

Joker: A Unified Interaction Model For Web Customization

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ABSTRACT

Tools that enable end-users to customize websites typically use a two-stage workflow: first, users extract data into a structured form; second, they use that extracted data to augment the original website in some way. This two-stage workflow poses a usability barrier because it requires users to make upfront decisions about what data to extract, rather than allowing them to incrementally extract data as they augment it.

In this paper, we present a new, unified interaction model for web customization that encompasses both extraction and augmentation. The key idea is to provide users with a spreadsheet-like formula language that can be used for both data extraction and augmentation. We also provide a programming-by-demonstration (PBD) interface that allows users to create data extraction formulas by clicking on elements in the website. This model allows users to naturally and iteratively move between extraction and augmentation.

To illustrate our unified interaction model, we have implemented a tool called Joker which is an extension of Wildcard, a prior web customization system. Through case studies, we show that Joker can be used to customize many real-world websites. We also present a formative user study with five participants, which showed that people with a wide range of technical backgrounds can use Joker to customize websites, and also revealed some interesting limitations of our approach.

1 INTRODUCTION

Many websites do not meet the exact needs of all of their users, so millions of people use browser extensions and user-scripts [4, 5] to customize them. However, these tools only allow end-users to install customizations built by programmers. End-user web customization systems like Sifter [15], Vegemite [18] and Wildcard [19] provide a more accessible

approach, allowing anyone to create bespoke customizations without performing traditional programming.

These tools each provide different useful mechanisms for end-user customization, but they share a common design limitation: they have a rigid separation between the two stages of the web customization process. First, in the *extraction* or *scraping* phase, users get data from the website into a structured, tabular format. Second, in the *augmentation* phase, users perform augmentations like adding new columns derived from the data, or sorting the data. For example, in Vegemite, a user can extract a list of addresses from a housing catalog, and then augment the data by computing a walkability score for each address.

This separation between extraction and augmentation poses an important barrier to usability. A user study [18] of Vegemite wrote that “it was confusing to use one technique to create the initial table, and another technique to add information to a new column.” The creators of Sifter similarly reported [15] that “the necessity for extracting data before augmentation could take place was poorly understood, if understood at all.” In Wildcard, end-users cannot augment a website at all until a programmer has written and shared extraction code for that website in Javascript [19]. These tools all impose a sequential workflow in which users must first extract all the data they need, and then perform all their desired augmentations. This workflow exemplifies the more general problem in interface design of forcing users to make premature commitments to formal structure [8, 22].

In this paper, we present a new approach to web customization that combines extraction and augmentation in a unified interaction model. Our key idea is to develop a domain specific language (DSL) that encompasses both extraction and augmentation tasks, along with a programming-by-demonstration (PBD) interface that makes it easy for end-users to program in the language. This unified interaction model allows end-users to move seamlessly between

extraction and augmentation, resulting in a more iterative and free-form workflow for web customization.

To demonstrate and evaluate this model, we have built a browser extension called Joker, an extension of the Wildcard customization tool. The original Wildcard system [19] adds a spreadsheet-like table to a website and establishes a bidirectional synchronization between the website and the table. This allows users to customize a website, by filtering and sorting page elements, and adding user annotations, derived values and calls to web services. Although Wildcard offers a declarative formula language for augmenting the page, a conventional imperative language (namely JavaScript) is used for the extraction step, and the extraction code cannot be modified during augmentation.

Joker makes two primary contributions:

A unified formula language for extraction & augmentation: Wildcard’s formula language only supported primitive values like strings and numbers aimed at augmentation. Joker extends this language by introducing Document Object Model (DOM) elements as a new type of value, and adding a new set of formulas for performing operations on them. This includes querying elements with Cascading Style Sheets (CSS) selectors and traversing the DOM. With this approach, a single formula language is used to express both extraction and augmentation tasks, even within a single formula expression.

A PBD interface for creating extraction formulas: Directly writing extraction formulas can be challenging for end-users, so Joker provides a PBD interface that synthesizes formulas from user demonstrations. A key aspect of our design is that the program synthesized from the demonstration is made visible as a spreadsheet formula that can be subsequently edited by the user, and is more easily understood than imperative code due to its declarative form.

