

1 Programming by example: efficient, but not 2 “helpful”

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20 — Abstract —

21 Programming by example (PBE) is a powerful programming paradigm based on example driven
22 synthesis. Users can provide examples, and a tool automatically constructs a program that
23 satisfies the examples. To investigate the impact of PBE on real-world users, we built a study
24 around StriSynth, a tool for shell scripting by example, and recruited 27 working IT professionals
25 to participate. In our study we asked the users to complete three tasks with StriSynth, and the
26 same three tasks with PowerShell, a traditional scripting language. We found that, although our
27 participants completed the tasks more quickly with StriSynth, they reported that they believed
28 PowerShell to be a more helpful tool.

29 **2012 ACM Subject Classification** Human-centered computing → Empirical studies in interac-
30 tion design

31 **Keywords and phrases** User Study, Scripting, Programming by Example

32 **Digital Object Identifier** 10.4230/OASICS.PLATEAU.2018.3

33 **Funding** This research sponsored by NSF grants CCF-1302327 and CCF-1715387.

34 **1** Introduction

35 Scripting languages, such as PowerShell and bash, help IT professionals to more efficiently
36 complete tedious and repetitive tasks. Those tasks can include file manipulations and
37 organizing data, where a simple error can destroy users’ data. As an example, consider the
38 disastrous attempt to remove all backup emacs files with the command `rm * ~`. Additionally,
39 small errors in scripts can lead to malicious behavior, such as data loss [25]. Scripts can
40 be difficult for users to write by hand, requiring users to have extensive experience with
41 regular expressions, programming, and domain expertise in the scripting language of their



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9th Workshop on Evaluation and Usability of Programming Languages and Tools (PLATEAU 2018).

Editors: Titus Barik, Joshua Sunshine, and Sarah Chasins; Article No. 3; pp. 3:1–3:11

OpenAccess Series in Informatics



OASICS Schloss Dagstuhl – Leibniz-Zentrum für Informatik, Dagstuhl Publishing, Germany

42 choice. Depending on the application, a user may need to be able to write a very complicated
 43 regular expression for a relatively simple task. Furthermore, users may not have access to
 44 their scripting language of choice, depending on the operating system and software policies
 45 used by their employer.

46 For these reasons, many end-users search for help on online forums when they need to
 47 write a script [3, 2, 4]. When users seek help in writing a script on forums, they will often
 48 provide a few illustrative examples that convey the goal of the script. This observation was the
 49 basis of StriSynth [12], a research tool that was proposed to make scripting easier and more
 50 efficient by allowing users to program scripts by example. While scripting is a challenging
 51 task, especially for novice programmers, providing examples of the intended behavior is a
 52 more natural interface for scripting. StriSynth supports various types of functions, such as
 53 transformations, filters, partitions, and merging strings.

54 In this work, we explore how scripting by example, specifically with StriSynth, is received
 55 by the real-world target end-users. We designed a user study around StriSynth and recruited
 56 27 IT professionals to participate in the study. In our study we asked users to complete
 57 three tasks with StriSynth, and the same three tasks with PowerShell, a traditional scripting
 58 language. When using StriSynth, users were statistically significantly faster at completing
 59 tasks as compared with PowerShell. However, in a post-study survey when users were asked
 60 which tool they perceived to be more “helpful”, users statistically significantly reported that
 61 PowerShell, with the traditional scripting paradigm, was more helpful. This was counter-
 62 intuitive result, as we expected that a faster tool should be considered to be more helpful by
 63 users. While the formal methods community has largely taken efficiency of task completion
 64 to be an indicator of a good language design, we explore our results here that show this is in
 65 fact a more complex issue.

66 **2 Background**

67 Programming by example (PBE) [6, 23, 27, 7] is a form of program synthesis. It works by
 68 automatically generating programs that coincide with the given examples. In this way, the
 69 examples can be seen as an incomplete, but easily readable and understandable specification.
 70 However, even if the synthesized program satisfies all the provided examples, it might not
 71 correspond to user’s intentions due to this incompleteness in the specification. In this case, a
 72 user must provide further examples to the synthesis tool.

