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 Data exploration tools based on code have many desirable characteristics. They can easily access a wide range of different data sources, result in reproducible scripts and encourage users to reuse and modify existing code. Unfortunately, most programming tools require expert coding skills. Can we make data exploration based on code accessible to non-experts? We present The Gamma, a novel text-based data exploration environment that answers the question in the affirmative. The Gamma is based on a novel interaction principle, *iterative prompting*, which lets users create transparent and reproducible scripts without writing code. The Gamma lowers the barrier to entry and learning from previously created data analyses. We evaluate the usability of The Gamma through a user study on non-technical employees of a research institute. Our work shows that we may not need to shy away from code in order to build accessible, reproducible and transparent tools that will allow a broad audience to benefit from the rise of open data.

CCS Concepts: • Human-centered computing \rightarrow Interaction paradigms; • Software and its engineering \rightarrow Integrated and visual development environments; *Domain specific languages*.

Additional Key Words and Phrases: data exploration; end-user programming; data journalism; programming languages; type providers

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1 INTRODUCTION

Despite the advances on visual tooling, programmatic data exploration remains the choice of expert analysts. It is flexibile, offers greater reusability and leads to transparent analyses. We aim to make programmatic data exploration accessible to a wider range of users. The design of a data exploration tool that would make this possible poses a number of challenges. First, the tool needs to have a low barrier to entry. Second, it needs to support a wide range of data sources in a uniform way. Third, the users should be able to learn how to use the tool by looking at existing data analyses.

We contribute The Gamma, a text-based data exploration environment for non-experts. The Gamma is based on a single, easy to understand interaction principle and provides a uniform access to a range of data sources including data tables, graph databases and data cubes. The resulting analysis is a transparent script that can be followed to reproduce the result from scratch. This allows learning from existing analyses and encourages readers to engage with data.

Iterative Prompting. The main contribution of our work is the *iterative prompting* interaction principle, which makes it possible to construct all valid data exploration scripts by repeatedly choosing an item from a list of options offered through auto-complete. The design favors *recognition over recall* and allows non-programmers to write entire scripts without typing code and without learning a programming language first. Yet, it still results in transparent and reproducible code. In other words, iterative prompting turns auto-complete from a programmer assistance tool into a non-expert programming mechanism. A crucial feature is that iterative prompting only offers operations that are valid in a given context and that it offers all such operations; it is both correct and complete.

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Fig. 1. Obtaining teams with the greatest number of gold medals from Rio 2016 Olympics with a reproducible The Gamma script (1), contextual iterative prompting mechanism offering ways of sorting the data (2) and an instant preview of results (3).

Data Exploration. The Gamma focuses on data exploration of the kind illustrated in Figure 1. The user accesses data available in a structured format. They make several experiments to find an interesting way of looking at the data, e.g. by applying different aggregations or filters. They may choose to view the results as a table or a basic chart before publishing their analysis. The Gamma makes such programmatic data exploration simple enough for non-experts, but scraping and cleaning of messy data or building custom data visualizations is currently outside of the scope of our work. Exposing those using iterative prompting remains an interesting and worthwhile future challenge.

Paper Overview. The Gamma is available (non-anonymously) at http://thegamma.net, both as a JavaScript library and a hosted data exploration service. In this paper, we describe and evaluate the design principles behind the project¹:

- We introduce the iterative prompting principle in The Gamma and show how it can be used for querying of distinct data sources including data tables, graph databases and data cubes (Section 3).
- We evaluate the system through a number of case studies (Section 4) and a user study (Section 5), which confirms that non-programmers can use The Gamma to construct non-trivial data queries.
- We reflect how our design lowers barriers to entry, supports learning without experts and offers a complete and correct program construction method (Section 6).

2 RELATED WORK

The key contribution of our work is that it develops a new, fundamentally different, way of using the established auto-completion mechanism. Unlike most past work dating back to Kaiser [24], we do not view it as a programmer assistance tool. Instead, we turn it into an interaction mechanism through which non-experts can create entire programs. We build on recent research on information-rich programming [53] and aim to make those advances available to non-programmers [35, 37], in the context of data exploration as done, for example, by journalists [15]. Our work features a novel combination of characteristics in that our iterative prompting interaction principle (i) is centered around editing and understanding of code, (ii) reduces conceptual complexity to a single basic kind of interaction, yet (iii) it is correct and complete in that it can be used to construct all meaningful queries for a variety of data sources.

