

Geometric Morphometric Analysis of Projectile Points from the Southwest United States

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Abstract

Traditional analyses of projectile points often use visual identification, the presence or absence of discrete characteristics, or linear measurements and angles to classify points into distinct types. Geometric morphometrics provides additional tools for analyzing, visualizing, and comparing projectile point morphology utilizing the whole or parts of the form in either two or three dimensions. This study is an analysis of the effectiveness of geometric morphometric methods for identifying technological similarity in 2D projectile point outlines for previously classified late prehistoric projectile points found in the U.S. Southwest and unclassified projectile points from Tonto Basin, Arizona. Various methods from geometric morphometrics were compared to determine which method best reproduced the original classification scheme. Elliptical Fourier analysis was compared with various configurations of semilandmark and landmark analyses using generalized Procrustes analysis. These methods were applied to the complete projectile point form, and the landmark analysis was also applied to half of the lower quadrant of the projectile point—essentially one corner of the projectile point. The landmark analysis applied to the corner of the projectile point provided the best results. This method was then applied to the Tonto Basin points. Hierarchical clustering was used on the Tonto Basin projectile point morphometric data to explore the variation in shapes between sites. To demonstrate that geometric morphometric methods can be used without relying on typologies, a network analysis of the morphometric distances was also conducted. This network graph produced distinct clusters of technological similarity in projectile point outlines, while also showing the continuous variation between points. These results demonstrate the effectiveness of geometric morphometrics for the 2D analysis of late prehistoric arrow points.

Keywords: American Southwest, Hohokam, Arizona, projectile points, lithics, computational archaeology, geometric morphometrics

23 **Introduction**

24 Geometric morphometrics (GM) is a quantitative approach to studying shape in two or three dimensions
25 that has recently been adopted in archaeology (see MacLeod, 2017; Okumura and Araujo, 2019; Shott and
26 Trail, 2010, for overviews). It has numerous advantages over traditional lithic analyses, particularly because
27 it can overcome the reliance on linear dimensions (Shott and Trail, 2010, pp. 196–197). Lithic artifacts can
28 be assigned to typologies or directly compared without the use of a typology, as will be demonstrated in this
29 chapter. There are several approaches within GM that provide similar results through different methods. One
30 of the more traditional approaches is to place landmarks at homologous locations around the object. Land-
31 marks can be augmented with semilandmarks, which are points placed relative to another using a consistent
32 rule—usually equidistant spacing between two points (Okumura and Araujo, 2019, pp. 2–4). Another common
33 approach is to use elliptical Fourier analysis to compare the outlines of objects. Each method has strengths
34 and weaknesses. A major purpose of this study is to evaluate the effectiveness of these methods for analyzing
35 projectile points in the U.S. Southwest during the late prehistoric period (after the introduction of the bow and
36 arrow).

37 Once the method of analyzing the points has been determined, the next step is to determine how to com-
38 pare points using the results of the analysis. One approach would be to use an existing regional typology
39 and to assign projectile points to the closest match. Another approach, would be to use cluster analysis to
40 assign points to newly created types. The final approach would be to ignore typologies and compare the
41 morphometric distance for each projectile point directly.

42 Regional analyses are fundamental parts of archaeology, but there are many challenges to overcome. One
43 of these challenges is harmonizing the different categorization schemes (i.e., ontologies) used throughout
44 the region. Another of these challenges, is determining whether the current categories are useful. The U.S.
45 Southwest has a long history of regional ceramic typologies (e.g., Colton, 1956; W Gladwin and HS Gladwin,
46 1930; Hargrave, 1932; Kidder, 1915; Martin and Willis, 1940), but there are still disagreements, challenges,
47 and competing definitions (Duff, 1996). Regional analyses in the Southwest, based in large part on pottery,
48 have produced many useful insights (e.g., Bernardini, 2005; Clark et al., 2019); Hegmon et al. (2016); Mills
49 et al. (2013); Peeples (2018)]. However, one type of material culture that has received little attention—in the
50 Southwest at least—is lithics (i.e., chipped stone). Projectile points are commonly discussed during the archaic
51 period of the Southwest, and they are common topics in many other areas of the North American continent
52 and world where they are found, but they are rarely discussed after the appearance of pottery.

53 Despite the over-emphasis on pottery in the Southwest, there are some excellent resources on projectile
54 point typologies (e.g., Hoffman, 1997; Justice, 2002; Loendorf and Rice, 2004); Sliva (2006)]. However, ad hoc
55 approaches are common, and these cannot be extrapolated beyond a specific project. Even using existing
56 resources can make comparisons difficult. How does Tagg's (1994, p.111) Type 23 compare to Sliva's (Sliva,
57 2006, p. 35) Cohonina Side-notched? There is an answer, but often it is easier to come up with a new typology
58 schema than to try to harmonize existing work.

59 Another challenge that is not unique to projectile points is that interpretations may differ between analysts.
60 Exactly when does a base begin curving enough to be called basal notched? Even the difference between a
61 side-notched and a corner-notched point can, at times, be ambiguous. Not to mention the frustrating situation
62 where a point appears to have one corner-notch and one side-notch. How should one place this point into
63 an existing typology? These are questions that can be handled in different ways that differ from analyst to
64 analyst. Idiosyncrasies and biases are impossible to be rid of entirely, but using approaches such as the those
65 described here can reduce them and increase the reproducibility of the process.

66 By necessity, this chapter covers a number of topics. The geographic area is the U.S. Southwest, but the
67 methods and analysis are applicable to any area. The code and data used are available for reuse and modifi-
68 cation. The primary purpose was to explore geometric morphometric methods using previously typed spec-
69 imens from the Southwest and untyped specimens from the Tonto Basin. As mentioned, the corner-based
70 landmark analysis proved most successful. Another purpose was to analyze the results with and without using

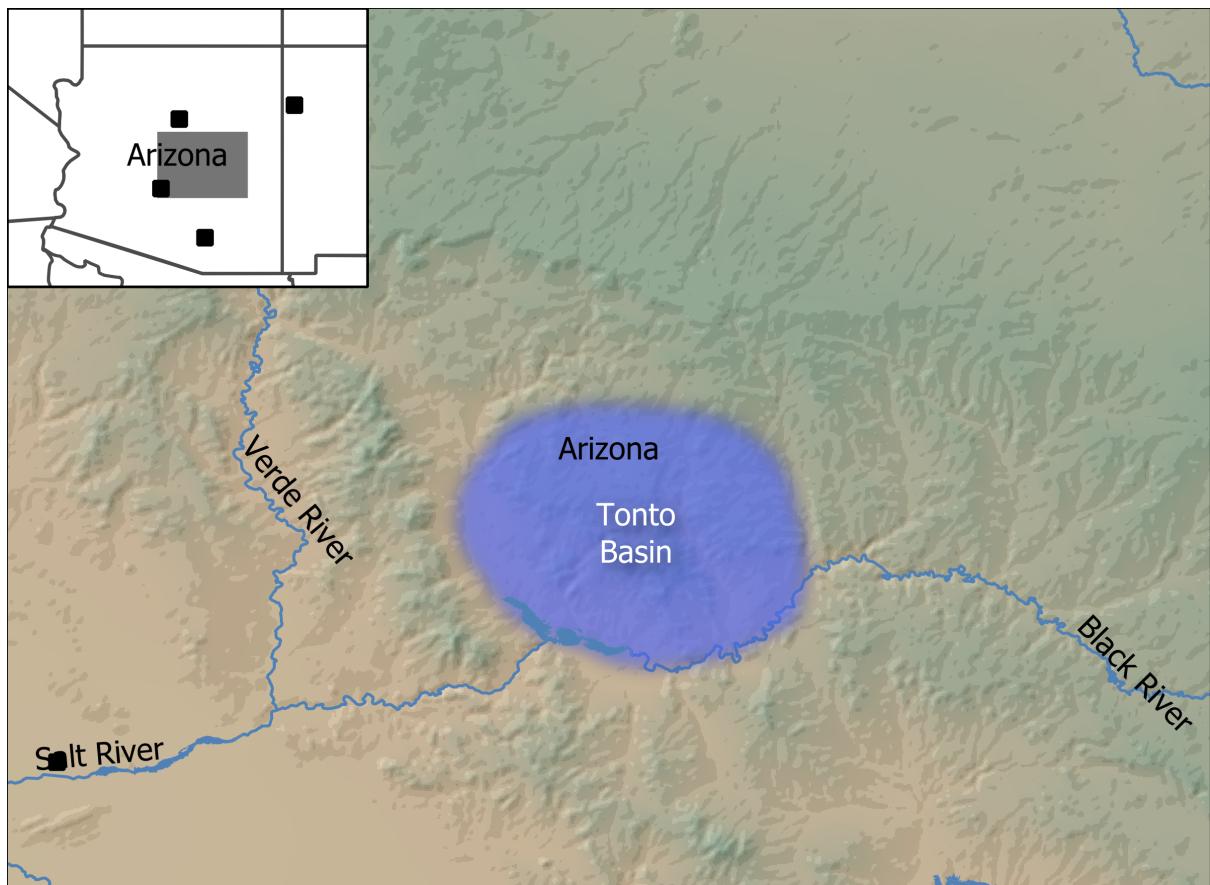


Figure 1. Location of Tonto Basin in the state of Arizona, United States.

71 typologies. The results demonstrate that both approaches are useful.

72 **Background**

73 In order to test the effectiveness of geometric morphometric methods, I needed a dataset of well-typed points
 74 that could be used as a validation set. I chose to use the typology published by Noel Justice (2002) for the
 75 simple reason that it is easily accessible and contains numerous illustrations. These illustrations were used as
 76 type specimens to compare projectile points from Tonto Basin in central Arizona (Figure 1). These points were
 77 excavated in a series of large cultural resource management projects necessitated by work on the Roosevelt
 78 Dam. The largest project—the Roosevelt Platform Mound Study—included 129 sites. Most of the sites date
 79 between AD 1275 and 1325 with occupation continuing until around AD 1450 (Rice, 1998). In the original
 80 analysis, Projectile points were classified according to small and large points and then subdivided based on
 81 morphological characteristics (Rice, 1994, p.727). The typology used is an excellent demonstration of the
 82 difficulty in conducting projectile point studies in this area, as the typology is idiosyncratic to this specific
 83 project, and cannot be easily compared with other datasets. This is not a criticism of the analyst's choice to
 84 create a new typology, as no existing typology met the needs of the researchers.

85 This is an exploratory analysis designed to minimize the amount of time spent collecting data and to be as
 86 reproducible as possible. There are a number of research steps that are often not addressed in publications.
 87 This missing documentation can make reproducing results challenging. I will describe the rational for relevant
 88 decisions, but the script used to generate the analysis will be included in an RMarkdown document in the
 89 supplemental material (see statement at end of manuscript).

