, the product-level data, namely the documentation and analysis of replicas, form the sole research subject of experimental archaeology and serve as the tangible foundation for analogical inference in the interpretation of archaeological materials. It can exist in the form of spreadsheets containing detailed technological attributes, photos and illustrations, or high-resolution 3D scans of individual artefacts or a whole assemblage. No particular modification regarding the collection procedure of product-level data is required in the context of the Triple P framework, although the definition of variables measured and the documentation techniques (models of camera/scanners, light setting, processing software version, etc.) should be always available in the relevant meta-data. I also strongly recommend adopting good habits in spreadsheet data organization (Broman & Woo, 2018).

Process-level data

While systematic behavioural coding methods widely used in the study of non-human animal behaviour (Fragaszy & Mangalam, 2018) are still largely neglected among archaeologists, attempts to reconstruct behavioural sequences involved in the manufacture of material remains are not infrequent. One such example is the cognigram, which was first systematically developed and applied in archaeological research by Haidle (Haidle, 2009, 2014). A cognigram is a graphical representation of the reconstructed behaviour behind archaeological artefacts in chronological order of appearance (Haidle, 2014), which essentially represents an abstracting process of a series of action sequences achieving a similar goal. This approach provides an elegant descriptive methodology yet is limited by its normative and analytical orientation, meaning it cannot handle variation very well. To some extent, it describes the minimal steps to achieve a goal from the perspective of reverse engineering and reflects the analyst's own causal perception. However, this may be biased because 1) certain causal insights in stone fracture mechanics remain opaque to academic knappers until they are revealed through controlled experiments by Dibble and his colleagues (Li et al., 2022) 2) ethnographic studies demonstrated that expert non-

academic practitioners can have a different set of causal understanding (Harris et al., 2021).

Consequently, we need to accumulate more real-world data by recording a large number of toolmaking videos and conducting systematic ethogram analysis. With the emergence of new software platforms such as BORIS (Friard & Gamba, 2016), the difficulty of coding has decreased significantly in recent years (Fig. 2). Here I use a modified version of action grammar developed by (Stout et al., 2021) as an example, among multiple coding schemes featuring different research focus (Muller et al., 2023) or granularity (Cueva-Temprana et al., 2019; Mahaney, 2014; Roux & David, 2005). The knapping action recorded in videos can be coded following the ethogram presented in Table 1. Depending on the original research question, sequences of coded actions can then be used in further analysis, such as complexity (Stout et al., 2021), similarity (Mobbs et al., 2021), etc.

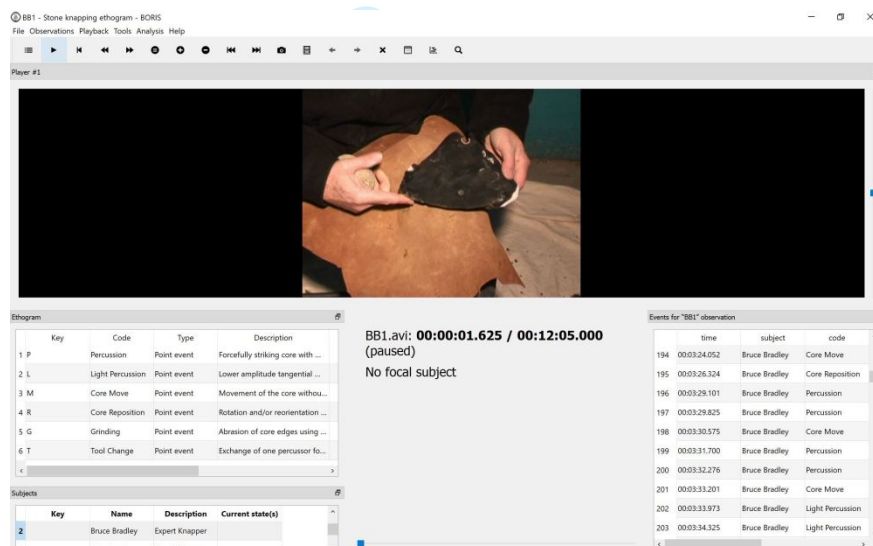


Fig.2. An example of coding a handaxe knapping session using the BORIS software.