73 To address this issue, StriSynth was implemented as a live programming environment [5]
 74 for PBE. In this way, a synthesized script can be refined with every new provided example,
 75 and thus yields a more interactive experience for the user. Interactive PBE allows end-users
 76 to provide a single example at a time, rather than guessing at the full example set that is
 77 necessary for synthesis.

78 In order to compare the PBE paradigm to more traditional scripting languages, we
 79 have chosen to use the tool StriSynth [12]. StriSynth is an existing tool for automating file
 80 manipulation tasks, in a similar style to Flash Fill’s [1] synthesis of spreadsheet manipulations.
 81 While the use of scripting language such as sed, awk, Bash or PowerShell requires a certain
 82 level of expertise, many tasks can be easily described using natural language or through
 83 examples.

84 **2.1 StriSynth example**

85 To give some context for how StriSynth compares to traditional scripting language paradigms,
 86 we give an example task that can be easily completed with StriSynth. This task comes from

87 a StackOverflow post, where the users discuss challenging regular expressions [3]. The user
 88 asked for a script that will create a link from every item in a directory. To better illustrate
 the goal of the script, the user provided two examples transformations:

```
Document1.docx      ↪      <A HREF='Document1.docx'>Document1</A>
Document2.docx      ↪      <A HREF='Document2.docx'>Document2</A>
```

89
 90 To accomplish this transformation, other users on the forum suggested a solution based
 91 on regular expressions in `sed`:

```
92 sed/\(^([a-zA-Z0-9]+\)\)\.([a-z]+\)/\<a href=\'\1\.\2\' \>\1</a\>/g
```

93 While it was very easy for the user to express the goal of the script by providing examples,
 94 the resulting script is arguably less readable, even for such a simple problem. In contrast, to
 95 solve this problem in StriSynth, a user provides an example showing what a script should do:

```
96 > NEW
97 > "f.docx" ==> "<a href='f.docx'>a</a>"
98 > val F = TRANSFORM
```

99 The keyword `NEW` denotes the start for learning of a new script, after which the user
 100 provides an example of the scripts desired behavior. Based on the provided example, StriSynth
 101 learns a string transformer, and the user saves it with the next command. Every learned
 102 function can be saved using the command `val name = ...` which creates a reference, `name`,
 103 to the learned script. The user may then check how `F` works on different examples to confirm
 104 the learned function is correct.

```
105 > F("Document1.docx")
106 <A HREF='Document1.docx'>Document1</A>
107 > F("Document2.docx")
108 <A HREF='Document2.docx'>Document2</A>
```

109 We observe that the learned transformer F is a function that exactly does what the user
 110 asked initially. However, it only takes a single string as input, while the user wanted a script
 111 that operates on a list of strings. To extend the learned transformer to work over a list, the
 112 user can use the `as map` function.

```
113 > val finalScript = F as map
```

114 If a function G has a type signature $G : T_1 \rightarrow T_2$, then applying the postfix operator `as`
 115 `map` will result in $G \text{ as map} : \text{List}(T_1) \rightarrow \text{List}(T_2)$. With `as map`, the user creates the final
 116 script which takes as input a list of file names and creates a list of HTML links.

117 Beyond the string transformation used above, StriSynth can also learn other types of
 118 functions from examples. StriSynth supports a *filter* function that takes a list of strings as
 119 input and removes some elements based on the filtering criterion. Similarly, StriSynth also
 120 supports learning a *partition* function takes as input a list of strings, and divides them into
 121 groups based on the partitioning criterion. Those groups are then returned as a list of lists
 122 of strings. This functions can be used in any way by the user, but are particularly useful
 123 for scripting tasks that require operations on certain types of files, or files matching some
 124 naming pattern.