90 One of the key elements of this study is reproducibility, which necessitates automation. Projectile point

91 analysis often includes assigning a point to a type based on linear metrics—sometimes angular measurements
92 as well, and the presence or absence of various features (e.g., concave base, serrated blades, corner-notches).
93 But often, the analyst is left to visually compare the point to various type specimens to identify the closest
94 match. In my experience, this can be a frustrating way to spend your time. This method is harder to reproduce
95 and subject to greater human error. Yet, algorithms are only part of the answer, and human judgment and
96 context are still critical to any analysis. The key is to minimize the possibilities for error and maximize the
97 opportunities for reproducibility, which I have tried to do here. Thus, one of the key questions of this research
98 is to determine what input should be left to the analyst and what can be left to automated or standardized
99 procedures.

100 **Data collection**

101 This study has two sources of data: illustrations of projectile points published by Justice (2002) and images of
102 projectile points from collection held at Arizona State University. The datasets include 74 illustrations from
103 Justice's publication and 90 projectile points from Tonto Basin. Many of Justice's types could not be included
104 because there were so few complete, illustrated examples.

105 It is worthwhile to question how an illustration compares to an image of a physical projectile point obtained
106 from a flatbed scanner. Fortunately, illustrations have been published for some of the projectile points in this
107 study. Figure 2 is a comparison of outlines created from an illustration and scan of the same projectile point.
108 There are subtle differences between the two mediums—the base is slightly more rounded in places than the
109 scan. These differences are detectable in a morphometrics analysis, however, the differences are minor as
110 seen in figure 3. The quality of projectile point illustration can vary, but from this brief comparison there
111 should be no hesitation using illustrations for 2D morphometric analysis.

112 Justice's projectile point illustrations were scanned, and the illustrations were converted into individual,
113 solid black outlines and saved as jpeg files using common image-editing software. The open source statistical
114 software R was used for all analyses (R Core Team, 2022). The Momocs package (Bonhomme et al., 2014)
115 has an import function to convert jpeg files into outlines. This is a major advantage over manual outlining
116 processes used in popular GM software, such as tpsDig (James Rohlf, 2015). These outlines form the basis of the geometric morphometric analyses conducted here with the exception of the landmark analyses.
117 Landmarking was performed using the tpsDig software. The Momocs package has that capability, but I have
118 found tpsDig's utility for landmarking to be superior. The Tonto Basin projectile point images were created
119 using a flatbed scanner at 1200 DPI¹. The images were converted to outlines using the same process as the
120 projectile point illustrations.

122 **Projectile points of the Southwest**

123 There are a few regional typologies for projectile points in the Southwest: common examples were authored
124 by Hoffman (1997), Justice (2002), Loendorf and Rice (2004), and Sliva (2006). Altogether, these four typologies
125 include 129 projectile point types, although many overlap. In some cases, the authors identify correlates of the
126 types. Many types date to the Archaic period, and thus predate the primary period we are interested in. Not
127 all of the points were ascribed dates by the authors. Justice lists 23 points that overlap with the AD 1100–1500
128 period (the maximal dates for the Hohokam Classic period). Projectile points have restricted geographical
129 boundaries, although these boundaries correspond to much greater areas than ceramic types typically do
130 (Buchanan, Hamilton, et al., 2019). Figure 4 shows that several projectile point boundaries defined by Justice
131 overlap with the Tonto Basin.

132 For this study, I digitized 74 projectile point images from Justice's publication representing 8 projectile point
133 types (table 1). Justice placed each projectile point type into a cluster of related points. Figure 5 shows the

¹Many of the images were obtained by the author but some were generously contributed by Josh Watts—see the results of his study here: (Watts, 2013).

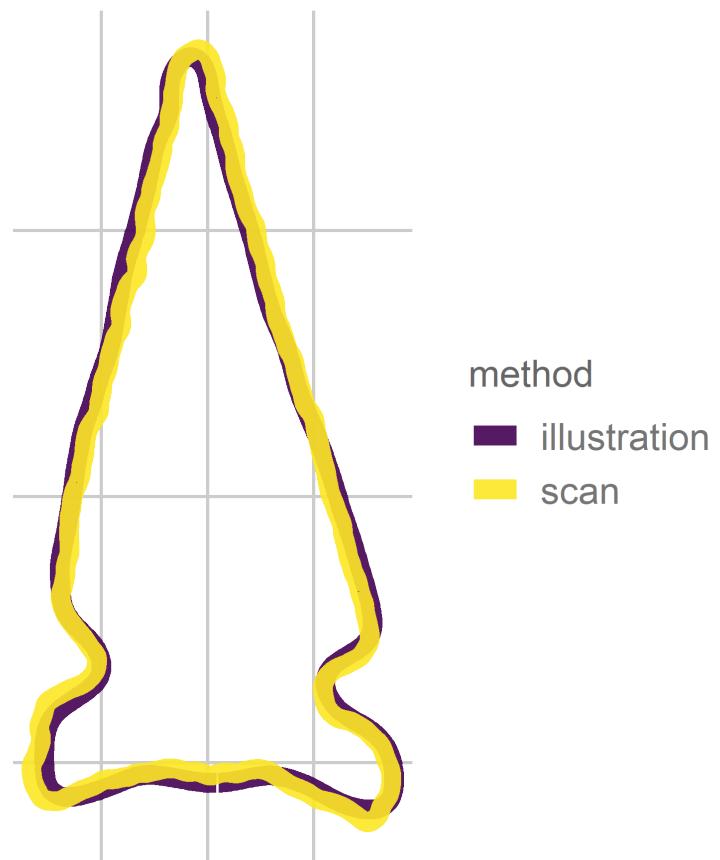


Figure 2. Comparison of projectile point outlines for an illustration and a scan of the same projectile point
(Oliver and Simon 1997: Figure 9.3; Specimen 33598)

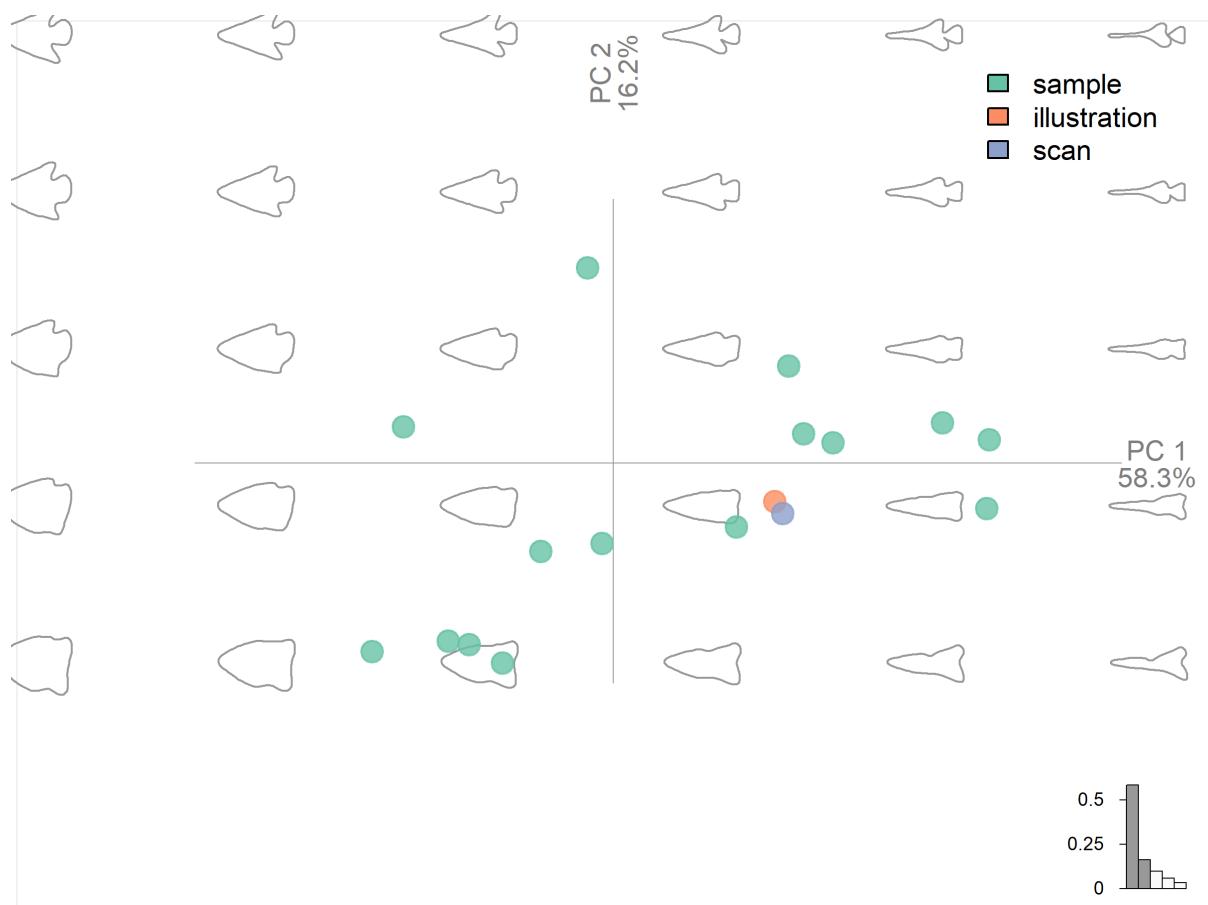


Figure 3. Principal component plot comparing the morphometric differences between a sample of 20 random projectile points and the illustration and scan of the same projectile point.