211 *Table 1. A modified version of the original action grammar presented in (Stout et al., 2021)*

Action	Definition
Percussion	Forcefully striking core with percussor (hammerstone or antler billet) in such a way as to potentially remove a flake
Light Percussion	Lower amplitude tangential strike to the tool edge of the kind often employed for platform preparation
Core Move	Movement of the core without a change in grip.
Core Reposition	Rotation and/or reorientation of the core involving repositioning of the hand.
Grinding	Abrasion of core edges using a hammerstone
Tool Change	Exchange of one percussor for another

212 **Perception-level data**

213 Ethnographies revolving around experimental archaeology as a field (Reeves Flores, 2012),
214 as well as practices of specific technologies like flintknapping, including contemporary U.S.
215 hobbyists (Whittaker, 2004) and knapping practitioners in various non-industrial societies
216 (Arthur, 2018; Stout, 2002), are far from novel. However, ethnography has never been
217 formally recognized as a legitimate research method in experimental archaeology. Echoing
218 with the recent trends of adopting embodied cognition (Varela et al., 2017) in
219 archaeological research (Malafouris, 2013), ethnographic data and methods can reveal
220 hidden information (e.g., intention, phenomenology) that is otherwise irretrievable and
221 thus should occupy a unique niche in experimental archaeology. This also echoes the post-
222 positivist turn in psychology in the past decades, particularly the emphasis on the value of
223 incorporating qualitative research (Stout, 2021; Syed & McLean, 2022; Weger et al., 2019).
224 Through participant observation, interviews, and detailed field notes, ethnography can
225 capture the subtle nuances of perception, such as sensory experiences, social interactions,
226 and cultural meanings associated with the experimental activities (Gowlland, 2019).
227 Compared with the ethological methods, the interview questions and participant
228 observation in ethnographic methods feature an even higher degree of freedom and rely

more heavily on the research question as well as ad-hoc interaction. One potential application of ethnographic methods in experimental archaeology of stone artefacts is asking knappers about the intentions of each action and see how it matches with the results as revealed by lithic analysis of replicas, which can provide crucial contextual information addressing the issues of equifinality and multifinality in the formation of lithic assemblage.

Multi-level data curation

The comparative study and large-scale synthesis of variation data require the building of centralized, open-access, and carefully curated data infrastructure, which unfortunately still does not exist yet in experimental archaeology. Among the three dimensions of the Triple P framework, the product-level data are usually stored in the format of spreadsheets, photos, and 3D models, and the perception-level data formats mainly include audio files and their transcribed texts, whereas videos are the main vector of process-level data, a rather non-traditional data format in archaeological research featuring the highest file size compared with the other two. As such, following data sharing principles of FAIR (Wilkinson et al., 2016) and CARE (Carroll et al., 2020), the Triple P framework recommends Databrary (Simon et al., 2015), a web-based library originally designed for developmental scientists, as the main data curation platform, where researchers can freely upload video files and related metadata that can connect with different types of data within the same project.

Conclusion

Through the broadening of traditional data types and recording methods revolving around experimental replicas *per se*, the Triple P conceptual framework allows the amplified multiscale expression of material cultural variation. It is also compatible with many theoretical orientations, ranging from behavioural archaeology (emphasis on video recording of behavioural processes) through evolutionary archaeology (emphasis on the amplification of variation) to post-processual archaeology (emphasis on perception

through ethnography). In terms of its research practice, it embraces a collaborative mode of knowledge production by involving a more diverse pool of stakeholders. The innovativeness, flexibility, and inclusiveness of the Triple P conceptual framework has enormous potential in redefining what can be and what should be studied by experimental archaeology as a field and thereby contributing to a better understanding of our deep past.

