Dataproofer

Spell check for data

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Spell check identifies potential spelling errors.

Dataproofer identifies potential data errors.

Primary Use Case

Quick check of data quality immediately following data collection

Strengths

- Fast, automated quality checks
- Cross-platform (Window, OS X, Linux)
- Supports XLSX, XLS, TSV, CSV, PSV, and Google Spreadsheets
- Customizable checks

Weaknesses

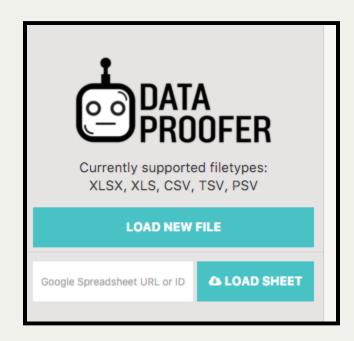
- Creating custom checks requires some programming experience
- Not for editing
- Does not like large data (> 500 KB)
- Does not support statistical software data formats

Using Dataproofer

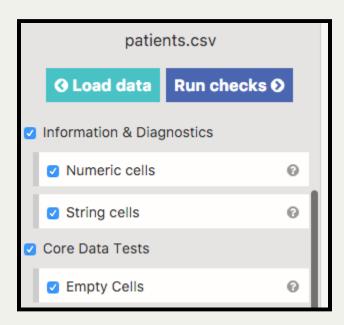
Install Dataproofer

dataproofer.org

Load Raw Data



Run Default Checks



Review Results

patients.csv		001	М	11/11/1998	88140	80	10	OVERVIEW
	1	002	F	11/13/1998	84120	78	X0	
9 Select checks	2	003	X	10/21/1998	68190	100	31	
_	3	004	F	01/01/1999	101200	120	5A	
9 passed out of 15 total	4	XX5	M	05/07/1998	68120	80	10	
No missing or duplicate column header	s 5	006		06/15/1999	72102	68	61	
	6	007	M	08/32/1998	88148	102	0	
Numeric cells	9 7	008	F	08/08/1998		210	70	
	8	009	M	09/25/1999	86240	180	41	
1 String cells	9	010	F	10/19/1999	40120	10		
	10	011	M	13/13/1998	68300	20	41	
Empty Cells	9 11	012	M	10/12/98	60122	74	0	
	12	013	2	08/23/1999	74108	64	1	
O Duplicate Rows	13	014	M	02/02/1999	22130	90	1	
• Determination or many	0 14	002	F	11/13/1998	84120	78	X0	
Potentially missing rows	15	003	M	11/12/1999	58112	74	0	
Words at their character limit	₀ 16	015	F		82148	88	31	
Words at their character innit	17	017	F	04/05/1999	208	84	20	
	o 18	019	M	06/07/1999	58118	70	0	
	19	123	M	15/12/1999		60	10	
Summed integer at its upper ⊘ limit	20	321	F		900400	200	51	
	21	020	F	99/99/9999	10 20	8	0	
Small integer at its SQL upper ⊘ limit	22	022	M	10/10/1999	48114	82	21	
	23	023	F	12/31/1998	22 34	78	0	
	24	024	F	11/09/1998	76 120	80	10	
Big integer at its SQL upper	₀ 25	025	M	01/01/1999	74102	68	51	
⊘ limit	26	027	F	NOTAVAIL N	166	106	70	
	27	028	F	03/28/1998	66150	90	30	
Outliers from the mean	0							

Demo

patients.csv

Custom Checks

Motivation

Working with survey data, we want to quickly identify the following missing values:

- NA
- N/A
- -99
- -98

Before the Custom Check

field.csv

Anatomy of a Custom Check

```
customDataprooferTest.name("Missing (NA)")
   .description("If a cell contains an NA value.")
   .conclusion("Check for any patterns in missing values.")
   .methodology(cellMethod(isMissingCheck))
```

isMissingCheck Function

```
var isMissingCheck = function(cell) {
    if (cell === 'NA' || cell === 'N/A' ||
        cell == -99 || cell == -98) {
        return true;
    }
    return false;
}
```