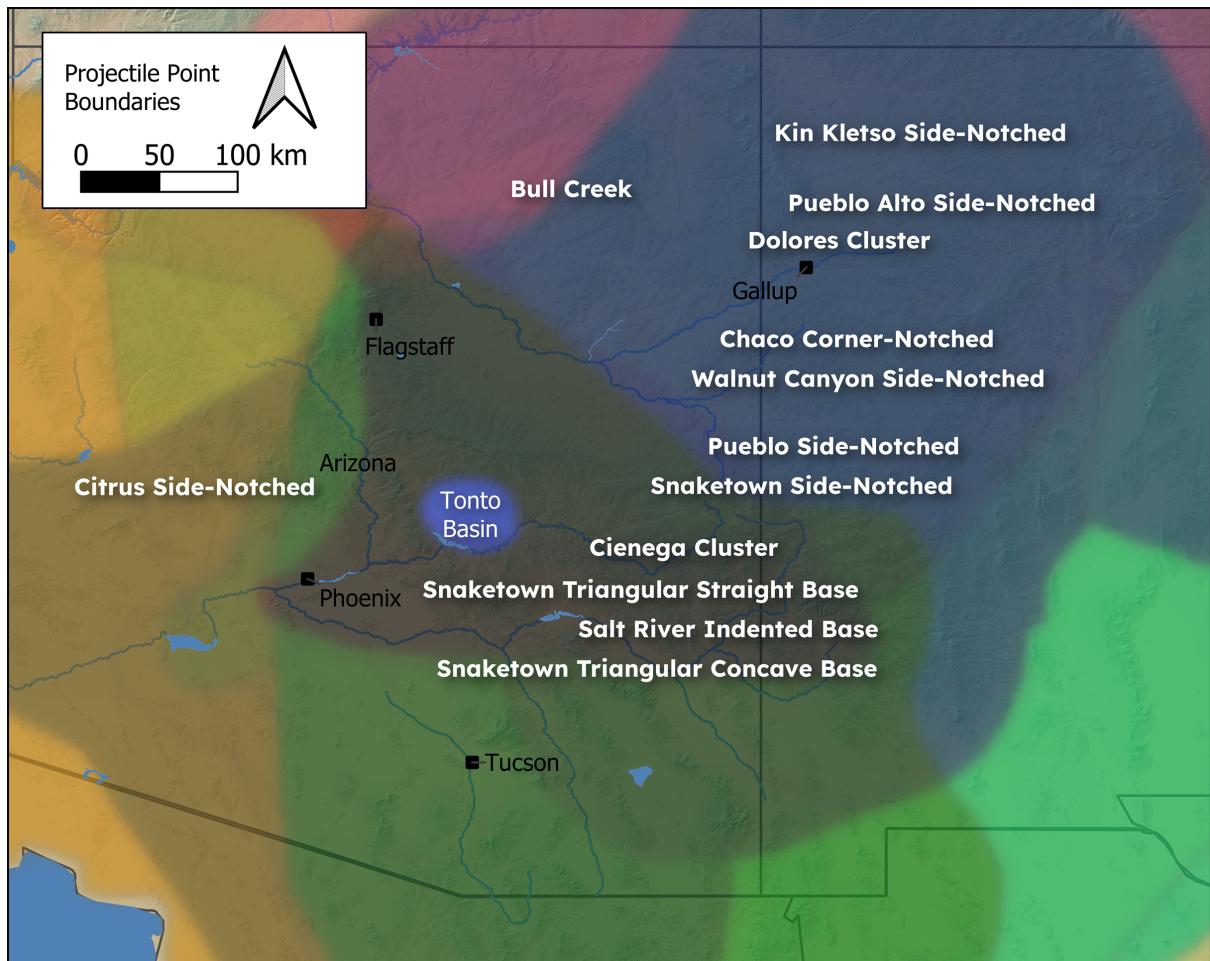


Figure 4. Location of selected projectile point boundaries defined by Justice (2002) as digitized by Buchanan and colleagues (2019). Darker colors represent greater overlap in the number of projectile point boundaries.

Table 1. Cluster Names, Types, and Number of Samples

Cluster	Type	total
Chaco	Chaco Corner Notched	9
Chaco	Pueblo Alto Side Notched	7
Cienega	Tularosa Corner Notched	13
Livermore	Guadalupe	12
Pueblo Side Notched	Pueblo Side Notched Concave Base	9
Pueblo Side Notched	Pueblo Side Notched Straight Base	7
Snaketown	Snaketown Triangular Concave Base	9
Western Triangular	Cottonwood Triangular	8

134 projectile point outlines by type. The included projectile point types include some Archaic points and types
135 not expected to overlap with the Tonto Basin projectile points. Archaic points are often found at later sites,
136 and these types form useful comparisons, as they should not match to non-archaic projectile points. One
137 limitation of this study is that projectile points must be complete, or nearly complete (minor damage to the
138 tip or another part of the point that was judged to not significantly impact the shape of the point was ignored).
139 Thus, not all of the illustrations included in Justice's book could be included in the outline analyses.

140 **Tonto Basin Projectile Points**

141 The sample of Tonto Basin points used in this study come from the Roosevelt Platform Mound Study. They
142 come from 18 different sites that were grouped into five clusters in the original reports (see Rice, 1998, for
143 an overview). Figure 6 shows the projectile points used in this study grouped by site cluster. The majority of
144 these sites were occupied during the Roosevelt phase (AD 1275-1325) and early portion of the Gila phase (AD
145 1325-1450). The sites consist primarily of compounds, room blocks, and platform mounds.

146 The projectile points exhibit a variety of forms (see figure 7). Rice (1994) classified Tonto Basin points into
147 small and large complexes (likely equivalent to dart and arrow points), and further classified small points into
148 the longer Salado series and the shorter Tonto series. These series were further subdivided using a custom
149 classification scheme based on blade, tang, and base shape, as well as notch style. This is a logical way to
150 classify the points, but it does not easily lend itself to regional comparisons, as other points were not classified
151 in the same way. Nearly all of the points in the original sample consisted of side-notched or triangular points.
152 Because the results of the analysis discussed below indicated that analyzing the points by shape was the most
153 logical choice, only the triangular and side-notched points were used in this study.

154 **Geometric Morphometrics**

155 I used two GM methods in this analysis: elliptical Fourier analysis (EFA) and full generalized Procrustes align-
156 ment (GPA). Something to keep in mind is that GM methods analyze the form of the object separated from size,
157 position, and orientation. Real-world measurements such as length and width are not explicitly included in
158 these methods, although relative dimensions, such as length to width ratio are captured in the overall form of
159 the object. Measurements such as length and weight can be included in various analyses but are not included
160 here. The purpose of this study is to determine whether GM methods alone are sufficient to discriminate
161 between types of projectile points and how they can best be used in the context of the U.S. Southwest.

162 EFA was developed by Kuhl and Giardina (1982) as a quantitative means for describing a closed outline.
163 There are a handful of papers that use EFA for lithic studies in archaeology (e.g., Cardillo, 2010; Fox, 2015;
164 Gingerich et al., 2014; Hoggard et al., 2019; Iovita, 2011; Iovita and McPherron, 2011). The mathematics behind
165 the method are complex to describe, which is one reason the method has not been adopted as quickly as it

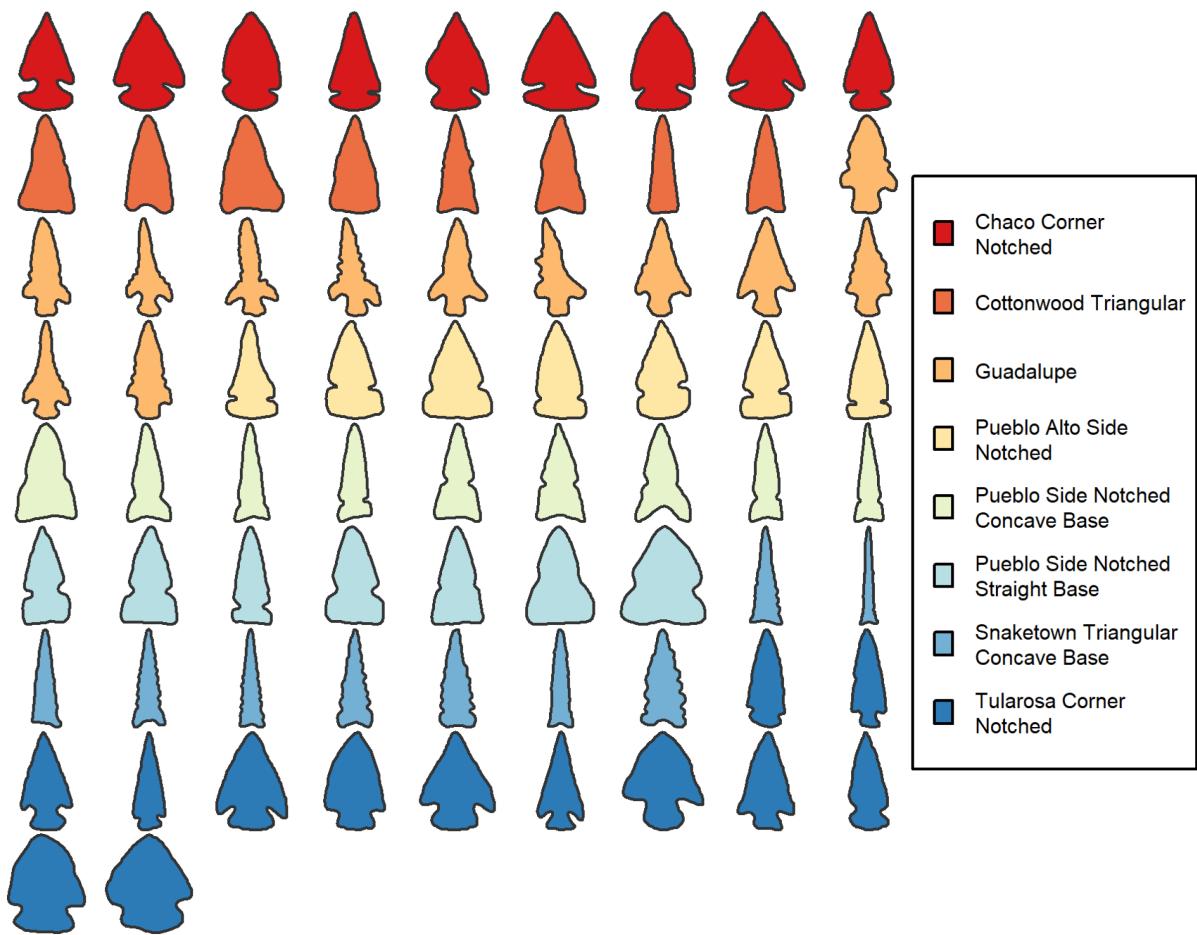


Figure 5. Outlines of projectile point illustrations taken from Justice (2002). Note that the projectile points are not scaled.