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125 In addition, StriSynth can learn a *reduce* function that merges the elements in a list into
126 a single string. StriSynth’s *split* function does the opposite: it returns a list of strings from
127 the input string. These types of functions are especially useful for scripting tasks that apply
128 operations to collections of files.

129 **3 Methodology**

130 A recent survey of the key challenges facing formal methods cites the need for more user
131 studies, especially on real-world users [15]. To test the impact PBE on real users, we
132 recruited 27 IT professionals, all of whom were 18 years of age or older. All materials
133 for the study, as well as the raw data results from the study are available open source at
134 <https://github.com/santolucito/StriSynthStudy>.

135 Our study design consisted of four stages:

- 136 1. A tutorial on both PowerShell and StriSynth that introduced the paradigm and syntax
- 137 2. Complete three scripting tasks (*Extract filenames* from a directory listing, *Move files*
138 with **.png* to *imgs/*, *Printing pdfs* from a list of various file types) in PowerShell
- 139 3. Complete the same three scripting tasks in StriSynth
- 140 4. A post-study survey

141 In the study, participants were told that they would be using the tools StriSynthA
142 and StriSynthB instead of StriSynth and PowerShell to avoid bias from participants’ prior
143 experience. The participants were randomly split into two groups, group A and group B,
144 where the two groups switched the order of steps 2 and 3 of the study to account for any
145 potential bias in earlier exposure to the tasks. Group A completed the tasks with PowerShell
146 first (N=12) and group B completed the tasks with StriSynth first (N=15). The entire study
147 generally took each participant 50 minutes, and the study was conducted in-person with a
148 researcher present. The scripting tasks were completed on the researcher’s laptop, which was
149 preloaded before each study with directories and files needed for the scripting tasks.

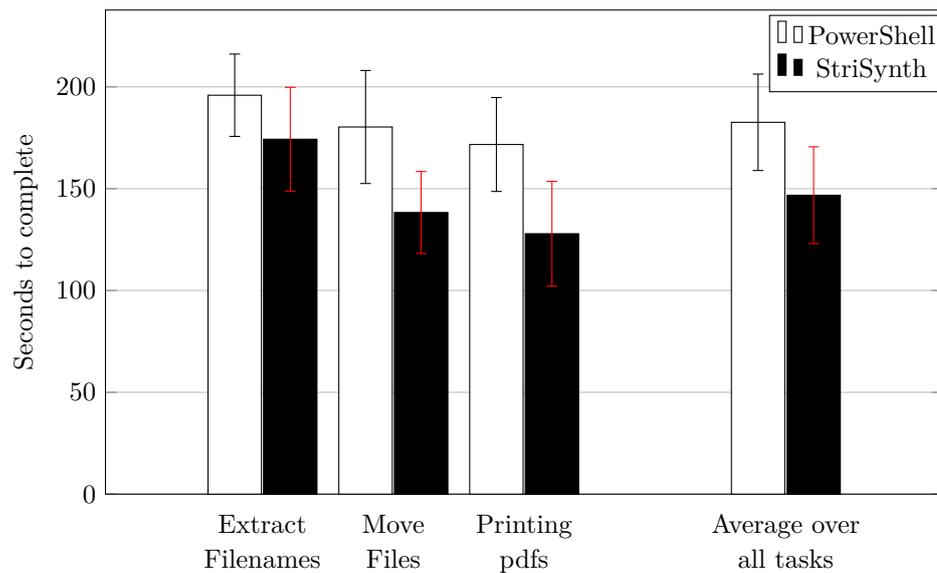
150 While each user was participating in the study, the researcher present recorded the overall
151 time that was used to complete each task. Following the completion of the six tasks, each
152 user was given a questionnaire. The questionnaire measured various responses: prior coding
153 experience, perceived helpfulness of each program as a whole, and perceived helpfulness of
154 each program for each specific task they completed.