Section 2 describes a concrete scenario, showing how Joker enables a user to complete a useful customization task. In Section 3, we briefly describe the implementation of our formula language and user interface, as well as the algorithms used by our PBD interface.

We have performed two evaluations of our approach, presented in Section 4. First, we describe a suite of case studies in which we used Joker to extract and augment a variety of websites in order to characterize its capabilities and limitations. Second, we describe a formative user study with five participants, which showed that users were generally able to use Joker to perform useful extraction and augmentation tasks, but which also uncovered limitations, particularly for less experienced users trying to extract data from more complex websites.

Joker relates to existing work not only in end-user web customization, but also in end-user web scraping and program synthesis, which we discuss in Section 5. Finally, we discuss opportunities for future work in Section 6.

2 EXAMPLE USAGE SCENARIO

This section describes an example scenario, illustrated in Figure 1, and demonstrated in the video accompanying this paper. Jen is searching for a karaoke machine on eBay, a shopping website. She wants to use Joker to sort products by price within a page of search results, a feature not supported by eBay. Because Joker supports interleaving extraction and augmentation tasks on the fly, Jen is also able to extract the URLs of products to determine whether they have been visited and use that to further sort the products. While we have described only a single illustrative scenario in this example, Joker is flexible enough to support a wide range of other useful customizations and workflows on various websites, described in more detail in Section 4.

3 SYSTEM IMPLEMENTATION

In this section, we briefly describe Joker’s formula language and the *wrapper induction* [16] algorithm that Joker’s PBD interface uses to synthesize the row element and column selectors presented in formulas. Figure 2 illustrates the entire process.

3.1 Extraction Formulas

The Wildcard customization tool includes a formula language for augmentation, including operators for basic arithmetic and string manipulation, as well as more advanced operators that fetch data from web APIs. As in other tabular interfaces like SIEUFERD [7] and Airtable [5], formulas apply to a whole column at a time rather than a single cell, and can reference other columns by name. Joker extends this base language with new constructs which enable it to apply to *data extraction* instead of just augmentation. More information can be found in Appendix A.

By providing a single formula language to express extractions and augmentations, Joker enables a *unified interaction model* that supports interleaving the two. Furthermore, the formula language enables users to specify logic using pure, stateless functions that reactively update in response to upstream changes. This *functional reactive paradigm* is easier to reason about than traditional imperative programming, as demonstrated by the use of formulas by millions of end users in spreadsheet programs and end-user programming environments [1–3, 5, 6, 9].

Alt + click (option instead of alt on Mac) on a field you wish to scrape

Restart Cancel Done

Show More

Type

- Compact System without Monitor (308)
- Component System with Monitor (157)
- Component System without Monitor (87)
- DJ Karaoke System (89)
- Portable Karaoke System (274)
- Not Specified (1,000)

Free shipping
Free returns
131 sold

Sponsored
EARISE T26 Pro Karaoke Machine with 2 Wireless Microphones Remote Control
Brand New
\$79.99
Buy it Now
Free shipping
Save up to 10% when you buy more

2 =QuerySelector(rowElement, "span.s-item__price")

	f(x) A	f(x) B	C
1	Rockville 8" Pro Karaoke Machine/System 4 ipad/phone/	\$149.95	
2	EARISE T26 Pro Karaoke Machine with 2 Wireless Microp	\$79.99	
3	NYC Acoustics X-Tower Bluetooth Karaoke Machine Syst	\$69.95	
4	Rockville Dual 12" ipad/phone/Android/Laptop/TV Youtub	\$289.95	
5	Technical Pro WASP420 Bluetooth Karaoke Machine Syst	\$84.95	

A) Extracting Names and Prices by Demonstration:

- (1) The user scrapes the name and price of each listing by successively clicking on a name and price on the webpage.
- (2) The table reveals the underlying formula that was synthesized to scrape this data.