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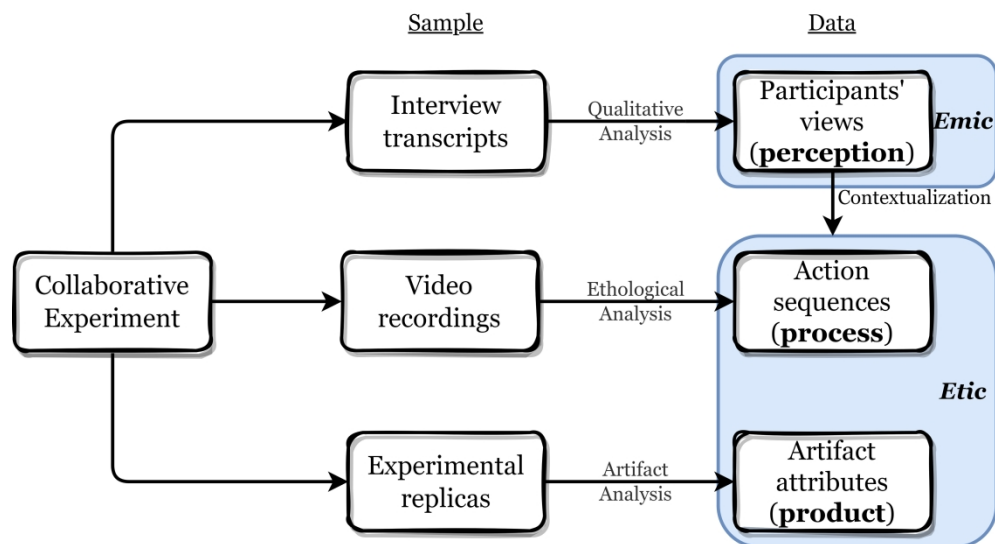


Fig.1. A schematic diagram demonstrating how to operationalize the Perception-Process-Product conceptual framework.

395x227mm (300 x 300 DPI)

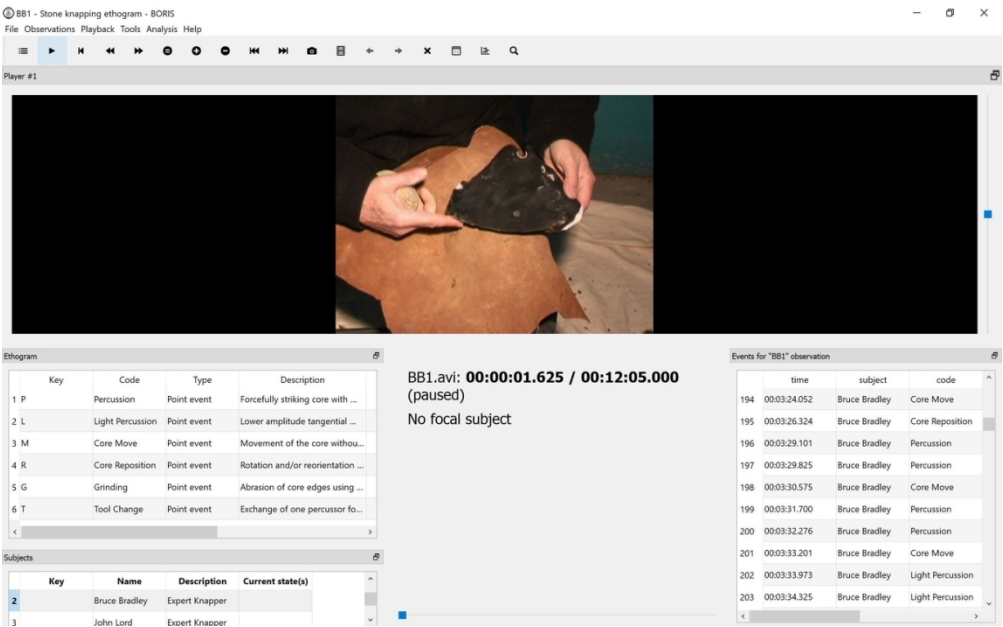


Fig.2. An example of coding a handaxe knapping session using the BORIS software.

115x71mm (300 x 300 DPI)