155 **4 Results**

156 In this section we present the results of the user study described in Sec. 3. Overall, users
157 completed the tasks more quickly when using StriSynth as opposed to PowerShell. This is
158 good evidence that StriSynth is an efficient tool, especially as none of our users had used
159 StriSynth before this study, while some already had experience with PowerShell. However,
160 despite this concrete measure of efficiency for StriSynth, users said that they believe that
161 PowerShell is a more helpful tool.

162 **4.1 Time to complete the user study tasks**

163 To estimate the usefulness of the programming by example tool StriSynth, we recorded the
164 time it took for users to complete each task with both StriSynth and PowerShell. The results
165 are shown in Fig. 1. In addition, Fig. 1 also contains standard error, depicted with line bars.



■ **Figure 1** The amount of the time each task took, as well as the average time over all tasks for all users (N=27). The smaller bars indicate standard error.

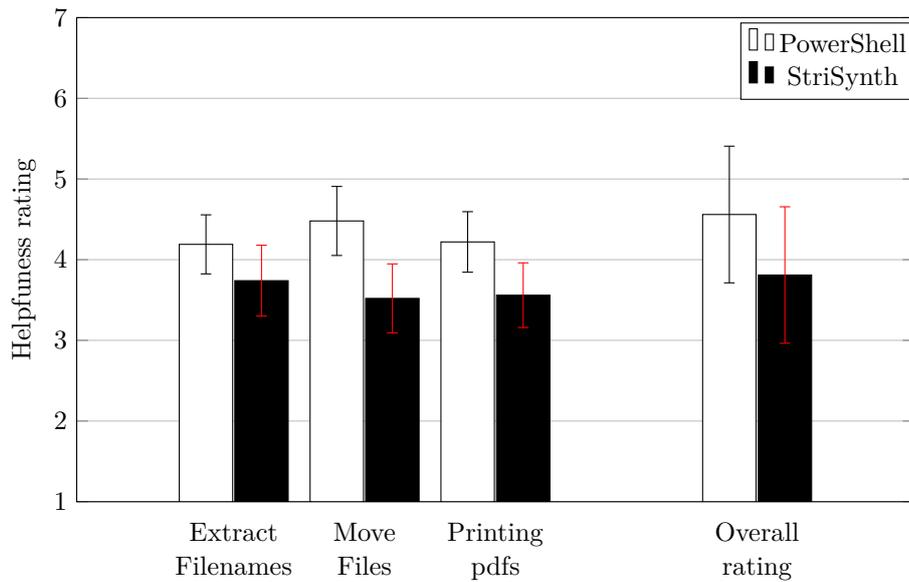
166 In the case of the first task (extracting filenames), from in Fig. 1 the standard error bars
 167 give us the intuition that true mean of the time it takes for overall population to complete
 168 this task using PowerShell is between 170 and 210 seconds. The smaller the standard error,
 169 the more likely is that we have achieved the exact, true value of the mean time, which it
 170 takes for the entire population of IT professionals to complete the tasks.

171 We can see in Fig. 1 that overall the users took less time to complete the tasks with
 172 StriSynth. However, our sample size was relatively small (N=27). Therefore, we wanted
 173 to measure the confidence that our observations are reflective of the larger IT population
 174 beyond our small sample size. To do this we ran a paired sample *t*-test [32].

175 When running the paired sample *t*-test, we are checking the null hypothesis that the
 176 difference between the paired observations in the two samples is zero. Without going into
 177 the details of statistical methods, we need to compute the *p*-value. Any *p*-value of less than
 178 .05 is called *statistically significant*, indicating we have met a generally accepted threshold of
 179 confidence in our results [32].