¹⁰³ ¹This paper is 11 pages (5500 words), without citations, to be commensurate with the size of the contribution.

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Code Completion for Data Science. A key component in The Gamma is the use of auto-complete for offering possible
 operations. Our work follows type providers [41, 53], which integrate external data into a static type system of F#,
 allowing the use of auto-completion; for querying data tables, we utilize the theory developed by Petricek [39]. The key
 difference in our work is that The Gamma can be used without a programming language expertise.

Most similar to our approach are tools that make recommendations when users begin interacting with data. Those based on machine learning-based code completion for domain specific languages [17, 18] differ in that they do not guarantee completeness, i.e. it is unclear whether the user can create all possible scripts. Approaches based on natural language are effective [45, 50], but hide the underlying structure and do not help the user understand it. Conversational agents [13] share similar characteristics, except that the construction process is iterative.

Code completion based on machine learning or statistical methods [5, 43] also exists for general-purpose programming languages used by data scientists such as Python [52], providing assistance to expert programmers. Finally, DS.js [59] is interesting in that it enables querying of data on the web. It uses JavaScript, but with rich contextual code completion.

120 Notebooks and Business Intelligence Tools. Notebooks such as Jupyter [28], which allow combining source code with 121 commentary and visual outputs, are widely used by data scientists, but require expert programming skills. The Gamma 122 123 targets non-experts, but could also be integrated with a multi-language notebook system [40]. Spreadsheets, business 124 intelligence tools [33, 58] and other visual data analytics tools [8, 19] do not involve programming, but require mastering 125 a complex GUI. In contrast, The Gamma is based on a single kind of interaction, through which all available operations 126 can be completed. Several systems [25, 42, 47] record interactions with the GUI as a script that can be modified by the 127 128 user. Unlike in The Gamma, the source code does not guide the user in learning how to use the system.

Easier Programming Tools. We aim to build an easy to use and learn programming system. Many approaches to 130 131 this goal have been tried. Victor [56] introduced design principles that inspired many to build live programming 132 systems [14, 29, 44] that give immediate feedback to help programmers understand how code relates to output and 133 exploratory systems [26, 27] that assist with completing open-ended tasks. A system combining textual language 134 with visualization also exists for graph querying [2]. To avoid difficulties with editing code as text, some systems 135 use structured editors [32, 38, 54]. In Subtext [11, 12] the language itself is co-designed with the editor to make the 136 137 interactions with code more natural. The Gamma is live in that our editor gives an instant preview of the results. 138 Many systems simplify programming by offering high-level abstractions, e.g. for interactive news articles [7], statistical 139 analyses [23] or interactive data visualization [48, 49]. The Gamma exposes a number of data sources through high-level 140 141 abstractions that support iterative prompting, but support for tasks other than querying remains furture work. 142

Programming without Writing Code. There are two main approaches to programming where the user does not write code. In programming by example [31], the user gives examples of desired results. This has been used, e.g. for specifying data transformations in spreadsheets and data extraction [16, 30]. In direct manipulation [21], a program is specified by directly interacting with the output. This has been used in the visual domain [20], but also for data querying [4, 51]. The VQE language [10] also considers how to allow code reuse and modification in this context. Direct manipulation can also support data exploration by letting users partially edit queries, e.g. by changing quantifiers as in DataPlay [1].

Gestures and Data Entry. Although our focus is on program construction, our work can be positioned in the broader
 context of input methods. Akin to Dasher [57], our system provides a way of navigating through a complete space of
 options, while on-screen feedforward [3] allows efficient selection in gesture-based interfaces. Those provide compelling
 alternatives to auto-completion menus, although the efficiency of input methods is typically not an issue in programming.