Alt + click (option instead of alt on Mac) on a field you wish to scrape

Restart Cancel Done

2,325 results for karaoke machine

Singing Machine Karaoke CD/Bluetooth/Auxiliary
Pre-Owned
\$15.00

3 =ExtractNumber(B)

4

	f(x) A	f(x) B	C	D
1	Singing Machine Karaoke CD/Bluetooth/Au	\$15.00		
2	Verkstar Ultra Portable Karaoke Machine, V	\$29.99		
3	Singing Machine Glow SML2200 Bluetooth	\$33.50		
4	Singing Machine SML2200: Karaoke Bluet	\$39.99		
5	New ListingThe Singing Machine Bluetooth	\$40.00		

B) Extracting the Number and Sorting by Price:

- (3) Autocomplete with function documentation helps the user write a formula that converts the price from a string with a dollar sign into a numerical value.
- (4) Now, the user can sort this column such that the rows are ordered by increasing numerical value. Correspondingly, the webpage is sorted such that the lowest-cost results appear at the top.

Alt + click (option instead of alt on Mac) on a field you wish to scrape

Restart Cancel Done

5

6 =GetAttribute(GetParent(A), "href")

7

8

	f(x) A	f(x) B	f(x) C	f(x) D	E
1	Rockville 8" Pro	\$149.95	149.95	https://www.eba	
2	EARISE T26 Pro	\$79.99	79.99	https://www.eba	
3	NYC Acoustics X	\$69.95	69.95	https://www.eba	
4	Rockville Dual 12	\$289.95	289.95	https://www.eba	
5	Technical Pro W	\$84.95	84.95	https://www.eba	

C) Extracting Links using Formulas:

- (5) The user wants to extract the link attached to the listing's name element, but the link cannot be extracted by demonstration. The user inspects the page to find that this is because the link is contained in the name element's parent.
- (6) The user composes a function to get the parent element of the name element (in column A) and extract its link attribute ("href").

Alt + click (option instead of alt on Mac) on a field you wish to scrape

Restart Cancel Done

2,325 results for karaoke machine

Sponsored
EARISE T26 Pro Karaoke Machine with 2 Wireless Microphones Remote Control
Brand New
\$79.99
Free shipping
Save up to 10% when you buy more

7 =Visited(D)

8

	f(x) A	f(x) B	f(x) C	f(x) D	f(x) E
1	EARISE T26 Pro Karaoke Machine	\$79.99		79.99	https://www.ebay.com/itm/313235
2	Verkstar Ultra Portable Karaoke V	\$29.99		29.99	https://www.ebay.com/itm/40282
3	Wireless Karaoke Machine Micro	\$179.00		179	https://www.ebay.com/itm/124621
4	Rockville Pro Dual 10" ipad/phone	\$334.95		334.95	https://www.ebay.com/itm/30390
5	New Listingsinging machine karat	\$75.00		75	https://www.ebay.com/itm/20344

D) Sorting to Filter by Unread Links:

- (7) The user writes a formula that determines whether a link has already been visited in the user's browser history.
- (8) The user can sort this column to see all listings with unread links at the top of the page, with already-read links at the bottom.

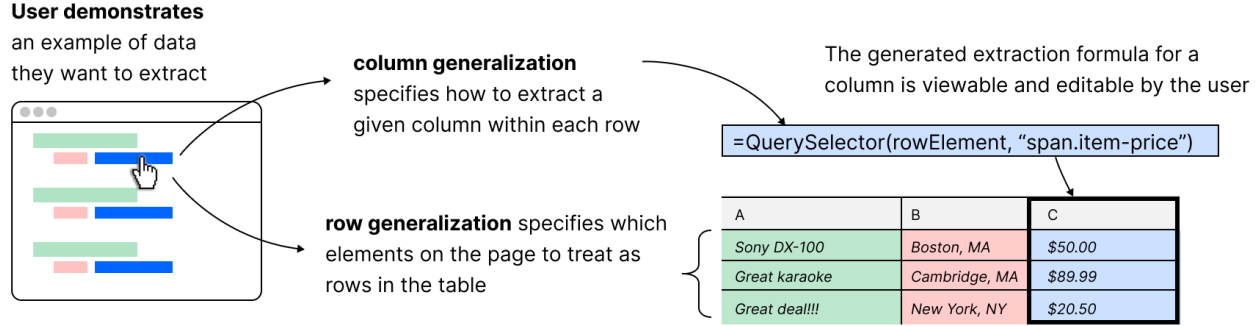


Figure 2: An overview of Joker’s interaction model and wrapper induction process.