180 By running these tests on our samples, we learn that a statistically significant difference
 181 was found in the *Move Files* ($p = .03$) and *Printing pdfs* ($p = .02$) tasks. The *p*-value of .03
 182 means that, assuming StriSynth does *not* actually have any impact on time to complete the
 183 *Move Files* task, there is only a 3% chance that we could have observed the timing difference
 184 (or even some larger difference) between StriSynth and PowerShell presented Fig. 1. In other
 185 words, given these low *p*-values, we can be confident that using StriSynth does in fact have
 186 an impact on time to complete the task.

187 All together, our results support the claim that, for small scripting tasks of the type we
 188 presented to our users, PBE can be a more efficient programming paradigm. This is the
 189 expected result that is in line with the literature [16].



■ **Figure 2** Users’ (N=27) self reported measure of the helpfulness of each tool with standard error bars.

190 4.2 Reported helpfulness

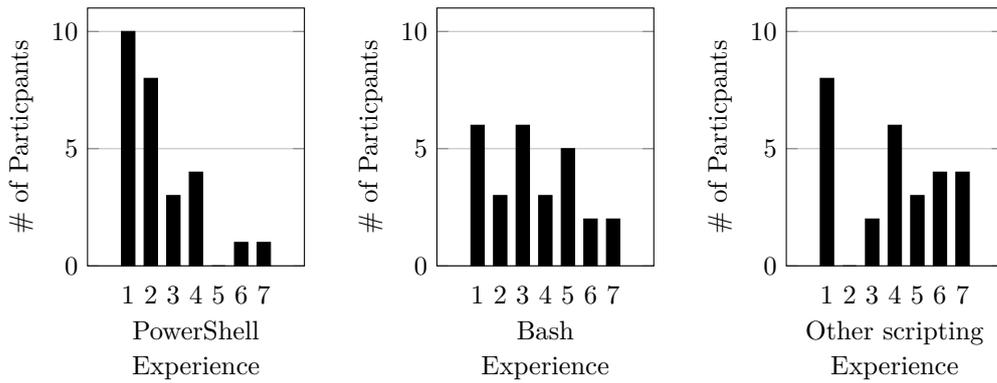
191 Reaching beyond traditional measures for PBE, at the end of the study we also asked users
 192 to report how “helpful” they found both StriSynth and PowerShell. At this point, users did
 193 not know how long they took to complete the tasks with each of the tools. Users were asked
 194 the rate the helpfulness only based on their experience of using the tools during the study.
 195 The exact questions asked were “The following program was helpful for scripting/completing
 196 *Extract Filenames/etc...*”, and users were asked to respond on a scale from 1 (strongly
 197 disagree) - 7 (strongly agree). We show the results from this survey question in Fig. 2, again
 198 with standard error bars. Users rated PowerShell as more helpful in all three tasks, with the
 199 *Move Files* task showing the most significant difference ($p < .01$).

200 The results in Fig. 2 show the surprising insight that, despite the efficiency of StriSynth as
 201 demonstrated in Fig. 1, users perceived PowerShell to be the more helpful tool. Unfortunately,
 202 as we did not anticipate such unexpected results, our study design did not include a more
 203 detailed definition of helpfulness, or ask users to give a more detailed description of their
 204 interpretation of what it means for a tool to be helpful. However, we can at least surmise from
 205 the results presented here, that efficiency is not a complete proxy measure for helpfulness of
 206 a tool.

207 4.3 Impact of prior user experience

208 Prior work observed that familiarity can be a stronger indicator of user preference than
 209 efficiency in the development of programming languages [31]. Our study asked users to self-
 210 report their prior experience with scripting languages in a post-study survey to understand
 211 the impact of user familiarity. The survey used a seven-point Likert scale for users assess the
 212 users’ prior experience. Fig. 3 shows the distribution of experience in three categories for all
 213 users.