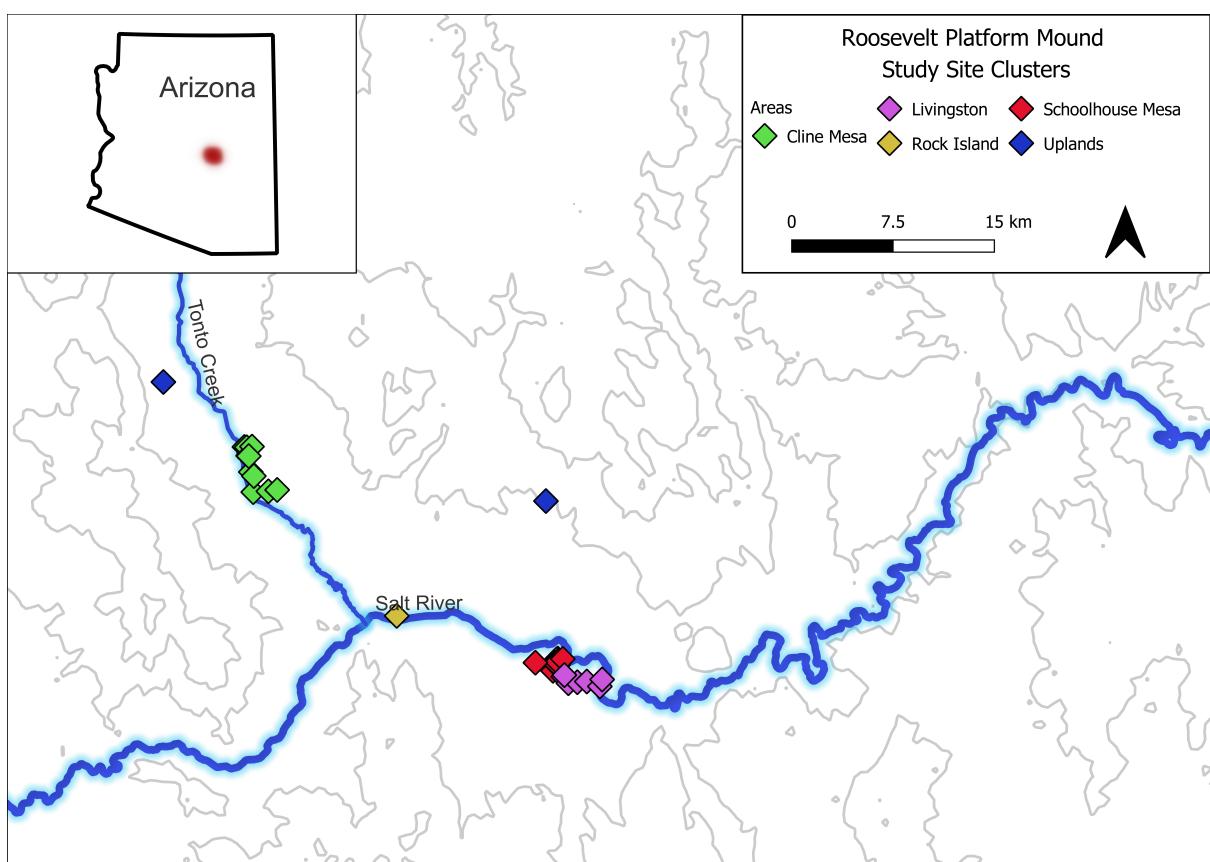


Figure 6. Archaeology sites from Tonto Basin included in discussion. The original reports grouped each site into different clusters. Note that two of the platform mound sites are labeled separately.

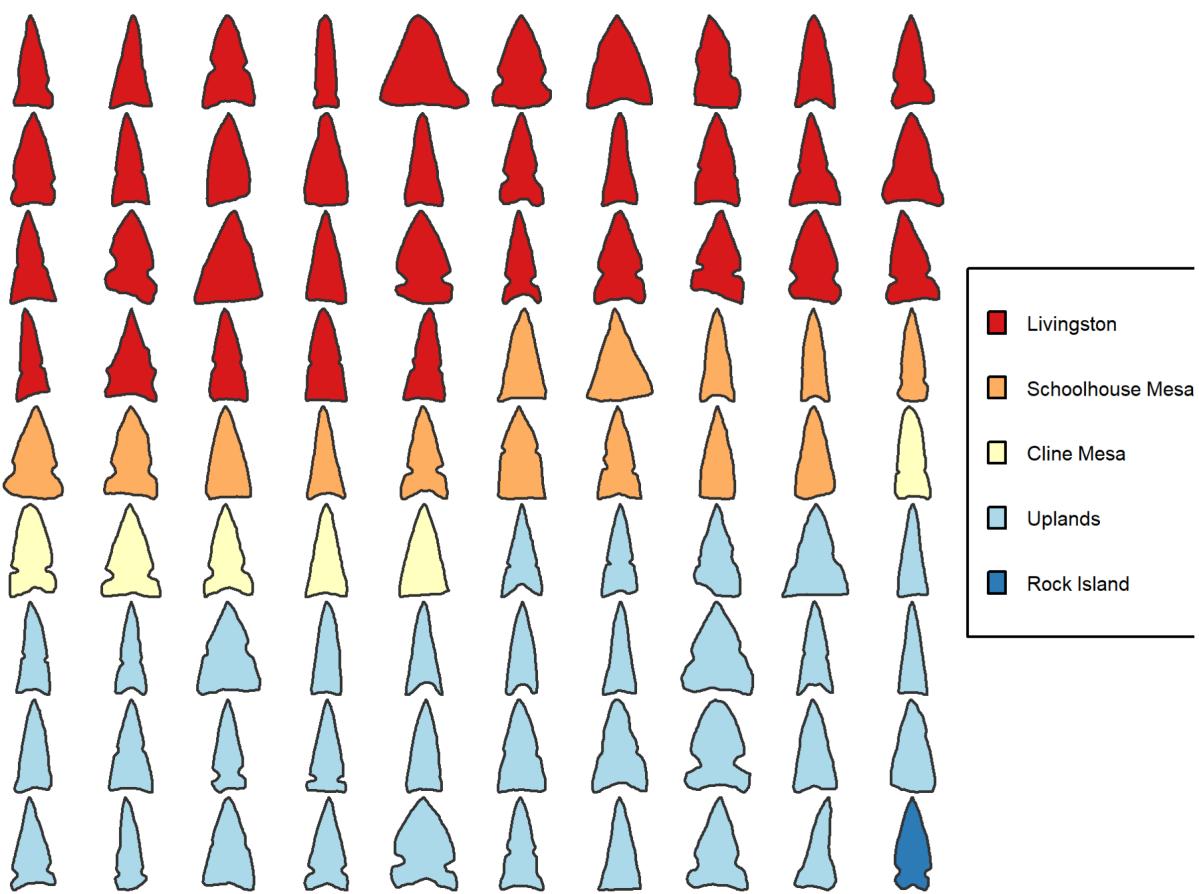


Figure 7. Outlines of projectile points from sites in the Tonto Basin. Note that the projectile points are not scaled.

166 should be (see Caple et al., 2017). Caple and colleagues (Caple et al., 2017) provide an excellent description
167 of EFA for non-mathematicians, and the reader is referred to their treatise for more details. For my purposes,
168 it is enough to know that EFA analysis requires a closed outline and a number of harmonics. The harmonics
169 can be thought of as ellipses in a time series used to describe the shape of the object. Three harmonics
170 can be used to create an oval shape, and 12 harmonics is sufficient for a complex projectile point outline.
171 The number of harmonics necessary to capture the outline can be computed. For example, if you wish to
172 reconstruct an outline with 99.9% accuracy, then the exact number of harmonics necessary can be calculated
173 using the formula:

$$HarmonicPower_n = \frac{A_n^2 + B_n^2 + C_n^2 + D_n^2}{2}$$

174 where n is the harmonic power and A, B, C, and D are the coefficients generated from the EFA. EFA creates
175 a series of coefficients—four for each harmonic (A,B,C,D)—which can be used in multivariate statistics. Most
176 commonly, principal components analysis (PCA) is used to transform the EFA values. The PCA results can
177 then be used in distance-based methods such as clustering or even network analysis.

178 Generalized Procrustes Analysis, or GPA, is primarily a way to align, scale, and rotate points (Gower, 1975).
179 Instead of outlines like EFA, GPA requires landmarks located on homologous locations for each object (Rohlf
180 and Slice, 1990). As an alternative to landmarks, semilandmarks can be placed at equidistant locations around
181 the object. There is substantial discussion on the validity of certain types of landmarks and the use of semiland-
182 marks as landmarks (e.g., De Groot, 2011; MacLeod, 2017; Okumura and Araujo, 2019; Shott and Trail, 2010).
183 Although there are some methods that can introduce error, there is no reason to exclude semi-landmarks
184 outright. One disadvantage of traditional landmark analysis, compared to EFA, is that the analyst must be
185 more involved in the selection of the number and placement of landmarks. Once the landmarks, or semiland-
186 marks are placed on the objects, they are iteratively modified to achieve the best possible alignment between
187 shapes without changing the relative positions between landmarks. This modification is done using the GPA
188 procedure. As with EFA, the next step is usually to perform a PCA analysis. Landmark analysis using GPA is
189 more common, so far, in archaeological analysis of stone tools than EFA (e.g., Archer et al., 2018; Bischoff
190 and Allison, 2020; Buchanan, Eren, et al., 2015; Charlin and González-José, 2018; Fisher, 2018; Gingerich et al.,
191 2014; Herzlinger et al., 2017; Lycett et al., 2010; Riede et al., 2019; Selden et al., 2020; Shott and Trail, 2010;
192 Smith et al., 2015; Thulman, 2012).

193 The project was initially designed to compare the EFA results with a semilandmark analysis using the full out-
194 line of the projectile point by taking a sample of the outline and using the sampled coordinates as landmarks.
195 Both approaches yielded similar results, but neither achieved satisfactory accuracy. The research design was
196 then modified to include a more traditional landmark analysis to determine whether it would improve upon
197 the initial design.

198 There are some disadvantages to using landmarks, which is why the EFA/semlandmark approach was
199 initially favored. The principal disadvantages to using landmarks are reproducibility and accuracy. Landmarks
200 are more subjective in many ways than the semilandmarks or EFA (see Shott and Trail, 2010, p. 205). The
201 analyst must decide how many points to place, what topological points should be used as landmarks, and
202 how many landmarks should be used. The placement of landmarks can vary between analysts and can be
203 affected by the instruments or software used to collect or create the landmarks. Another major concern is
204 the loss of detail from not considering the entire outline. Serrated points and points with more than one
205 notch (this occurs more often than one might expect in Southwestern points) are difficult to capture without
206 including a lot of landmarks, which are only applicable in a minority of situations. Secondary to these points,
207 but still a concern, is that placing landmarks can be a more time-consuming process, as it is not as subject to
208 automation as semilandmarks or EFA.

209 Despite the disadvantages, landmarks are widely used for good reasons. I see two main advantages to
210 landmark analysis in the context of projectile point analysis. The first is that the analyst can use their prior
211 experience to determine what topological locations on the projectile point are most useful for discriminating
212 between types. Decades of research on projectile points has refined many typologies into useful tools, despite

their limitations. This knowledge can be applied to choosing appropriate landmarks. The second advantage is that outline analysis requires complete points, whereas landmark analysis can use damaged points. If chunks of the projectile are missing then the outline is not usable. Possibly, the missing portion could be estimated and filled in, but that process is more error prone than estimating missing landmarks. Landmarks can be placed on reconstructed projectile point illustrations or missing landmarks can be estimated mathematically (Gunz et al., 2009). Most projectile points suffer from some type of damage and some of the projectile points I classified as "complete" suffer from minor damage to the tip of the point or elsewhere. The use of damaged points can greatly increase the available sample size for studies, which is often a major limitation in projectile point studies.

Landmark configurations can vary significantly, depending on what the analysis is designed to measure and on the point type. Most of the area of a projectile point is usually in the blade—the portion above the notches. The base of the point, the portion below the notches, is also the hafting element. For projectile point typologies, the base of the point usually contains the most important elements for determining the type—notching style and basal shape being the two major elements. Thus, if most of the landmarks are on the blade margins then the base of the point is not getting as much coverage. It is more than just tradition that the base gets the most attention. Hafting a point is an important technological choice, more so than how long the point is. Furthermore, projectile points can be resharpened. Resharpening the blade margins can modify the shape of the blade and change its appearance. While it is possible to modify the base of the point and even convert a side-notched point into a corner-notched point and vice-versa, it is unlikely that this happened regularly with the small arrowpoints used in this study (Loendorf, Rogers, et al., 2019).