3.2 Wrapper Induction

When users demonstrate a specific column value to extract, Joker must synthesize a program that reflects the user’s general intent. This is an instance of the *wrapper induction* problem of synthesizing a web data extraction query from examples. Prior work on this topic [13, 16] prioritizes accuracy and robustness to future changes, which makes sense for a fully automated system, but can lead to very complex queries. In our work, we chose to prioritize the readability of queries by less sophisticated users, so that users can more easily author queries and repair them when they break. We implemented a set of heuristics inspired by Vegemite [18] for wrapper induction, outlined in Appendix B.

4 EVALUATION

We evaluate our interaction model and tool in terms of two research questions:

RQ1: What kinds of websites can this model operate effectively on? We evaluate this with a suite of case studies that demonstrate its capabilities and limitations.

RQ2: How are users of different backgrounds able to use the system? We evaluate this with a small formative user study.

4.1 Case Studies

Our first evaluation describes the results of our use of Joker to extract data and perform customizations on popular websites.

4.1.1 Successful Applications. We have used Joker to achieve a variety of customizations across many websites.

Sorting search results by price on Amazon. We have found Joker to be useful for sorting the contents of various websites. One example of a useful sort achieved by Joker is sorting search results by price within the Featured page on Amazon. (Using Amazon’s sort by price feature often returns irrelevant results.) In Amazon’s source code, the price is

split into three HTML elements: the dollar sign, the dollar amount, and the cents amount. A user can extract by demonstration only the cents element into column A. Subsequently, because the parent element of the cents element contains all three of the price elements, the user can extract the full price using the formula `GetParent(A)`. Next, the user can write the formula `ExtractNumber(B)` to convert the string into a numeric value. Finally, the user can sort this column by low-to-high prices. In a similar manner, we have used Joker to extract and sort prices and ratings on the product listing pages of Target and eBay.

Filtering titles of publications on Google Scholar. We have also found Joker can be useful for filtering a website’s listings based on the text content of an element in the listing. For example, we have used Joker to filter the titles of a researcher’s publications on their Google Scholar profile which is not natively supported. First, a user can extract the titles into a column (A) by demonstration. Then, the user can write the formula `Includes(A, "compiler")` that returns whether or not the title contains the keyword “compiler.” Finally, the user can sort by this column to get all of the publications that fit their constraint at the top of the page. We have also used Joker to filter other text-based directory web pages such as Google search results and the MIT course catalog, in similar ways.

Retrieving information about links on Reddit. Additionally, we have used Joker to augment web pages with external information. For example, Joker can augment Reddit’s user interface, which has a list of headlines with links to articles, with the links’ read times and whether the link has already been read. To achieve this customization, a user first extracts the headline elements into column (A) by demonstration. The user can then extract the link into the next column (B) with the formula `GetAttribute(A, "href")`. Then, the user can write the formula `ReadTimeInSeconds(B)` that calls an API that returns the links’ read times. Similarly, the user

can write the formula `Visited(B)`, which uses another API that returns whether that link has been visited in the user’s browser history. The user can also extract elements such as the number of comments and the time of posting and sort by these values. We have performed similar customizations on websites such as ABC News and CNN.

4.1.2 Limitations. Joker is most effective on websites whose data is presented as a collection of similarly-structured HTML elements. Certain websites, however, have designs that make it difficult for Joker to extract data:

- *Heterogeneous row elements.* Some websites break their content into rows, but the rows do not have a consistent layout, and contain different types of child elements. For example, the page design of HackerNews alternates between rows containing a title and rows containing supplementary data (e.g. number of likes and the time of posting). Because Joker only chooses a single row selector, when extracting by demonstration, Joker will only select one of the types of rows, and elements in the other types of rows will not be extracted.
- *Infinite scroll.* Some websites have an “infinite scroll” feature that adds new entries to the page when a user scrolls to the bottom. Joker’s table will only contain elements that were rendered when the table was first created. Additionally, for websites that render a very large number of DOM elements, the speed of the live feedback provided by Joker’s PBD interface might significantly decrease. This is because the wrapper induction process used by the PBD interface queries the DOM which takes longer as the size of the DOM increases.