157 3 OVERVIEW

The Gamma aims to make text-based data exploration easy enough for non-experts. The aim is motivated, in part, by the desirable properties of text-based data exploration tools such as transparency, reproducibility and learnability and, in part, by an aim to explore an unexplored point in the design space of data exploration tools. Although various efforts make text-based programming easier, most systems that target non-experts shy away from code.

The Gamma is a text-based data exploration environment that allows non-experts explore data using iterative prompting – by repeatedly selecting an item from an auto-complete list. The study presented in Section 5 confirms that the kind of data exploration shown in the next section can, indeed, be successfully done by non-experts.

3.1 Querying Travel Expenses

The walkthough in Figure 2 shows a typical task completed using The Gamma. A data analyst from Kent is exploring travel expense claims by members of the House of Lords published by the UK government [55]. The following shows a subset of the data in the CSV format:

	🖋 drop colum	20					Name	0	Davis Athended	D	Travel C
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	≯ paging ≯ sort data						Lord Harris of Peckham	Kent	3	0	0
	🔑 windowing						Baroness Noakes	Kent	б	0	135
2						(5)	Bishop of Rochester	Kent	1	0	10
	expenses .'filter data' ≁Kent	.'County	is'.ke			U	expenses .'filter data' .'sort data'.'l		is'.Kent.then		
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3	.'filter data'. Name Lord Astor of Hever	County Kent	Days Attended	0	85	6	<pre> then expenses 'sort data'.'b .'filter data'paging.take(5) .'get series'.' </pre>	oy Travel 'County :	Costs descend is'.Kent.then	-	Costs
3)	Name Lord Astor of Hever Lord Freud	County Kent Kent	Days Attended 7 1	0	85 0	6	<pre>✓ then expenses .'sort data'.'b .'filter data'paging.take(5) .'get series'.' Lord Swinfen Baroness Noakes</pre>	oy Travel 'County :	Costs descend is'.Kent.then	-	Costs
3)	Name Lord Astor of Hever Lord Freud Lord Harris of Peckham	County Kent Kent Kent	Days Attended 7 1 3	0 0 0 0	85 0 0	6	<pre>✓ then expenses .'sort data'.'t .'filter data'paging.take(5) .'get series'.' Lord Swinfen Baroness Noakes Baroness Suttie</pre>	oy Travel 'County :	Costs descend is'.Kent.then	-	Costs
3	Lord Astor of Hever Lord Astor of Hever Lord Freud Baroness Noakes	County Kent Kent Kent Kent	Days Attended 7 1 3 6	0 0 0 0 0 0 0	85 0 0 135	6	<pre>✓ then expenses .'sort data'.'b .'filter data'paging.take(5) .'get series'.' Lord Swinfen Baroness Noakes</pre>	oy Travel 'County :	Costs descend is'.Kent.then	-	Costs

Fig. 2. Constructing a script that charts the top 5 members of the House of Lords for Kent, based on their travel costs.

²⁰⁹ Name, County, Days Attended, Days Away, Travel Costs

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<sup>210</sup> <sub>2</sub> Lord Adonis, London, 8, 0, 504
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3 Baroness Afshar, Yorkshire, 2, 0, 0

Lord Alderdice, Oxfordshire, 3, 0, 114

5 Lord Alli, London, 5, 0, 0

The analyst imports the file through a web interface, the environment is initialised with code that refers to the imported data as expenses and she starts exploring the data using the type provider for tabular data (Section 3.3):

- (1) The analyst types '.' (dot) to trigger auto-completion on expenses. The type provider offers a list of operations that the analyst can perform. To find House of Lords members from Kent, the analyst chooses filter data.
- (2) The analyst is offered a list of columns based on the schema of the tabular data and choses County is. She is then offered a list of counties in the data set and types ke to search for Kent and she selects Kent.
- (3) The Gamma evaluates the code on-the-fly and shows a preview of results. The analyst now sees a table with House of Lords members from Kent. She wants to see if there are any members who missed any House sessions.
- (4) The analyst finishes specifying the (possibly compound) sorting key by choosing then and is offered the same list of querying operations as in the first step. She selects sort data followed by by Days Away descending.
- (5) The analyst sees that there are no reported "days away" and decides to compare travel costs. She hits the backspace key a number of times, is offered the list of keys again and selects by Travel Costs descending.
- (6) The analyst chooses then and is, again, offered the list of querying operation. She uses paging to get top 5 records, which requires typing 5 as the argument. She then uses the get series operation to obtain a data series associating travel expenses with a name, which is automatically visualized using a bar chart.