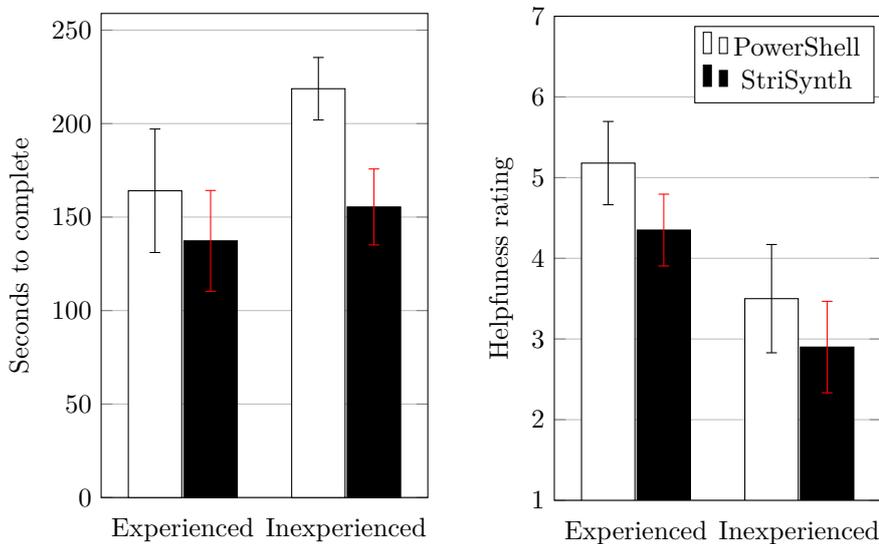
214 To understand the impact of prior experience on how the users interacted with StriSynth,



■ **Figure 3** Users’ (N=27) self reported prior experience with various scripting languages from 1 (Unfamiliar) to 7 (Expert User).

215 we split our user population into two categories. We have the inexperienced user group,
 216 which is the users who rated their prior experience with PowerShell as a 1 (unfamiliar), and
 217 the complement set of users as the experienced user group, who rated their prior experience
 218 with PowerShell as (≥ 2). In Fig. 4, we show how these two groups performed in the study.

219 Fig. 4 shows that both groups of users completed the tasks faster with StriSynth. A more
 220 subtle and interesting insight is that inexperienced users had a greater relative speedup in
 221 task completion when using StriSynth. That is, inexperienced users benefited more from
 222 using StriSynth as compared to the benefit to experienced users. This provides evidence
 223 for the widely stated perception that programming by example is a domain well-suited for
 224 novice programmers.



■ **Figure 4** We grouped users as Experienced (PowerShell experience ≥ 2 , N=17) and Inexperienced (PowerShell experience=1, N=10). We report average time to complete the tasks, and self reported helpfulness of the tools, as separated by these two groups.

225 4.4 Threats to Validity

226 In a usability study, it is important to avoid any possible selection bias in the call for
227 participants. Selection bias can be an issue if the set of users selected systematically differs
228 from the target population. The results we have presented are from a set of users that work
229 as professional IT support specialists. We do not believe that we have any selection bias here
230 because in this work, we specifically wanted to explore the impact synthesis can have in the
231 real-world on such professionals.

232 A further potential threat to the validity of our results is in the social desirability bias, or
233 need-to-please phenomena, whereby users will subconsciously try to produce the results they
234 expect the researcher would like to see. This potential bias can occur when users are asked
235 to compare a tool that is a known standard with an alternative that the user knows to be
236 developed by the researcher. To combat this issue, we presented StriSynth and PowerShell
237 as tools named StriSynthA and StriSynthB respectively. In this way, we framed the study as
238 a comparison between two different tools that we had developed, eliminating the potential
239 need-to-please bias. This was a critical component to our study design that allowed us to
240 observe the disconnect between efficiency and users’ perceived helpfulness of each tool.