4.2 User Study

Our second evaluation reports a small, formative user study we conducted to understand how users would interact with Joker.

4.2.1 Participants. We recruited 5 participants with varying backgrounds. 3 participants were familiar with spreadsheet formulas. 3 participants had extensive web development experience, 1 had a small amount of prior web development experience, and 1 had no web development experience. 3 participants had previously extracted data from websites.

4.2.2 Protocol. The participants completed 7 web customization tasks across 2 websites. All participants attempted all the tasks.

First, we asked participants to customize a website with a relatively simple HTML structure: the MIT EECS course catalog website. All data *extraction* on this site can be performed with demonstrations alone in Joker, although augmentation

still requires writing formulas. The specific tasks were the following: 1a) Extract course titles, 1b) Extract course prerequisites, 1c) Add a column that indicates whether a course has a prerequisite & 1d) Add a column that indicates whether a course has no prerequisites and is offered in the fall term.

Next, we asked participants to customize a website with a more complex HTML structure: the search results page for the eBay shopping website. Due to the website’s complexity, demonstrations alone are not sufficient to extract data; users must also directly edit extraction formulas. The specific tasks were the following: 2a) Extract title from listings of Apple iPhones for sale, 2b) Extract the listing price for the phone, & 2c) Create a column that indicates whether a listing for a phone is sponsored.

Because some of the tasks build on results of previous tasks, we wanted to ensure all participants made enough progress to gather useful feedback. Therefore, whenever a participant got stuck for several minutes, we recorded why they were stuck and then offered hints on how to proceed (such as suggestions to read formula documentation or open the browser dev tools). While all participants were able to complete all tasks with hints, this obviously does not mean they could have completed the task unassisted. Our goal was not to simply measure whether users completed the task, but rather to gain qualitative insight into the barriers they faced.

4.2.3 Results. We have categorized our results into the following four groups:

Unified interaction model. Most participants took advantage of the unified interaction model to interleave extraction and augmentation tasks, rather than performing all extraction up front. For example, on task 1, most participants extracted the prerequisites by demonstration, added one or more columns to the table to perform some string operations on the prerequisites, and then continued on to extract more information from the web page by demonstration. Furthermore, we hypothesize that in a less controlled setting, users would be even more likely to interleave extraction and augmentation, since the task may be less well defined at the beginning.

One usability issue with the unified model was that participants sometimes got confused about how their demonstrations would affect the contents of the table. For example, multiple participants intended to add a new column by demonstrating an extraction, but instead accidentally overwrote the contents of an existing column. This poses a design challenge because the user’s demonstrations occur in the website, so they cannot directly interact with the table while demonstrating; this suggests that the interface needs to do a better job indicating where the results of a demonstration will be inserted.

Extracting simple data. On the relatively simple MIT course catalog website, all participants were able to extract the relevant data from the page within seconds, simply performing demonstrations with a few clicks. This suggests that when Joker’s generalization algorithm works well, it can be an effective tool for data extraction, even for users with limited programming experience. P1 said: “*you could hover and [the data] was already selected... that was very nice*”. P3, upon seeing the tutorial for extraction by demonstration, said “*that’s like black magic*.”

Extracting complex data. On the more complex eBay website where demonstration alone was not sufficient, results were more varied. P1, who had no prior web development experience, struggled to complete the task, saying that “*looking at HTML is a bit much*”; this suggests that more work could be done to make the experience usable for novices. However, users with more web development experience were able to use the tool to perform complex extractions, such as directly writing CSS selectors into the formula bar. P2 and P3 both reported that Joker’s live feedback loop was easier to use and faster than other approaches to web extraction; P3 noted that “*[with any other approach], it would have been slower to specify and slower to validate that I specified it correctly*.”

