

# Wrangell: A Data-Wrangling DSL in Haskell

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# ABSTRACT

Managing data from various document types (e.g. txt, csv files) often requires manual work by a user. We present Wrangell, a fast, simple domain-specific language for common data manipulation tasks.

Run from the command line, Wrangell is specifically designed to modify tables of data wherein columns are separated by a user-specified delimiter. Wrangell has functions to read in a file, parse the data, and run a transformation on the table, and write the modified table to an output file.

We implemented Wrangell in Haskell, a functional programming language. We chose Haskell for its powerful type system which allows a large amount of control over how operations behave when handed various data types. Furthermore, as a functional programming language, Haskell is a good choice for tasks which depend on running transformations on data.

We based Wrangell on Scheme, a Lisp-like language known for its simplicity. Wrangell’s simplicity is desirable as it would facilitate ease of integration into potential future data management projects.

## 1. INTRODUCTION

For our project, we wanted to design and implement a small language. Haskell was an attractive tool because of its type system which makes data types immutable and allows for algebraic definition of types. Furthermore, our team members had either a beginner or intermediate level of Haskell knowledge, so we saw this as an opportunity to learn something new.

To keep the scope of our language’s features manageable, we chose to gear our language to a specific domain. Because Haskell is a functional language, we thought that *data-wrangling*, with its emphasis on data transformations, would be a good application for our language implemented in Haskell. We began creating Wrangell, a DSL for “WRANGLing” data via “haskELL”.

We focused our desired data-wrangling tasks on one format: a table of data in a csv file, wherein we recognize rows via lines of the file, and columns per a user-specified delimiter. The following operations would initially be included in Wrangell:

1. Read in an input filename, output filename, () , ()
2. Read in and parse a csv file per a filename
3. Run some transformation on the data in the file

4. Write the transformed data to an output file

More specifically, the transformation operations in Wrangell would allow users to:

1. Remove a column
2. Insert a column
3. Remove a row
4. Insert a row

A motivating example for Wrangell could be the following: A small business has many csv files (say, one for each year the company has been active) with customer data. Each file has a column for various items including the customer's name, address, phone number, and credit card number. Now say the business wanted to contract some advertising company to mail a promotional catalog to each of its customers. In this case, the business would want to share its customer database with the advertising company. However, sharing customer credit card information with the advertising company would be both unnecessary and dangerous. In this case, the business would want an easy way to make copies of the customer information files with the credit card information column removed.

## 2. APPROACH

Because our team members only had beginner or intermediate levels of Haskell knowledge, we chose to keep Wrangell’s data management operations fairly simple. As a result, Wrangell’s application subjects are limited to tables of data wherein columns were separated by either a comma, or a user-specified delimiter (example, figure () )

We determined big-step semantics for Wrangell’s operations in figure (placeholder).

Finally we planned to apply Wrangell’s data manipulation features to a variety of csv files. Examples of our desired behavior for Wrangell are illustrated in figure (placeholder).

In this example, we might want to share the files peopleA.txt, peopleB.txt, and peopleC.txt with some third party. But first we need to remove the sensitive information (e.g. "SocSecNum" ). We could manually go through all of the

filename: peopleA.txt

```
name,age,gender,hometown,Favorite Food,SocSecNum
kenan,40,M,Cupertino,Crepes,11112222333
dana,25,F,Santa Barbara,Cheetos,11133661
donald,17,M,Davis,Milk,888176162
```

filename: peopleB.txt

```
name,age,gender,Favorite Food,school,SocSecNum
Dale,40,M,Coffee,Dartmouth 8716268282
Audrey,18,F,Coffee,Twin Peaks High,82377266378
Laura,17,F,Cereal,Twin Peaks High,828138921
Bobby,17,M,Bacon,Twin Peaks High,898709809
Hawk,34,M,Doughnuts,Brandeis,55154226
Ben,48,M,Brie sandwich,USC,556363
Hank,34,M,Dominoes,Prison GED Program,8872211
Leo,32,M,Raw Hamburger,Sarah Lawrence,837428937
```

filename: peopleC.txt

```
name,age,gender,SocSecNum,hometown,Spouse
Jax,33,M,8376499281,Charming,Tara
LuAnn,18,F,98349823,Twin Peaks,Otto
Tig,45,M,0002928181,Phoenix,n/a
Bobby,51,M,73727627618,Austin,n/a
Clay,54,M,67257632617,Charming,Gemma
Juice,31,M,8794861321,Philly,n/a
Tara,3,M,6534654215,Charming,Jax
```

files, manually removing the social security number column from each. But if we had, say, 100 files, this would be very expensive. By using Wrangell we would hope to create some modified versions of the files ("mod\_peopleA.txt", "mod\_peopleB.txt", and "mod\_peopleC.txt") with easily copy-pasted commands in the command line.