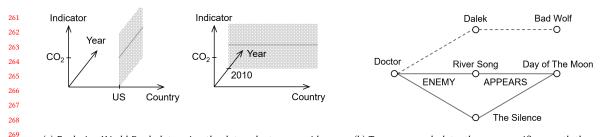
The constructed code is not unlike an SQL query, except that the whole script is constructed using iterative prompting, by repeatedly selecting one of the offered members. Those represent both operations, such as sort by and arguments, such as Kent. The only exception is when the analyst needs to type the number 5 to specify the number of items to take.

3.2 The Gamma Programming Environment

A program in The Gamma is a sequence of commands. A command can be either a variable declaration or an expression that evaluates to a value such as a data table or a chart. An expression is a reference to a data source followed by a chain of member accesses. A member can be either an ordinary member such as paging or an operation which takes a list of parameters enclosed in parentheses as in take(5). Names with non-alphanumerical characters are escaped using quotes.

The Gamma uses a type system to infer what members are available at a given point in a chain. Each expression has a type with a list of members that, in turn, have their own types. The types are not built-in, but are generated by type providers for individual data sources. The programming environment for The Gamma is based on the Monaco editor [34]. When the user types '.' the editor triggers auto-completion and retrieves a list of available members based on the type information. The Gamma evaluates scripts on-the-fly and shows a preview as illustrated in Figure 1.

There is a handful of situations where The Gamma does not yet fully support the iterative prompting principle. First, it allows operations with parameters such as take(5). This is currently needed when writing a query that skips or takes the first N elements from a table. Second, The Gamma allows the user to declare (immutable) variables using let. This is not needed for basic data exploration, but allows advanced users to better structure more complex code.



(a) Exploring World Bank data using the data cube type provider, users choose values from two dimensions to obtain a data series.

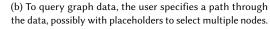


Fig. 3. Design of type providers for exploring data cubes and graph databases.

3.3 Type Providers for Data Querying

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The Gamma can be extended to support any data source by implementing a *type provider*, which defines a domain specific language for exploring data of a particular kind. A type provider generates object types with members (such as paging or Kent) that are accessed via iterative prompting. We describe type provider for exploring data cubes (inspired by Syme et al. [53]), tabular data (based on theory developed by Petricek [39]), and graph databases.

Data Cube Type Provider. Our first type provider allows users to explore data cubes, which are multi-dimensional arrays of values. For example, the World Bank collects a range of indicators about many countries each year while the UK government expenditure records spending for different government services, over time, with different adjustments:

- worldbank.byCountry.'United States'.'Climate Change'.'CO2 emissions (kt)'
- 2 expenditure.byService.Defence.inTermsOf.GDP

The dimensions of the worldbank cube are countries, years and indicators, whereas the dimensions of expenditure are government services, years and value type (adjusted, nominal, per GDP). Figure 3a how the provider allows users to slice the data cube. Choosing byCountry. 'United States', restricts the cube to a plane and selecting 'CO2 emissions (kt)' then gives a series with years as keys and emission data as values. Similarly, we could first filter the data by a year or an indicator. The same mechanism is used to select UK government spending on defence in terms of GDP.

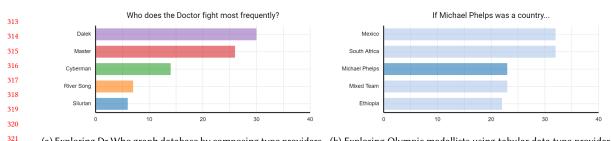
Tabular Data Type Provider. Our second type provider allows users to construct queries to explore data in tabular
 formats. Unlike the data cube provider, the provider for tabular data does not just allow selecting a subset of the data,
 but it can be used to construct SQL-like query. Consider the example from Figure 1:

- 1 olympics.'filter data'.'Games is'.'Rio (2016)'.then
 - 2 .'group data'.'by Team'.'sum Gold'.'sum Silver'.then
- 302 3 .'sort data'.'by Gold descending'

303 The example queries a table that records individual medals awarded in Olympic games. The chain constructs a query that 304 selects rows corresponding to the Rio 2016 Olympics and then calculates total number of gold and silver medals for each 305 team (country) before sorting the data. When using the provider, the user specifies a sequence of operations. Members 306 307 such as 'filter data' or 'group data' determine the operation type. Those are followed by operation parameters. For 308 example, when grouping data, we first select the key and then choose a number of aggregations to calculate over the 309 group. Unlike SQL, the provider only allows users to choose from pre-defined aggregations such as calculating the sum, 310 average or the number of distinct values. Section 4 shows that this is sufficient to construct a range of practical queries. 311

Anon

Woodstock '18, June 03-05, 2018, Woodstock, NY



(a) Exploring Dr Who graph database by composing type providers (b) Exploring Olympic medallists using tabular data type provider Fig. 4. Charts produced by two case studies of using The Gamma.

Graph Database Type Provider. Our third type provider allows users to explore data from graph databases, which store nodes representing entities and relationships between them. The following example explores a database of Doctor Who characters and episodes. It retrieves all enemies of the Doctor that appear in the Day of the Moon episode:

drwho.Character.Doctor.'ENEMY OF'.'[any]'.'APPEARED IN'.'Day of the Moon'

We start from the Doctor node and then follow two relationships. We use 'ENEMY OF'.'[any]' to follow links to all enemies of the Doctor and then specify 'APPEARED IN' to select only enemies that appear in a specific episode. The result appears in in Figure 3b. The provider works with any graph database and generates members automatically, based on the data. In the above, ENEMY OF and APPEARED IN are labels of relations and Doctor and Day of the Moon are labels of nodes. The [any] member defines a placeholder that can be filled with any node with the specified relationships. The results returned by the provider is a table of properties of all nodes along the specified path. As illustrated by an example discussed in Section 4, the returned table can be further queried using the tabular data type provider.

4 CASE STUDIES

The Gamma aims to simplify programmatic data exploration while keeping enough expressive power to allow users to create interesting data explorations. In this section, we consider two case studies that evaluate expressivity and show what can be achieved using the simple iterative prompting principle². We used The Gamma for larger projects exploring the UK government expenditure, activities of a research institute adn Olympic medal winners³.

The Most Frequent Doctor Who Villains. Our first case study uses a graph database with data from the Dr Who series. It lists Dr Who villains by the number of episodes in which they appear. This case study is interesting as it combines the graph database provider for fetching the data with the tabular data provider for summarization:

- 351 drWho.Character.Doctor.'ENEMY OF'.'[any]'.'APPEARED IN'.'[any]'.explore
- 352 2 .'group data'.'by Character name'.'count distinct Episode name'.then
- 353 .'sort data'.'by Episode name descending'.then
- 354 4 .paging.take(8).'get series'.'with key Character name'.'and value Episode name'
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Line 1 use the graph provider to find all paths linking the Doctor with any character linked via ENEMY OF, followed by any episode linked by APPEARED IN. This produces a table that can be analysed using the tabular data provider by selecting explore. For each character (the villain) we count the number of distinct episodes. The result is shown in Figure 4a. Despite performing a sophisticated data analysis that involves a graph database query, followed by an SQL-like data aggregation, the code can be constructed using iterative prompting, with the exception of the numbers in paging.

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³⁶² ²Available (non-anonymously) at: http://gallery.thegamma.net/86/ and http://gallery.thegamma.net/87/, respectively.