It was challenging for some participants to switch between using the browser’s developer tools and the Joker interface when doing complex extraction tasks. While we chose not to build HTML inspection into the Joker UI because the browser already provides a very rich set of tools, users sometimes were not able to tell how elements in the Joker table corresponded to elements in the browser’s element inspector.

Writing formulas. In general, participants were able to learn the formula language by using an autocomplete dropdown with inline documentation, which we developed as part of the Joker extension. In some cases, participants were able to immediately construct correct formulas on the first try; in other cases it took several attempts and some hints from the moderator to try a relevant function. While better documentation and error messages could help improve the learnability of the formula language, we also did not find it surprising that participants required some time to learn a completely unfamiliar formula language.

5 RELATED WORK

5.1 End-user Web Customization

Joker builds on web customization ideas implemented by previous tools. Our contribution is a new interaction model that allows for interleaving extraction and augmentation, enabled by a unified formula language and a PBD interface.

Joker is an extension of the Wildcard customization system [19], and preserves its foundational idea of synchronizing a table with a website. Wildcard only allows for extraction

logic to be written by programmers in Javascript; our work has substantially extended the Wildcard formula language and added an entire new system for dynamically creating data extraction logic within the user interface. We also improved the formula editing interface by adding an autocomplete dropdown and documentation popup, which proved important in our testing for allowing end-users to reliably edit and create formulas.

Vegemite [18] is a tool for end-user programming of web mashups. Like Joker, it allows users to perform demonstrations to extract data, but Vegemite only displays a table after all the demonstrations have been provided, which rules out interleaving extraction and augmentation. Vegemite does allow users to directly view and edit some of the logic generalized from demonstrations, but it only allows for editing augmentation logic, not extraction logic. The wrapper induction algorithm used in Joker is also very similar to Vegemite’s algorithm.

Sifter [15] is a tool that augments websites with advanced sorting and filtering functionality. It attempts to automatically detect items and fields on the website with a variety of heuristics. If these fail, it gives the user the option of demonstrating to correct some parts of the result. In contrast, Joker makes fewer assumptions about the structure of websites, by giving control to the user from the beginning of the process and displaying an editable synthesized program.

5.2 End-user Web Scraping and Program Synthesis

Joker builds on insights from other tools that synthesize web scraping (i.e. data extraction) code from user demonstrations, and give users ways to inspect and modify the generated code.

Rousillon [10] is a tool that enables end-users to extract hierarchical web data across multiple linked web pages. It presents the web extraction program generated from demonstrations in an editable, high-level, block-based language called Helena [3]. While both Rousillon and Joker create an editable program, they have different focuses. Because Rousillon allows users to extract data across multiple pages (e.g., extracting details from each linked page in a list), it uses an imperative language, with nested loops as a key construct. In contrast, Joker can only extract within a single page, and therefore can use a simpler declarative formula language. Also, Rousillon only allows editing high-level control flow and treats some details of the extraction logic as opaque; Joker offers finer-grained control over details like CSS selectors.

Mayer et al propose a user interaction model called *Program Navigation* [20] which aims to give users another mechanism beside examples for guiding the generalization process of PBE tools like FlashExtract [17] and FlashFill [14]. This is important because demonstrations are an ambiguous specification for program synthesis [21]: the set of synthesized programs for a demonstration can be very large. Joker shares the general idea of displaying synthesized programs, but only presents the top-ranked program. More broadly, Joker’s use of PBD to generate editable code embodies Ravi Chugh’s notion of *prodirect manipulation* [11], implemented in Sketch-N-Sketch [12], which aims to bridge the divide between programmatic and direct manipulation.

6 CONCLUSION AND FUTURE WORK

In this paper, we presented a unified interaction model for web customization. Our key idea is a spreadsheet formula language that encompasses both extraction and augmentation tasks, along with a programming-by-demonstration (PBD) interface that makes it easy for end-users to create formulas. The main area of future work involves testing with more users and making the formula language more accessible to end-users not familiar with CSS selectors. Our ultimate goal is to enable anyone that uses the web to customize websites in the course of their daily use in an intuitive and flexible way.