Because of the vagaries of placing landmarks, I used two configurations in this study. In the first configuration, the full outline was used. Only what I term, the corner of the projectile point was used in the second configuration. Figure 8 shows the first landmark configurations. The landmark configuration was designed to place fewer landmarks along the blade margins and more landmarks along the notches and the base. Separate curves were placed between the tip of the point and the notches and the notches and the base (or the tip and the base for triangular points), and landmarks were placed at equidistant locations along the curves. The second configuration is much sparser (figure 9). The landmarks were placed only on the right side of the point—this was arbitrarily chosen. For the side-notched and corner-notched points, landmarking started from the top of the notch, moved to the middle of the notch and then the bottom portion of the notch. For corner-notched points, this last landmark marked the right corner of the point, but for side-notched points an additional landmark was placed. The final landmark was placed at the center of the basal margin. Triangular points differed by placing the first landmark in the center of the blade margin. The first approach contains between 30 and 42 landmarks that cover the entire point outline, whereas the second ranges from 3 to 5 landmarks that cover only a portion of the projectile point. These extremes were chosen to provide significant contrast between approaches.

Comparisons

The first step in the analysis was to determine how well projectile points typed by Justice could be correctly assigned using GM methods. Linear discriminant analysis (LDA) was used to type the projectile points using the GM results. A general target of 0.85 was arbitrarily chosen as a minimum target for acceptable results—meaning that 85% of the projectile points were classified correctly. As mentioned previously, Justice placed each projectile point type into a cluster. Presumably, projectile point types in the same cluster should be more closely related than they are to projectile point types in other clusters. This gives another level of comparison that was used in addition to the types.

The original intent was to compare EFA versus semilandmarks placed at equidistant locations around the outline. However, these results were unsatisfactory, and a more traditional landmark analysis was also completed. Because the number and placement of landmarks has a significant impact on the outcome of the

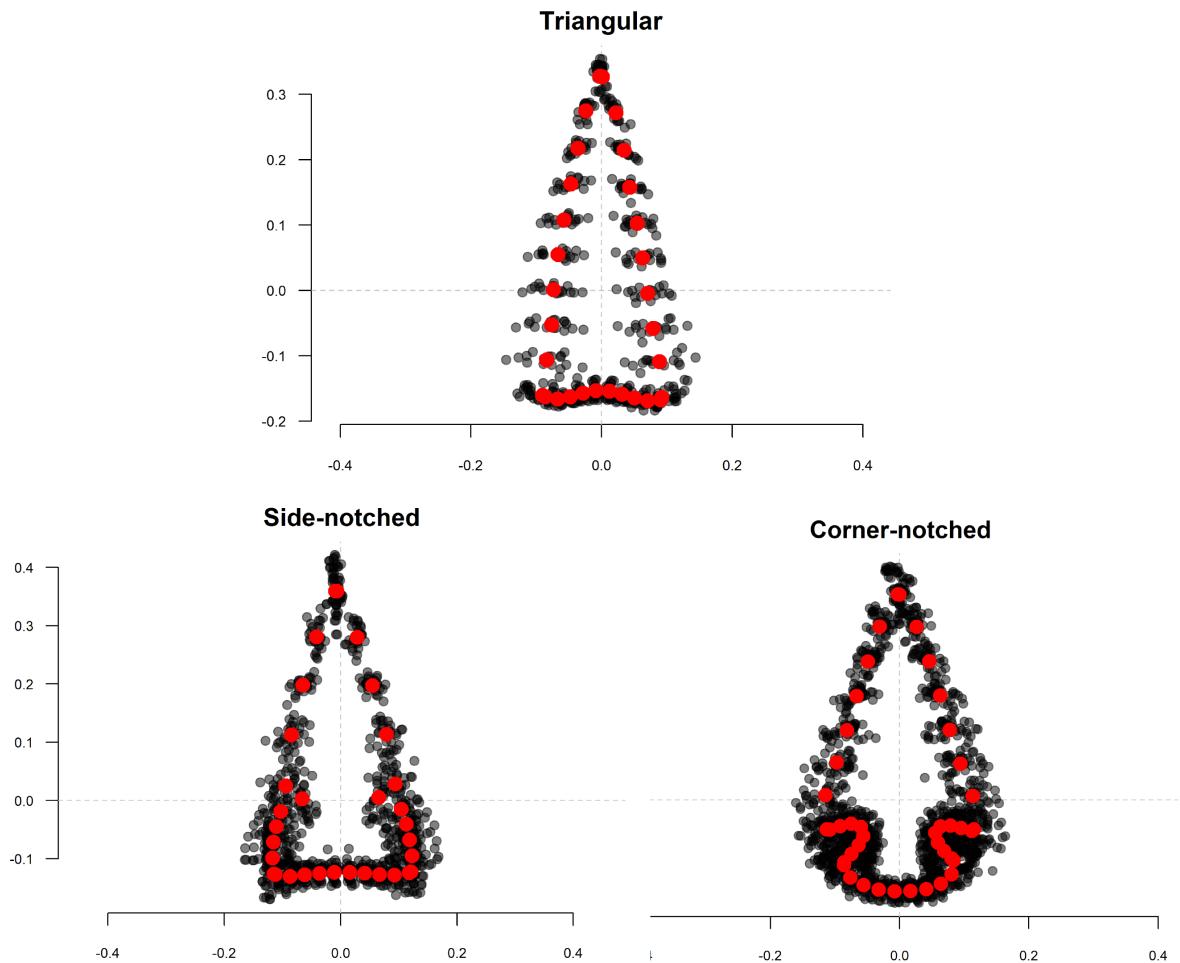


Figure 8. Comparison of the full outline landmarks for corner-notched, side-notched and triangular shaped points from Justice's (2002) projectile point illustrations. Red dots indicate the mean location for each of the landmarks.

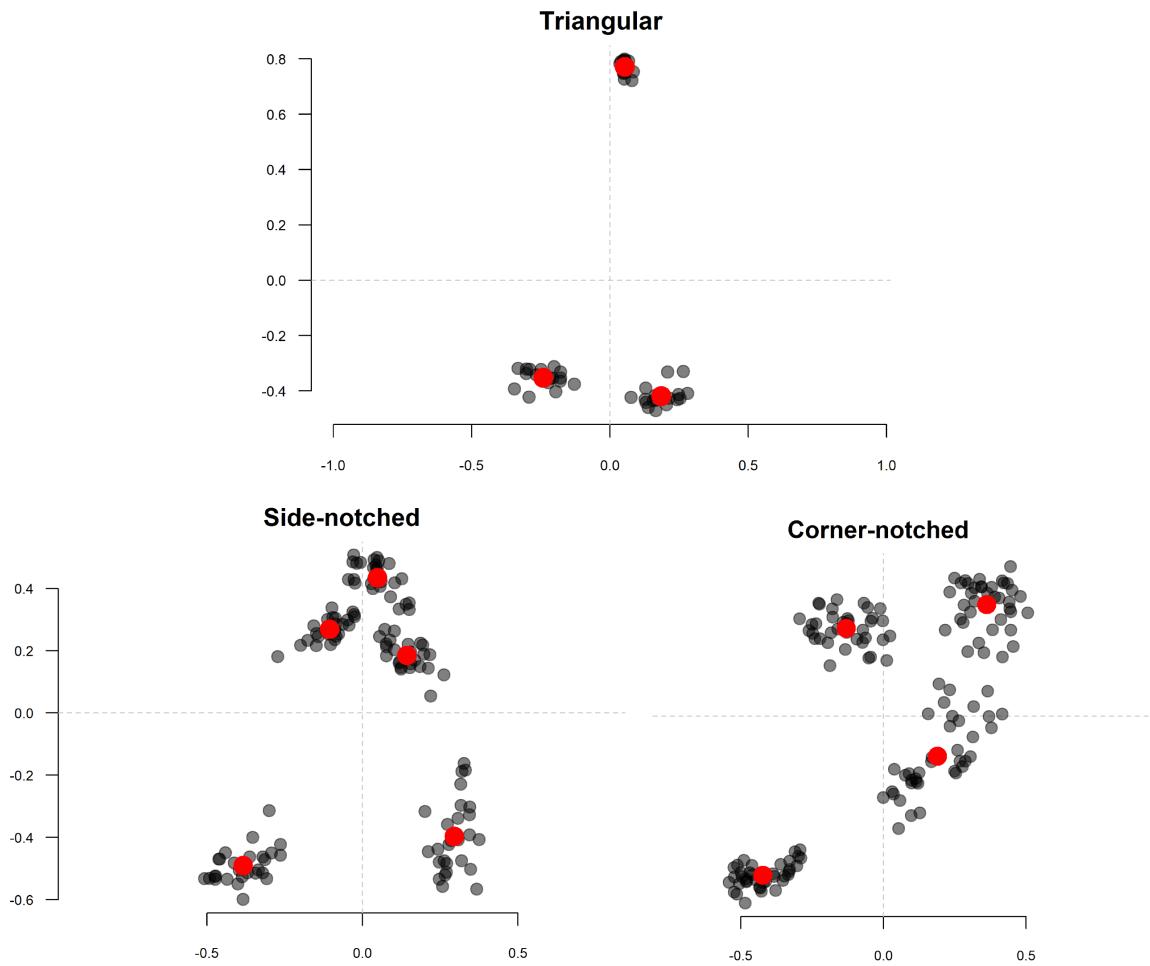


Figure 9. Comparison of the projectile point corner's landmarks for corner-notched, side-notched and triangular shaped points from Justice's (2002) projectile point illustrations. Red dots indicate the mean location for each of the landmarks.

Table 2. Linear Discriminant Analysis Results for Projectile Point Types

Type	EFA	semiLdk	Ldk	Ldk-corner	Mean
Chaco Corner Notched	0.56	0.89	0.85	0.85	0.79
Cottonwood Triangular	0.38	0.38	1.00	0.75	0.63
Guadalupe	0.75	0.83	0.93	0.93	0.86
Pueblo Alto Side Notched	0.86	0.71	1.00	1.00	0.89
Pueblo Side Notched Concave Base	0.44	0.78	0.73	0.73	0.67
Pueblo Side Notched Straight Base	0.57	0.57	0.43	0.71	0.57
Snaketown Triangular Concave Base	0.78	0.89	0.78	0.89	0.84
Tularosa Corner Notched	0.77	0.69	0.60	0.80	0.72
Mean	0.64	0.72	0.79	0.83	0.74

Table 3. Linear Discriminant Analysis Results for Projectile Point Clusters

Cluster	EFA	semiLdk	Ldk	Ldk-corner	Mean
Chaco	0.81	0.88	0.92	0.92	0.88
Cienega	0.69	0.69	0.60	0.80	0.70
Livermore	0.83	0.75	0.93	0.93	0.86
Pueblo Side Notched	0.75	0.56	0.89	1.00	0.80
Snaketown	0.78	0.89	0.78	0.89	0.84
Western Triangular	0.38	0.50	1.00	0.75	0.66
Mean	0.71	0.71	0.85	0.88	0.79

study, two different landmark configurations were used.