³⁶³ ³Available (non-anonymously) at http://turing.thegamma.net and http://rio2016.thegamma.net

Woodstock '18, June 03-05, 2018, Woodstock, NY

If Michael Phelps were a Country. Michael Phelps has won so many medals that media compared the number to
 countries [36], often using a chart that shows a country league table including Michael Phelps as an additional data
 point. We reproduce the chart, shown in Figure 4b, using the tabular data type provider:

- 369 1 let data = olympics.'group data'.'by Team'.'sum Gold'.then
 - 2 .'sort data'.'by Gold descending'.then
 - 3 .paging.skip(43).take(4).'get series'.'with key Team'.'and value Gold'
 - 4
 - 5 let phelps = olympics.'filter data'.'Athlete is'.'Michael Phelps'.then
 - 6 .'group data'.'by Athlete'.'sum Gold'.then
 - 7 .'get series'.'with key Athlete'.'and value Gold'
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9 charts.bar(data.append(phelps)).setColors(["#aec7e8","#aec7e8","#1f77b4"])

The data analysis is done in three commands. The first counts gold medals by countries and uses paging to fetch 4 countries with suitable number of medals. In the second, we use the grouping operation to aggregate data for just a single group. The two data series are then assigned to local variables (for readability) and passed to the chart.columns function. The example illustrates a case when more advanced language features are necessary. The data exploration itself has been completed via iterative prompting, but producing the final chart currently requires some manual programming.

5 USER STUDY

³⁸⁸ Using the characterization by Olsen [22], data exploration environments are complex systems that do not yield to simple
 ³⁹⁰ controlled experimentation. Consequently, our goals are modest and we do not attempt to quantitatively compare our
 ³⁹¹ work with other tools. We merely aim to study whether The Gamma can be successfully used by non-programmers.

We conducted a study in which we gave volunteers one of four data exploration tasks and assessed whether they were able to complete the task and how much assistance, if any, they needed. Some aspects of the study offer insights into how users learn and understand The Gamma. We summarize those points, but do not claim conclusive results.

5.1 Study Design

We performed a between-subjects study to evaluate the first experience of using The Gamma. We recruited 13 participants (5 male, 8 female) from a business team of a research institute working in non-technical roles (project management, partnerships, communications). Only one participant (#12) had prior programming experience. We split participants into 4 groups and asked each group to complete a different task. We gave participants a brief overview of The Gamma (content depended on the task). The participants worked for 30 minutes, after which we conducted a 30 minute semi-structured group interview. We let participants work independently, but offered guidance if they got stuck. The four tasks were:

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- *Expenditure*. Participants were given a demo using *worldbank*. They were asked to use the *expenditure* data source to compare the UK government spending on "Public order and safety" and "Defence" in terms of GDP.
- *Lords*. Participants were given a demo using *worldbank*. They were asked to use the *lords* data source (a table with House of Lords members expenses) to find a members representing London with the highest travel costs.
- *Worldbank*. Participants were given a minimal demo of iterative prompting and a code sample using *worldbank*. They were asked to solve a different task using the *worldbank* data source.
- Olympics. Participants were given a basic demo using olympics. They were asked to solve a more complex
 problem, involving grouping and aggregation, using the same data source.
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	Task	Kind	Done	Notes
#1	expenditure	cube	0	Obtained one of two data series
#2	expenditure	cube	•	Explored furhter data series independently
#3	expenditure	cube	•	Explored further data series independently
#4	expenditure	cube	J	Completed following a hint to use another member
#5	expenditure	cube	•	Explored further data series independently
#6	worldbank	cube	J	Completed after a syntax hint about whitespace
#7	worldbank	cube	•	Completed very quickly
#8	worldbank	cube	•	Completed, but needed longer to find correct data
#9	lords	table	J	Struggled with composition of operations
#10	lords	table	•	Completed very quickly
#11	lords	table	J	With a hint to avoid operations taking arguments
#12	olympics	table	J	With a hint to avoid operations taking arguments
#13	olympics	table	J	With hints about 'then' and operations taking argument

Table 1. Overview of work completed by individual participants in the study.

The marks denote: \bullet = completed, \bullet = required some guidance, \bullet = partially completed

Our primary hypothesis was that non-programmers will be able to use The Gamma to explore data. This was tested by all four tasks for one of the supported data sources. Some aspects of the study shed light on questions concerning the learnability of The Gamma. The tasks *expenditure* and *lords* test if knowledge can be transferred between different data sources by using one sources in the introduction and another in the task. The task *worldbank* explores whether users can learn from just code samples by providing only minimal upfront explanation. The task *lords* lets us study to what extent participants form a correct mental model of the more complex query language used in the tabular data source.