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APPENDIX

A: Extraction Formulas

Joker adds DOM elements as a data type to Wildcard’s formula language, alongside strings, numbers, and booleans. Because the language runs in a JavaScript interpreter, we simply use native JavaScript values to represent DOM elements in the language. DOM elements are displayed visually by showing their inner text contents. They can also be implicitly typecast to strings for use in other formulas; for example, a string manipulation formula like `Substring` can be called on a DOM element value, and will operate on its text contents.

We also added several functions to the formula language for traversing the DOM and performing extractions, summarized below with their types:

- `QuerySelector(el: Element, sel: string): Element`. Executes the CSS selector `sel` inside of element `el`, and returns the first matching element.
- `GetAttribute(el: Element, attribute: string): string`. Returns the value for an attribute on an element.
- `GetParent(el: Element): Element`. Returns the parent of a given element.

To extract data from a row, formulas need a way to reference the current row, so we added a construct to support this use case. Every row in the table maps to one DOM element in the page; we allow formulas to access this DOM element via a special keyword, `rowElement`. In some sense, `rowElement` can be seen as a hidden extra column of data in the table containing DOM elements.

While many more functions could be added to expose more of the underlying DOM API, we found that in practice these three functions provided ample power through composition. For example, in Section 2 we showed how `GetParent` and `GetAttribute` can be composed to traverse the DOM and extract the URL associated with a product listing.

B: Wrapper Induction Algorithm

Joker’s wrapper induction algorithm implements a set of heuristics inspired by Vegemite as described below:

Determining Row Elements. The user starts by demonstrating an element v , representing a value that should be in the table. From that demonstration, we must find a set of *row elements* that represent the rows of the table. We could naively assume that $\text{parent}(v)$ is the row containing v , but often v is deeply nested inside its containing row; we must determine which ancestor of v is likely to be the row.

Intuitively, we solve this problem by assuming that all rows share some similar internal structure. In particular, we expect most rows to contain a value for the demonstrated

column. (If there were no missing data, we’d expect *all* rows to contain data for this column.)

Formally: assume a function $\text{select}(el, s)$ which runs a CSS selector that returns the set of elements matching s within el . We generate a set of plausible candidates P , consisting of pairs of a row element and a CSS selector:

$$P = \{(r, s) \mid r \in \text{ancestors}(v) \wedge \text{select}(r, s) = \{v\}\}$$

For each candidate $(r, s) \in P$, we compute a weight function w , which is based on the number of siblings of r that have “similar structure”, defined by checking whether running s within the sibling also returns a unique element.

$$w(r, s) = |\{r' \mid r' \in \text{siblings}(r) \wedge |\text{select}(r', s)| = 1\}|$$

We then choose the candidate with the highest weight. In case of ties, the candidate closer to v in the tree (i.e., lower in the tree) wins. Given a winning candidate (r, s) , the full set of row elements is $\{r\} \cup \text{siblings}(r)$.

Synthesizing CSS Selectors For Column Values. Once we have determined the row elements, next we must choose a CSS selector that will be used to identify the demonstrated value within its row.

Given a demonstrated value v within a row element r , we generate two kinds of plausible selectors:

- selectors using CSS classes, which are manual annotations on DOM elements added by the website’s programmers, typically for styling purposes (e.g. “item_price”)
- selectors using positional indexes within the tree, using the *nth-child* CSS selector (e.g. *nth-child*(2), representing the second child of an element)

The minimum criteria for a plausible selector s is that it uniquely identifies the value within the row: $\text{select}(r, s) = \{v\}$. But there may be many plausible selectors, so we must pick a best one.

We first prioritize selectors using classes, because they tend to be more robust to changes on the website. A single selector can combine multiple classes, but we prefer using fewer classes when possible. If no plausible class-based selector can be generated (for example, if the relevant elements don’t have any classes to query), we fall back to using a positional index selector. This kind of selector can always be generated regardless of the contents of the page, but tends to be less accurate and robust.