Part of the reason the results were unsatisfactory for the EFA and semilandmarks was that the LDA analysis had trouble discriminating between notched and unnotched points and between side-notched and corner-notched points. These are some of the most basic distinctions that are made when analyzing projectile points. While it would be convenient if the analysis did not require an additional step, it is not difficult to separate the points into these basic shapes prior to the GM analysis. I separated the Justice points into three classes: side-notched, corner-notched, and triangular. For this study, I combined stemmed points into the corner-notched category.

Tables 2, 3, and 4 show the LDA results by type, cluster, and by shape. These tables will be referred to in the sections that follow.

Table 4. Linear Discriminant Analysis Results for Projectile Point Shapes

Shape	EFA	semiLdk	Mean
Corner-notched	0.88	0.91	0.90
Side-notched	0.65	0.74	0.70
Triangular	0.76	0.59	0.68
Mean	0.76	0.75	0.76

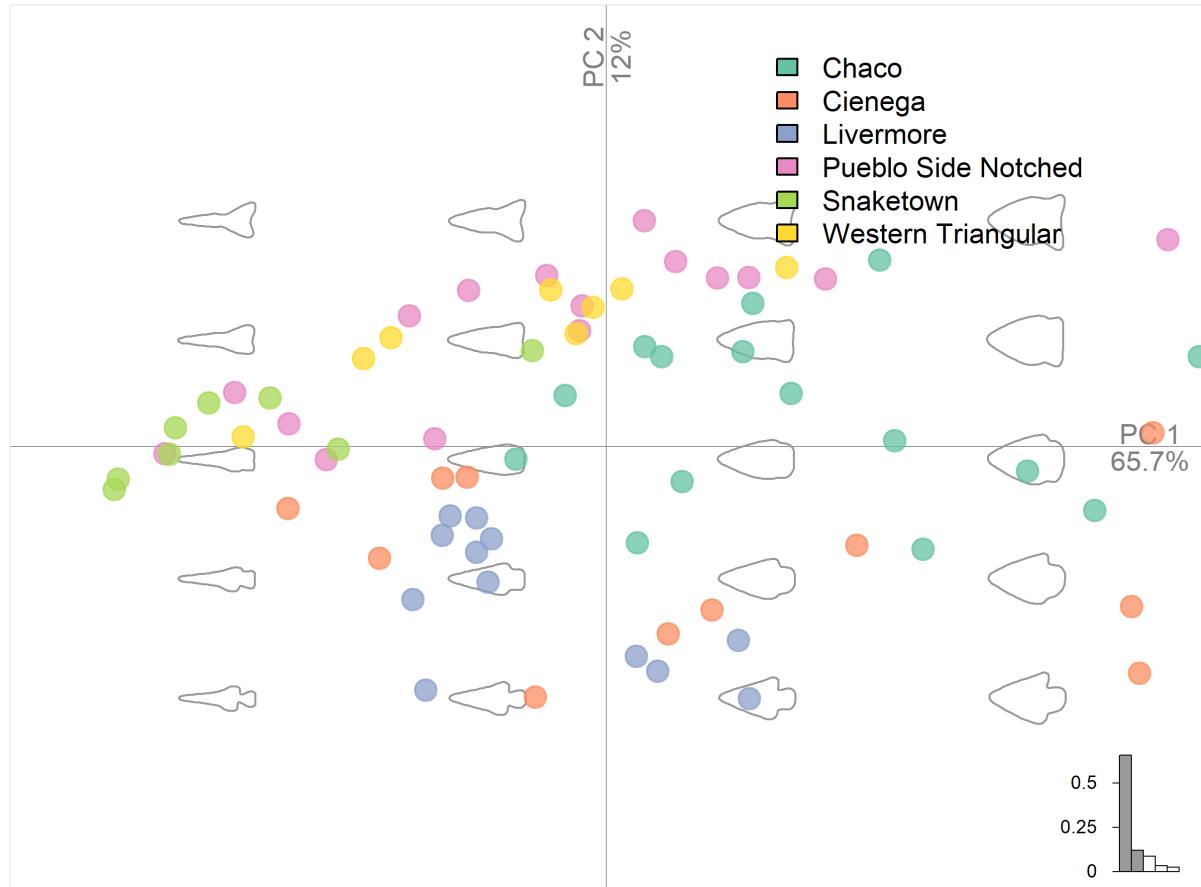


Figure 10. Principal components plot showing projectile points from Justice (2002) and the morphospace based on an elliptical Fourier analysis. The points are labeled by the cluster assigned by Justice.

269 Elliptical Fourier Analysis

270 The EFA analysis was conducted using the Momocs packages (Bonhomme et al., 2014) in R (R Core Team, 2022).
 271 The first step was to calculate the number of harmonics to use. In this case 12 harmonics described 99% of the
 272 variation in the projectile point outlines. Next, the EFA function was used on each projectile point, and then a
 273 PCA was used to reduce the dimensionality. Figure 10 shows the results of the PCA analysis. A useful feature
 274 of PCA plots using these data is that the morphospace can be plotted with the PCA results. The morphospace
 275 shows how the shapes vary along each axis of the PCA. In this case, PC1 (the first principal component) varies
 276 between short, wide points and long, narrow points. Some of shapes on the top and bottom left have inverted
 277 into impossible shapes, but note that no points fall into these areas. PC2 varies primarily from stemmed points
 278 to side-notched points.

279 The first objective for the analysis of the Justice points is to determine how well the different point types can
 280 be discriminated. Meaning, how well can GM methods classify these points into their original categories. The
 281 LDA results were far from the target goal of 0.85 for most projectile point types, and only one type (Pueblo
 282 Alto Side Notched) met the target (see EFA results in Table 2). The results were better when the projectile
 283 point types were grouped into clusters, as shown in Table 3; however, none of the clusters met the target
 284 of 0.85. Even more disconcerting were the results shown in Table 4, as only the corner-notched points were
 285 discriminated with an accuracy greater than the target.

286 The differences in classification accuracy between corner-notched, side-notched, and triangular points can
 287 perhaps best be explained by examining the mean shapes of each point. Figure 11 shows the mean shapes
 288 for the selected projectile point types. These are the mathematically average shapes when all of the projectile

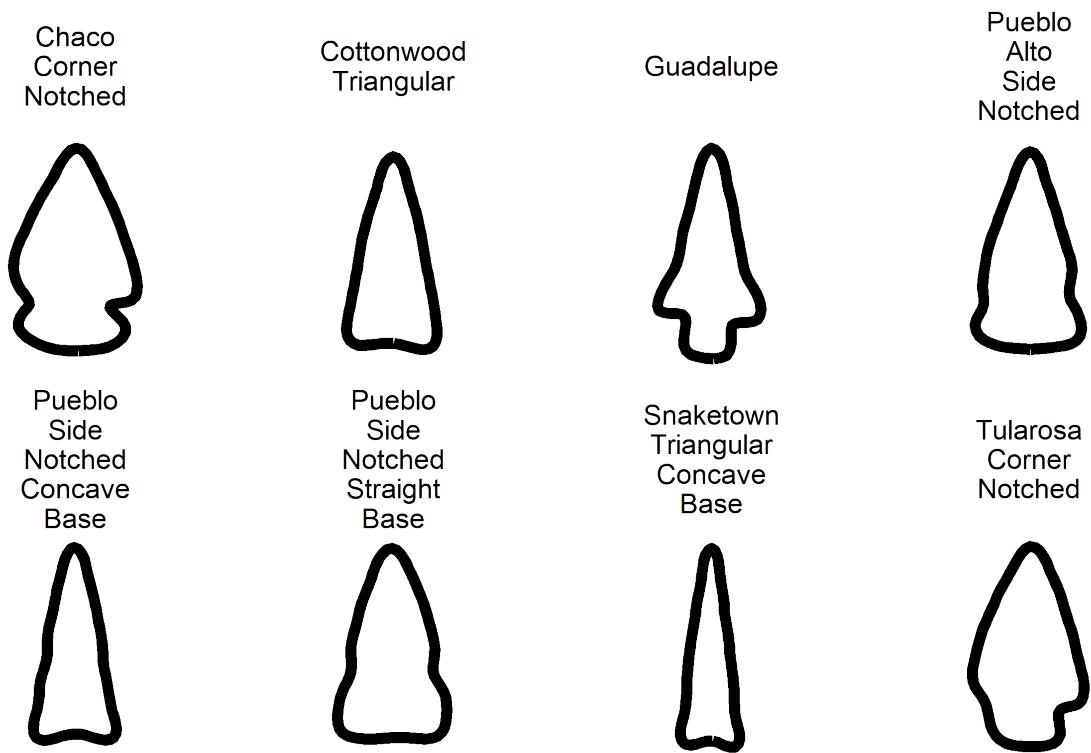


Figure 11. Mean shapes by projectile point type using elliptical Fourier analysis.

289 points in the type are combined. This has a tendency to average out the notches for the side-notched points,
 290 as the placement of these notches vary in height. Pueblo Alto Side Notched points appear to be an exception
 291 to the side-notched problem, as they have the highest classification accuracy. Corner notched and stemmed
 292 points must, by definition, always have their notches or stems in the same location, even though the shape
 293 of the notches and stems still varies. This explains why it is easier to discriminate them from other point
 294 types. As for the side-notched and triangular points, sometimes the notches are subtle and the notches are
 295 only a small part of the whole form, which is clearly not a strong enough element to separate triangular and
 296 side-notched points consistently.

297 **Semilandmarks**

298 Analyzing projectile points using semilandmarks is comparable to the EFA analysis. The major analytical choice
 299 is how many landmarks to use. Each projectile point is represented by a varied number of coordinates that
 300 represent its outline. The number of points must be standardized so that each projectile point has an iden-
 301 tical number of coordinates, which is done using the Momocs package. The choice of how many points to
 302 use does affect the GM analysis. To solve this problem, I tried different numbers of semilandmarks varying
 303 from 10 to 100. More than 100 points appears to no longer have a substantial effect on the results. Each
 304 projectile point was sampled multiple times and then all of the points were classified using LDA according to
 305 the same procedures used for the EFA analysis. The results of this test ranged from 65% accuracy (10 points)
 306 to 73% accuracy (30 points—the number used for the final analysis) with somewhat higher numbers of points
 307 consistently measuring in at 64% accuracy.