5.2 Study Results

Table 1 summarizes the work done by the study participants. For each participant, we record the task, the kind of data source used and the level of completion. For participants who needed assistance, the notes section details the help given. The experience suggests a number of possible design improvements for The Gamma, which are discussed below.

Can non-programmers explore data with The Gamma? All participants were able to complete, at least partially, a non-trivial data exploration task and only half of them required further guidance. Participants spent 10-25 minutes (average 17 minutes) working with The Gamma and 12 out of 13 completed the task; 6 required assistance, but 3 of those faced an issue related to operations taking arguments (discussed later), which could be addressed in the introduction. A number of participants also shared positive comments in the interviews. For example, participant #3 noted taht "this is actually pretty simple to use. You think about the logic of what you're actually asking and then you try to get it into the format you can." Participant #2 noted that The Gamma alleviated their unease about code: "For somebody who does not do coding or programming, this does not feel that daunting. It's not like you're giving me large screen full of code, which is reassuring."

How users learn The Gamma? There is some evidence that knowledge can be transferred between different data sources.
 Two of the tasks (*expenditure* and *lords*) used different data sources in the introduction and the task. Participants were
 able to complete those, although *lords* has been challenging as it involves a complex data source. Participant #2 also
 shared a positive comment: "I found it quite easy to translate what you showed us in the demo to the new dataset.".

There is also some evidence that users can learn just from code samples. In the *worldbank* task, participants were given only a minimal demo of how to invoke iterative prompting together with print-out of 2 code samples. All three

participants were able to complete a related task using the same data source. When discussing suitable educational
 materials for The Gamma, participant #7 also confirmed that having code is sufficient when they noted that *"a video would just be this [i.e. a code sample] anyway*". This supports our hypothesis that, once a user understands the iterative
 prompting principle, they can learn how to use any specific data source just from code samples.

How users understand complex query languages? The tabular type provider has a rich structure and uses a member 475 476 named then to complete the specification of a current operation, for example when specifying a list of aggregation 477 operations. We asked participants who worked with tabular data (lords or olympics) about their understanding of the 478 then member. Two participants (#12 and #13) initially thought that then is used to split a command over multiple lines, 479 but rejected the idea after experimenting. Participant #12 then correctly concluded that it "allows us to chain together 480 the operations" of the query; after a hint, participant #13 reflected that "if I knew this from the start, it would [have been 481 482 easier]." In summary, iterative prompting allows users to start exploring new data sources, but the structures exposed 483 by more complex data sources have their own further design principles that the users need to understand. 484

What would make The Gamma easier to use? The type provider for tabular data generates operations that take arguments such as take(5) in a number of places. When filtering data, it allows writing olympics. 'filter data'. 'Year is greater than' (2004). Those operations violate the iterative prompting principle as one cannot type '.' after 'Year is greater than'. Three participants (#11, #12, #13) struggled to complete a task, because they initially attempted to use those operations. This suggests that we should either avoid such operations, or hide them under an "advanced operations" tab.

The Gamma uses an ordinary text editor. This has both benefits and drawbacks compared to structured editors [32, 38, 54]. Most participants had no difficulty navigating around code, making edits or deleting fragments, which is arguably harder in a structured editor. Some participants used text editor effectively, e.g. participant #5, who used copy-and-paste to fetch the same data for multiple countries. However, we also observed two issues. Participant #2 struggled with indentation and participant #6 had a syntax error in an unrelated command, which prevents charts from rendering.

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6 DISCUSSION

The Gamma examines an unexplored point in the design space of tools for data exploration. It is a text-based programming environment for non-programmers. Its design has been motivated by a curiosity as to whether iterative prompting can make text-based programming with data accessible to non-experts. In this section, we theoretically assess the resulting design, making use of criteria for judging whether a system advances the state of the art proposed by Olsen [22].