Wrangell expects data to be formatted per the following example (by default, Wrangell expects a comma as the delimiter between columns:

1. Table

$$\Sigma = \bigcup_{n=1, r=1}^{\infty} \{M | t_i \in T, m_{*i} \in V(t_i : r), l_i \in \{Strings\}, \forall l_i \# l_j s.t. l_i = l_j \wedge j \neq i\}$$

$$\sigma[M] = (t_1, t_2, \dots, t_n), (m_{*1}, m_{*2}, \dots, m_{*n}), (l_1, l_2, \dots, l_n)$$

2. Drop Column

$$\frac{< name, \sigma > \Downarrow l_i}{< dropCol name, \sigma > \Downarrow \sigma'}$$

$$\sigma'[M] = \{(t_1, t_2, \dots, t_{i-1}, t_{i+1}, \dots, t_n), (m_{*1}, m_{*2}, \dots, m_{*i-1}, m_{*i+1}, \dots, m_{*n}), (l_1, l_2, \dots, l_{i-1}, l_{i+1}, \dots, l_n)\}$$

3. Drop Row

$$\frac{m'_{*i} = \{m_{ji} | < \lambda m_{ji}, \sigma > \Downarrow False\}}{< dropRow \lambda, \sigma > \Downarrow \sigma'}$$

$$\sigma'[M] = \{(t_1, t_2, \dots, t_n), (m'_{*1}, m'_{*2}, \dots, m'_{*n}), (l_1, l_2, \dots, l_n)\}$$

4. Insert Column

$$\frac{j \in [1, n+1], name \in \{l_1, l_2, \dots, l_n\}}{< insertCol type name j, \sigma > \Downarrow \sigma'}$$

$$\sigma'[M] = \{(t_1, t_2, \dots, t_{j-1}, type, \dots, t_n), (m_{*1}, m_{*2}, \dots, 0(), \dots, m_{*n}), (l_1, l_2, \dots, name, \dots, l_n)\}$$

Big step semantics for our functions in Wrangell:

### 3. IMPLEMENTATION AND RESULTS

We ran into several challenges when writing our Wrangell code in Haskell. Most of our challenges centered on our need to mimic imperative behavior in Haskell, a functional language.

#### 3.1 Modifying Information in an Immutable State

One challenge was to store information about the contents of the input file. This conflicts with the idea of a functional language's "immutable states." For example, Wrangell would need to store the parsed data table from each input file. Additionally, Wrangell needs to modify the data table. Our solution was to perform operations within the *IO Monad*, an abstract data type which provides a system wherein we can build "composite actions." The below example demonstrates our use of the IO monad during construction of an empty data table for later use:

```
import Data.IORef
data Table' =
  Table' { rows :: [[WVal]],
          format :: [WType],
          labels :: [String],
          delimiter :: String}
type Table = IORef Table' -- TODO: This will be a bit different

--creates a new empty table context
emptyTable :: IO Table
emptyTable = new IORef Table' {
  rows = [[]],
  format = [],
  labels = [],
  delimiter = ",",
}
```

Because emptyTable is in the IO monad, we can later () as follows:

```
--blah blah blah
```

Without defining emptyTable as an IORef, we would have (encountered an error?) above.

Additionally, Use of the IO Monad via IORef also allows Wrangell to be used to make "helper functions",

We describe monads and IO further in "Related Work" below.

#### 3.2 Type System Flexibility

The second challenge was creating a type system which was flexible enough to achieve our desired level of expressiveness while still behaving within the confines of Haskell's type system. This issue is illustrated in funcTable below:

"Our solution was type-safe exception handling" this is handled in the Either monad.

#### 3.3 Getting Wrangell to read in files

#### 3.4 Data.csv

We used data.csv (<https://github.com/jgoerzen/missingh/blob/master/data.csv>) to read in and parse the data files.

Monads

Monad transformers

Type-safe error checking

We created our own parser in Haskell... One of our biggest challenges was to improve our level of Haskell coding expertise...

Getting Wrangell to read in files:

What Wrangell can do:

Lenses:

### 4. RELATED WORK

Application to other file formats (hdf5)

Our research centered on learning Haskell and finding ways to generate desired behaviors.

The most challenging concept for us was Monad Transformations.

We researched current data-wrangling tools, and found (). Tool1() Tool2() ...The benefits of Wrangell over these tools is (speed?) .

Data wrangling likes functional programming because functional programming is all about input transformation and output.

### 5. CONCLUSIONS

Wrangell

Hypothetical future improvements of Wrangell might include improvement in label analysis, a feature which would allow users to identify columns purely via their label. For example, an operation meant to aggregate email addresses for all customers since 1995 would need to handle customer data files wherein the column "email address" is not present (e.g. files prior to 2002). Additionally, it would also be useful if Wrangell could handle files with inconsistent ordering of columns. Finally, we would want Wangell to be able to handle data tables with an unknown number of columns.

### 6. REFERENCES