308 The morphospace (see figure 12) based on the PCA analysis has similar dimensions of variation as the EFA

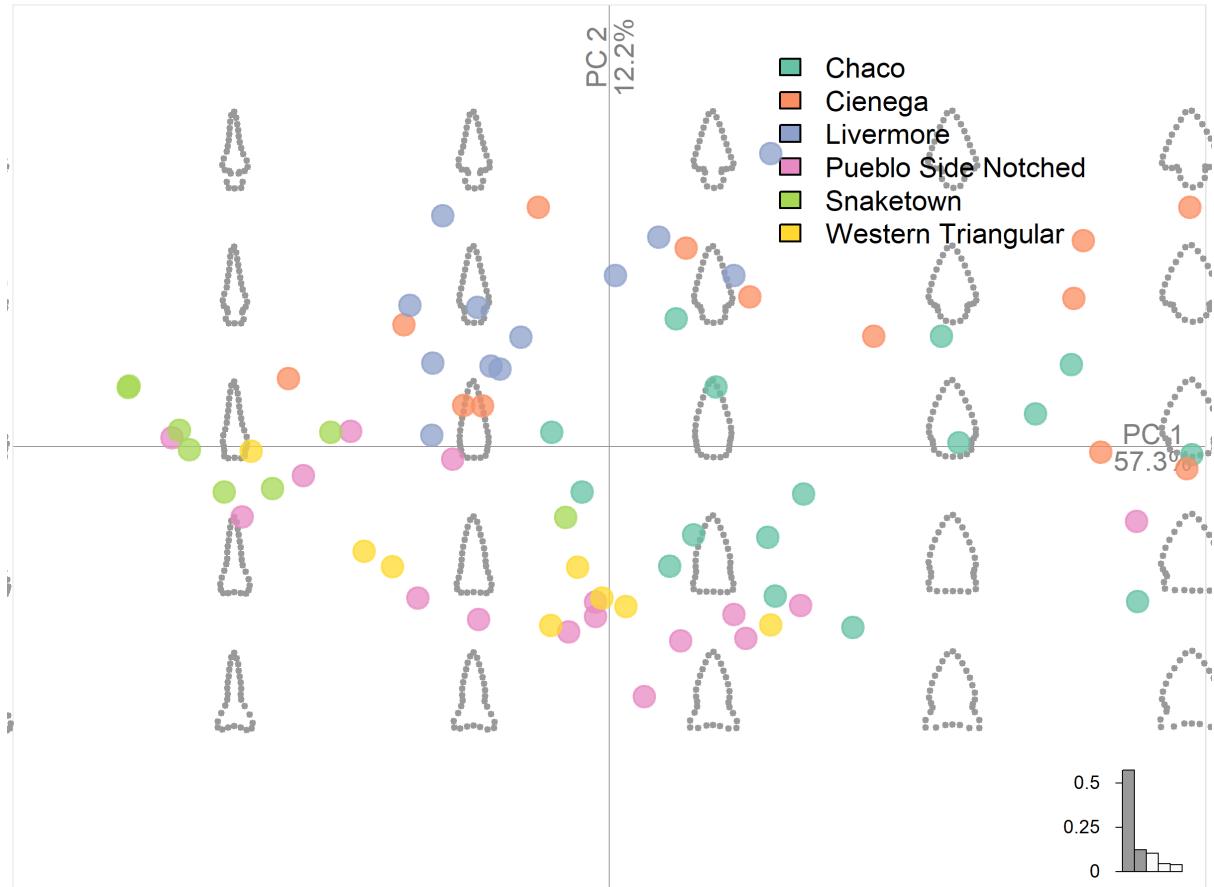


Figure 12. Principal Components Plots showing projectile points from Justice (2002) and the morphospace based on a semilandmark generalized procrustes alignment. The points are labeled by the cluster assigned by Justice.

analysis. The major improvement in accuracy (the 'semiLdk column of Tables 2-4) is perhaps due to the better alignment generated by the GPA procedure, but that is only speculation.

Regardless of the reason, a jump from 64% classification to 72% is a substantial improvement. It does not reach the target of greater than 85% classification accuracy, but it is a step in the right direction. The mean shapes are nearly identical to the EFA analysis, which indicates that this method suffers from the same problems with side-notched and triangular points, but it does a somewhat better job differentiating corner-notched and side-notched points. Curiously, it does a worse job differentiating triangular points. It seems the side-notches are still a problem.

317 Landmarks

Neither the semilandmarks nor EFA adequately distinguished between point types or between point shapes. Identifying notches was particularly troublesome. A solution to this problem was to use landmarks and explicitly identify the notches or lack of notches. No comparison was made between projectile point shapes using landmark analysis, as initial experiments determined that it was best to use different landmark procedures for the different shapes. Perhaps machine learning may solve this problem (see Castillo Flores et al., 2019; MacLeod, 2018; Nash and Prewitt, 2016). Triangular points require a different approach than side-notched points, and even side-notched and corner-notched/stemmed points require different procedures.

The LDA results for the first landmark configuration (Ldk in Tables 2-4) are much better than EFA and better than the semilandmark analysis, but still not as accurate as desired. The biggest underperformer by far was

327 Pueblo Side Notched Straight Base at 0.43. All of the previous analyses struggled to capture basal shape
328 distinctions, but this analysis struggled more so. What is particularly notable is that Cottonwood Triangular
329 points were classified perfectly whereas they were previously the worst performing type. As Table 3 shows,
330 the cluster assignments performed well. If the problems sorting Cienega from Snaketown can be sorted out,
331 then the results would be excellent.

332 The final analysis used the second landmark configuration—the projectile point corners. These results
333 proved superior to the first landmark configuration and are almost 20% points higher in accuracy than the
334 EFA results on average. The lowest type for accuracy was again Pueblo Side Notched Straight Base, but it
335 improved from the first landmark configuration to 0.71 from 0.43. The accuracy results were more consistent
336 and accurate. With some enhancement, this configuration could likely achieve better results.

337 Not only did landmark analysis provide superior accuracy, but it will also make it easier to use larger sample
338 sizes. Presumably, notching style is an important attribute that should be captured in the analysis. If EFA or
339 the semilandmark analysis as conducted here fails to sufficiently emphasize the notches, then these methods
340 are insufficient for my purposes. While landmark analysis is more time-consuming, the use of the second
341 configuration does reduce the burden of landmarking.

342 Tonto Basin Points

343 The initial intent was to classify the Tonto Basin points using the analysis of the Justice points; however, the
344 limited sample size limits the validity of the exercise. The analysis was not futile though, as the second land-
345 mark configuration using the corners of the projectile points proved the most effective. I therefore used the
346 same landmark configuration to analyze the points from Tonto Basin. The results of the GPA and PCA analysis
347 were used in a hierarchical cluster analysis using Ward's method (see Murtagh and Legendre, 2014). Figure
348 13 is a network graph showing the results. Several sites in Tonto Basin only had one or two types of projectile
349 points (low sample sizes were again problematic), but some of the larger, well-excavated sites shared all or
350 most of the projectile point types. It is beyond the purpose of this study to explore the patterns in this data,
351 but the methods clearly provide useful data for exploratory analysis.

352 The final question I wished to address in this study was whether it was necessary to use a typology in a GM
353 analysis of projectile points. There are many ways to answer this question, but in short, the answer is no. That
354 does not mean typologies are not useful, but they can mask important variation. The following is one way to
355 approach analyzing projectile points without using a typology.

356 Because the results of the GM analyses can be projected into multidimensional space, the distance between
357 these values is meaningful and can be directly compared. One way to flatten multidimensional space is to
358 calculate the Euclidean distance between each point and display the results as a network graph, as in figures
359 14 and 15. This way each point can be compared directly without grouping the points into types. The results
360 are messier, but subtle variation in morphology is easier to visualize this way. The closer the points are to
361 each other, the more similar they are in shape, keeping in mind that only the corners of the projectile points
362 (from the notches down for the side-notched points) were used in this analysis. Many of the points clustered
363 closely together, indicating a common shape across the sites. The side-notched points have a particularly
364 large cluster of typical Hohokam side-notched points. Yet there are also a large number of points that do not
365 closely match the points, which indicates that there is also a lot of variation. This variation may represent
366 idiosyncrasies, exchange, migration, novice knappers, or a different intention for the point. Regardless of the
367 purpose, the GM analysis better captures the “otherness” of a point than classifying a point as other/unknown
368 or worse, forcing it into a category it does not belong in.

369 The clustering analysis provided several different projectile point types and provided an overview of what
370 sites shared similar projectile points. This type of analysis can be combined with architectural or other data
371 to look for correlations or patterns that provide insights into the behavior of the people who made and used
372 these points. Yet the analyst must take care to ensure the clustering groups are appropriately sized and the

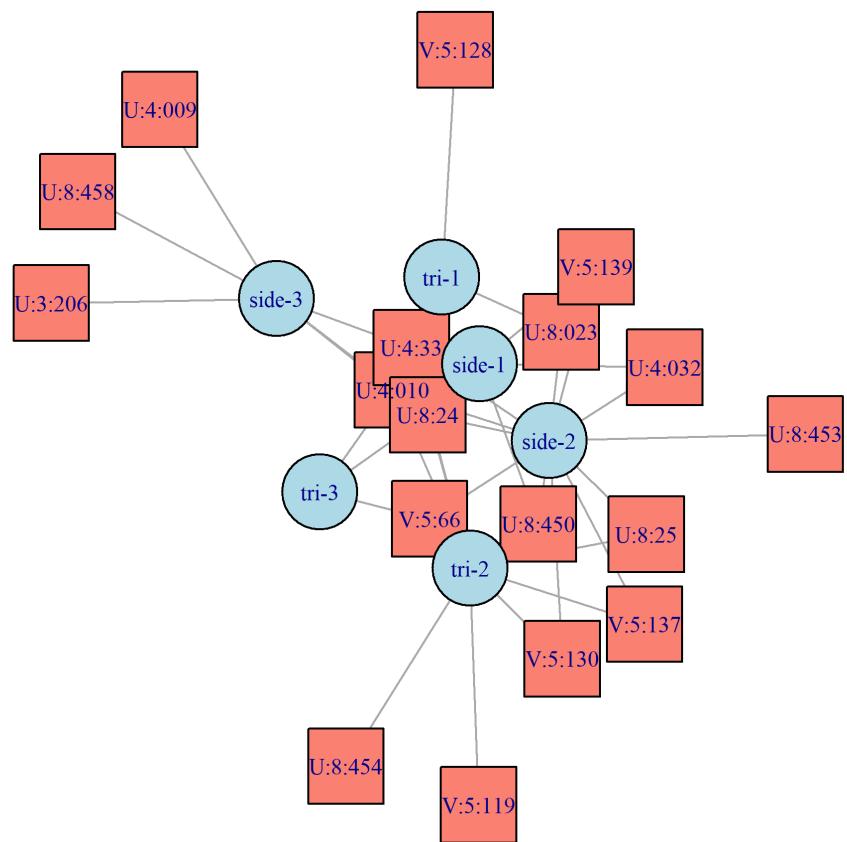


Figure 13. Bipartite network graph displaying assigned projectile point clusters for side-notched and triangular points in Tonto Basin and Tonto Basin sites.