Learning without experts. The Gamma allows non-programmers to produce transparent and reproducible scripts
 that explore data from a wide range of sources. It allows new participants to benefit from the capabilities offered
 by programmatic data exploration, satisfying the *exmpowering new participants* criteria proposed by Olsen [22]. To
 empower new participants, our design aims to make The Gamma suitable for users who cannot dedicate significant
 amount of time to learning it in advance and may not have access to experts. This is supported in two ways.

First, the iterative prompting principle makes it easy for users to start experimenting. The user needs to select an initial data source and then repeatedly choose an item from a list of choices. Iterative prompting is easier to use than e.g. a command line or a REPL (read-eval-print-loop), because it follows the *recognition over recall* usability heuristic. The users are not required to recall and type a command. They merely need to select one from a list of options. Second, the resulting source code serves as a trace of how the analysis was created. It provides the user with all information that they need to recreate the program, not just by copying it, but also by using iterative prompting. Such design has

been termed design for percolation by Sarkar [46], who studies how Excel users learn. He points out that users learn 521 522 new features when the usage of a feature is apparent in a spreadsheet. For example, users can learn different functions 523 in formulas, because those are visible in the cell. Learning how to use a wizard for creating charts is hard because the 524 operation leaves no full trace in the spreadsheet. Design for percolation thus supports learnability. 525

Lowering barriers to entry. Data exploration has a certain irreducible essential complexity. To make a system usable, 527 528 this complexity needs to be carefully stratified. The Gamma uses a two level structure. The first level consists of the 529 language itself with the iterative prompting mechanism. The second level consists of the individual members generated 530 by a type provider. This can be seen as a *domain specific language*, embedded in The Gamma language. Although 531 the complexity of individual domain specific languages differs, the user can always start exploring through iterative 532 prompting, even when faced with an unfamiliar data source. In tackling complexity, The Gamma satisfies two criteria 533 534 proposed by Olsen [22]. The Gamma satisfies the generality criteria in that it can be used uniformly with a wide range 535 of data sources. It also satisfies the expressive leverage criteria in that it factors out common aspects of different data 536 queries into the core language (first level) and leaves the specifics of each data source to the second level. 537

Correctness and completeness. An important characteristic of our design is that the iterative prompting mechanism 539 is both correct and complete with respect to possible data exploration scripts. The two properties are a consequence 540 541 of the fact that a program is a formed by a chain of operations and that the auto-completion leverages a static type 542 system. When invoking iterative prompting at the end of a well-typed script, a selected option, which is a valid object 543 member, is added to the end of the script, resulting in another well-typed script. This distinguishes our system from 544 auto-completion based on machine learning, which may offer members not valid in a given context. Auto-completion 545 546 lists offered via iterative prompting contain all available members and so the user can construct all possible scripts. Two 547 exceptions to completeness in our current design are the let binding and specifying numerical parameters as in take(5). 548

549 Importance of applications. Although The Gamma targets a broad audience of non-programmers, some of our work has 550 been particularly motivated by the use of data in journalism. It has the potential to enable journalists to make factual 551 claims backed by data more commonplace and enable wider audience to engage with such claims. As such The Gamma 552 553 satisfies the importance criteria proposed by Olsen [22]. Although The Gamma is open-source, it has not been deployed 554 in a newsroom so far. This would lead to valuable insights, but it requires finding a suitable fortuitous opportunity. The potential is indicated by a comment from a former journalist who participated in our study (#13): "There's a lot of effort going into data journalism that programming could make much quicker, (...) like this would really simplify things."

7 CONCLUSIONS

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Exploring data in a programming environment that makes the full source code available increases transparency, reproducibility and empowers users to ask critical questions about the data analysis. But can we make those features accessible to non-programmers? In this paper, we presented The Gamma, a simple data exploration environment for non-programmers that answers this question in the affirmative.

The Gamma is based on a single interaction principle, *iterative prompting*. It can be used to complete a range of 566 567 data exploration tasks using tabular data, data cubes and graph databases. The design lowers the barrier to entry 568 for programmatic data exploration and makes it easy to learn the system independently through examples and by 569 experimentation. We implemented The Gamma, make it available as open source and conducted a user study, which 570 lets us conclude that The Gamma can be used by non-programmers to construct non-trivial data exploration scripts. 571

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