241 5 Discussion

242 A key question from these results remains - how are efficiency and helpfulness different in the
243 eyes of the users, and why is this difference manifest in our study? One tempting explanation
244 is that this is the result of two vastly different interfaces. PowerShell is an industrially
245 developed tool, while StriSynth is a research prototype. However, both are command line
246 utilities with a qualitatively similar user experience. Another possible point of departure is
247 in the ability of the user to understand the function of a synthesized script from StriSynth.
248 Trust in the result of program synthesis is a direction that needs further exploration, but
249 StriSynth is unique in this respect in that it provides an English text explanation of the
250 synthesis result. Another possible interpretation may be tied to the expressivity of the
251 paradigm - StriSynth and other PBE tools are generally limited in their ability to directly
252 work with a traditional programming language and use familiar concepts such as variables
253 and loops. This may make a language seem less helpful for new users.

254 Finally, the results from our user study are specifically targeted at the impact of pro-
255 gramming by example systems for scripting in IT professional populations. We must also
256 consider how our results can be interpreted and extended to other PBE domains and program
257 synthesis more generally.

258 5.1 Application to Related Work

259 Gulwani et al. [11] show that PBE is an effective paradigm for industrial application in
260 spreadsheet manipulation, such as string transformations [1, 9], table transformations [10]
261 and database look-ups [29]. Another approach is based on the abstraction of ‘topes’ [28],
262 which lets users create abstractions for different data present in a spreadsheet. With topes, a
263 programmer uses a GUI to define constraints on the data, and to generate a context-free
264 grammar that is used to validate and reformat the data. These application domains of PBE
265 are focused on a similar population of non-expert programmers, and so it may be possible to
266 observe a similar efficiency vs helpfulness phenomena.

267 Unlike programming by example, in which the user provides input-output examples,
268 programming by demonstration is characterized by the user providing a complete trace

269 demonstration leading from the initial to the final state. There are several programming
270 by demonstration systems [6], such as Simultaneous Editing [26] for string manipulation,
271 SMARTedit [22] for text manipulation and Wrangler [17] for table transformations. As
272 programming by demonstration requires intermediate configurations instead of just input and
273 output examples, this paradigm is usually less flexible [21] than programming by example,
274 but the synthesis problem is easier. Based on our results here, it is possible that this
275 reduced flexibility may indicate users would rate programming by demonstration even less
276 helpful (but possible more efficient) than PBE in certain domains. There has been work to
277 overcome the limited expressivity of programming by demonstration by combining program
278 synthesis with direct manipulation of output [24]. This approach may present a way to
279 resolve the disconnect between efficiency and helpfulness as it allows users to interact in both
280 a traditional programming style as well as with synthesis.

281 The Myth [27] and Λ^2 [7] systems support PBE for inductively defined data-types in
282 functional languages. In contrast to StriSynth which focuses on scripting tasks, these tools
283 are focused on synthesis for more general purpose programming languages. The results from
284 our study may be cautiously extrapolated other domains - while the theme of PBE is the
285 same, interaction preference for users may differ when looking at general purpose languages.

286 Instead of providing specification in terms of examples or demonstrations, specification can
287 also be given in more formal and complete ways. InSynth [14, 13], CodeHint [8] and the C#
288 code snippets on demand [33] are systems that aim to provide code snippets based on context
289 such as the inferred type or the surrounding comments. Leon [20] and Comfusy [19, 18]
290 synthesize code snippets based on complete specifications, which are written in the same
291 language that is used for programming. Sketch [30] takes as input an incomplete program
292 with holes, and synthesizes code to complete the so that it meets the specification. These
293 techniques provide a more nuanced interface that may seem, from a perspective of helpfulness,
294 to be more similar to a traditional language paradigm.

295 **6** Conclusions

296 Our study shows that users do not always correlate an efficient programming paradigm with
297 a helpful paradigm. A more thorough exploration of this finding requires a follow up study, in
298 particular to discover the definition of helpfulness that participants are using. A key question
299 to answer would be whether users had erroneously perceived PowerShell to be more efficient
300 and therefore helpful, or if users consciously have other metrics in mind that constitute the
301 helpfulness of programming paradigm.

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