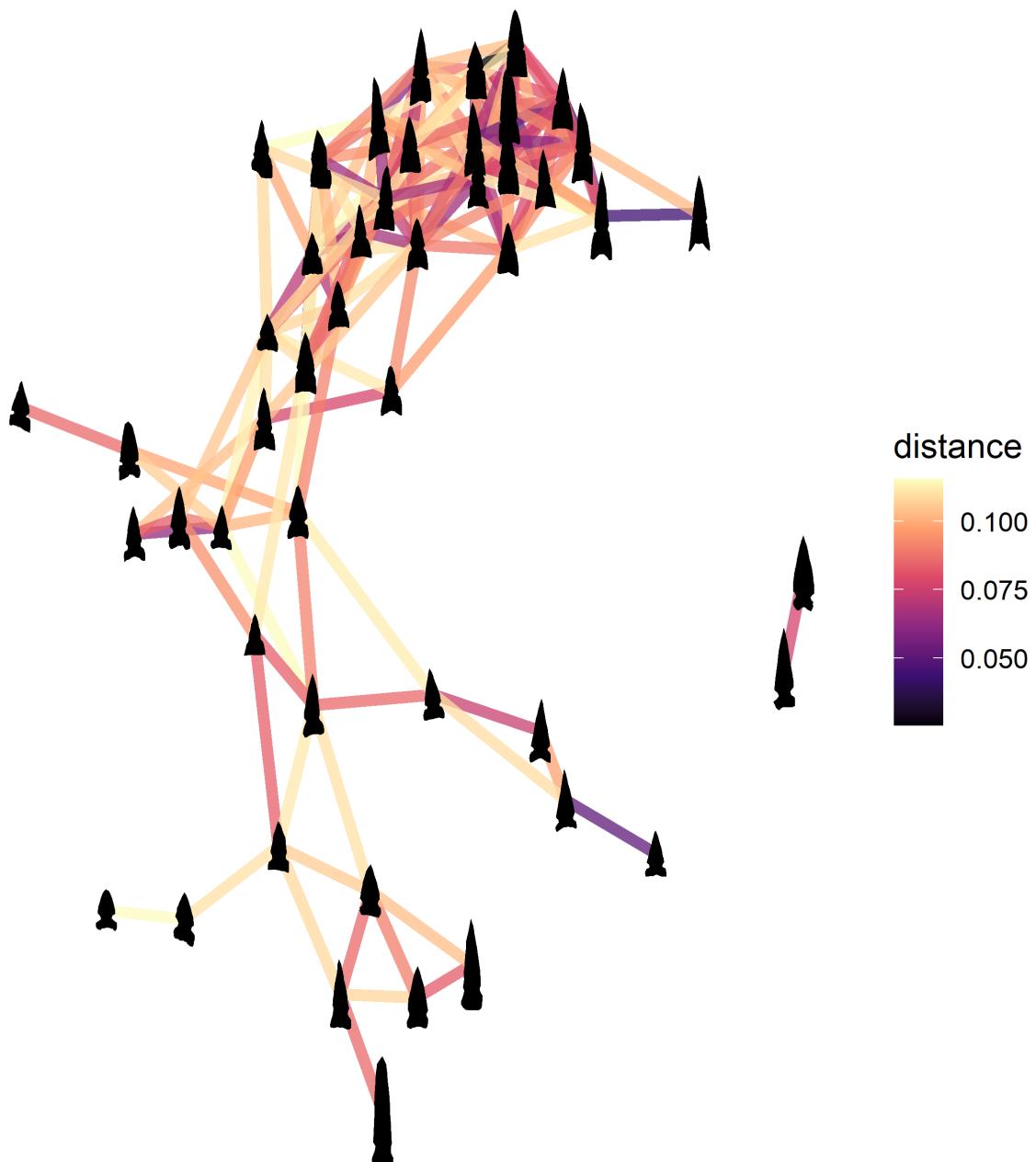


Figure 14. Network graph displaying side-notched points from Tonto Basin as nodes with ties showing the morphometric distance between points. Darker colors represent stronger ties. Note that only the strongest 10% of ties are shown.

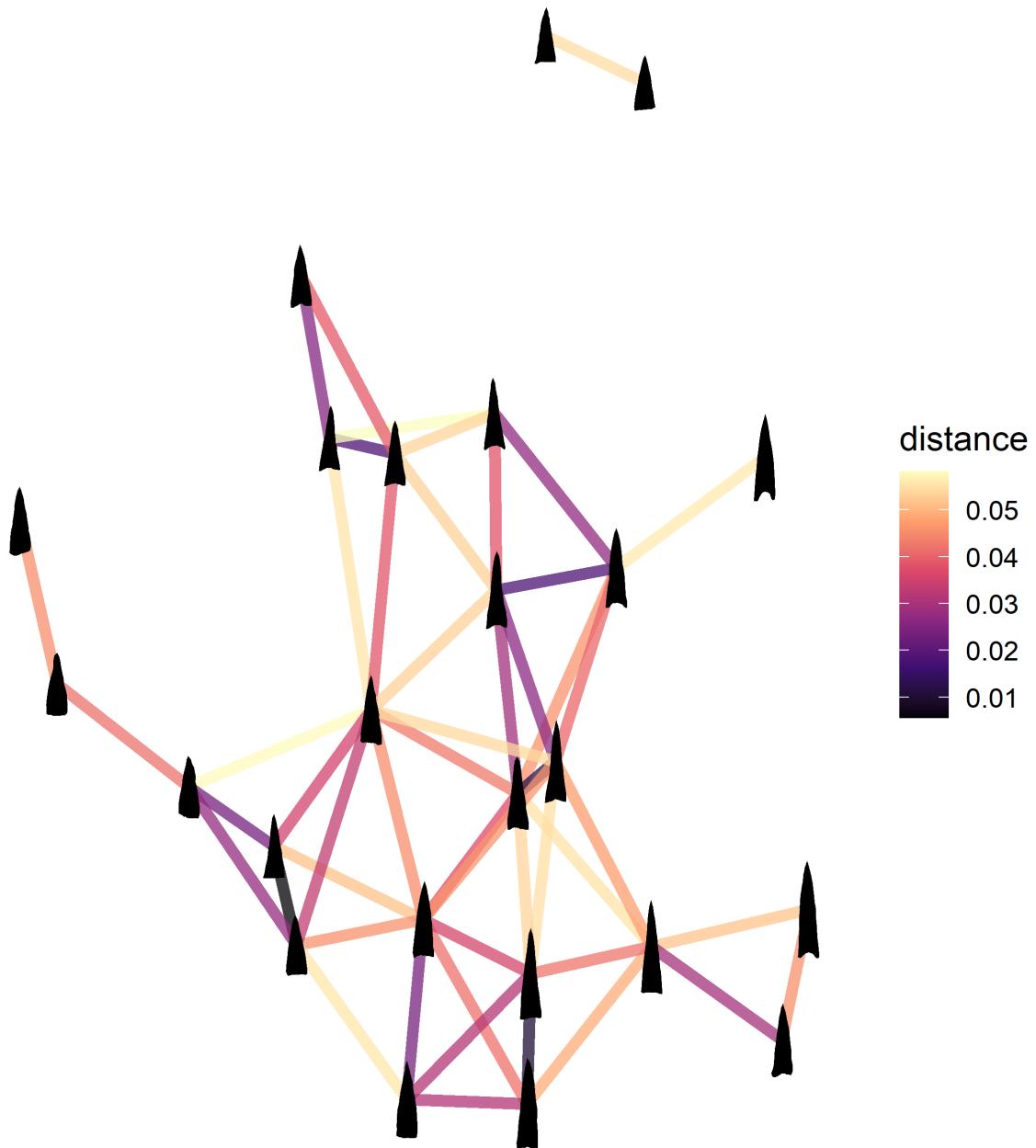


Figure 15. Network graph displaying triangular points from Tonto Basin as nodes with ties showing the morphometric distance between points. Darker colors represent stronger ties. Note that only the strongest 10% of ties are shown.

373 results make sense. One way to view the data more closely is to look at a distance network graph to view the
374 variation in morphometric shape. This way typological distinctions will not mask the variation.

375 Conclusion

376 These analyses provided significant variation in their results, yet they also demonstrated positive results.
377 While EFA underperformed all of the other analyses, a different dataset may favor this analysis. Some of the
378 types performed better for EFA than other types, which suggests that EFA may be the optimal choice for some
379 datasets. A clear result from this exploratory analysis is that GM analysis is not a one-size-fits-all approach. In
380 this case, using the corner of the projectile point—from the base to the middle of the basal margin or from the
381 middle of the blade if the point is triangular—proved to be the most useful method. The main advantage of this
382 method was that it provided the most accurate reproduction of Justice's original classification of the projectile
383 points. Another advantage is that broken points are easier to use with this method. If one half of the point
384 is missing, either the top or the lateral margin, it does not affect the analysis. This increases the number of
385 points available for analysis tremendously compared to only using whole points. The final advantage, though
386 minor, is that the landmarking analysis is simple and only requires three to five landmarks.

387 I mentioned in the introduction how difficult it is to conduct regional analyses with projectile points in the
388 U.S. Southwest. The main difficulty is harmonizing existing typologies and then fitting new projectile points
389 into this typology. While the sample size available for this study was too small to attempt classifying the Tonto
390 Basin points according to Justice's typology, it would be possible to do so with these methods. However, this
391 chapter also demonstrated that it is possible to type points using common clustering methods which may bet-
392 ter capture the variation in projectile point morphology than previously used types. Furthermore, it is possible
393 to analyze projectile points without resorting to types. The distances between projectile point morphologies
394 can be computed and compared directly. These distances could even be aggregated and summarized region-
395 ally. The main challenge for the regional analysis is obtaining the projectile point outlines or landmarks. Once
396 these are obtained, thousands of points can be analyzed and assigned to clusters relatively quickly.

397 Compared to a traditional analysis of linear metrics and weights, a GM analysis can capture much more in-
398 formation and provide more informative ways to analyze and visualize the data. The visualization capabilities
399 of GM is one of its greatest strengths, as it allows the analyst to see the data they are working with, visually
400 validate their results, and share their findings in visually compelling ways. Additionally, this analysis is more
401 reproducible and adaptable than traditional lithic analyses. While the analyst still has a lot of control over a
402 GM analysis, the results should be less biased than analyses based on visual type comparisons.

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407 Data, scripts, and supplementary information availability

408 All relevant data and scripts are available at the following DOI [10.17605/OSF.IO/ZGE9Q](https://doi.org/10.17605/OSF.IO/ZGE9Q) and on GitHub at
409 github.com/bischrob/TontoBasinPoints. The code used in the analysis is included in the manuscript.Rmd file
410 used to create this manuscript, although some lines have been commented out to improve efficiency.

411 Conflicts of interest disclosure

412 The author declares that they comply with the PCI rule of having no financial conflicts of interest in relation to
413 the content of the article.

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