After Custom Check

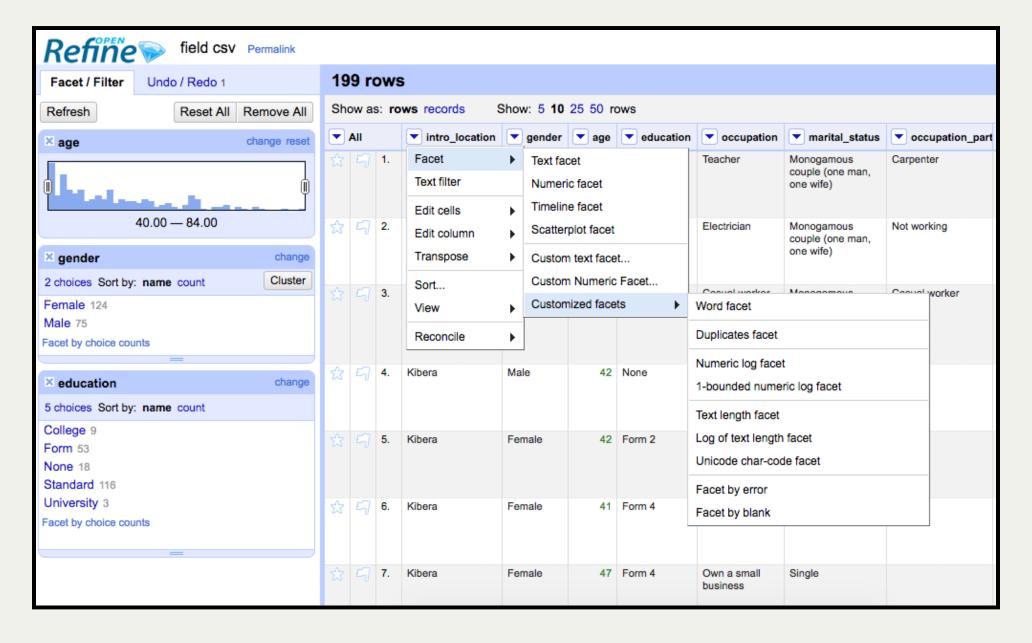
fieldv2.csv			fo	name	contact	contact2	date	location	consent		OVERVIEW	
	1	350	8.92540293		fe05932c-14	Mery Njoki N	Simon Maige	704113535	-99			1
⊙ Select checks	2	350	8.92540293		f42d6a5f-46	Mery Njoki N	Muraya Kima	799999999	-99			
	3	350	8.92540293		b6613436-5	Mery Njoki N	Bernard mbu	707026796	-99	-		
1 passed out of 2 total	4	350	8.92540293		f6fbc6c3-e3	Mery Njoki N	Olive Mwihal	728738426	792365406	100		
No missing or duplicate column headers	5	350	8.92540293		5e8d7ca6-9	Mery Njoki N	Njoroge nga	79999999	-99	1		1 4
	6	056	8.925402100		7cdf2552-39	Maureen Nje	Evelyne Ndu	712665330	-99			
Missing (NA)	7	056	8.925402100		acf3caab-e8	Maureen Nje	Francis Kima	799999999	-99	1.5		
1	8	056	8.925402100		1931e5ac-3a	Maureen Nje	Henry Karan	720093842	-99	10.0		1
	9	056	8.925402100		1bc50b60-f7	Maureen Nje	Moffat Ngati	708066916	-99			
	10				6a19cf7b-67	Moffat Ndur	Lilly Waringa	723310963	79999999	1		11
	11	056	8.925402100		Ofbd189e-c7	Maureen Nje	Ruth Wanjiru	718865716	-99			
	12				5069e1e7-2	Moffat Ndur	Eliud Mwaura	711158253	725522213			
	13				080d55c1-4	Moffat Ndur	Lucy Muthor	714170921	712740352			
	14				703b2353-e	Moffat Ndur	James Waith	725993888	720475915			- 11
	15				ba26a975-d	Mercy Mum	David ndichu	721987416	-99			
	16				554f9b7c-6	Mercy Mum	Agnes wanjir	714368895	-99			
	17				7612ef4c-d4	Mercy Mum	Samuel wain	99999999	-99			
	18				51b282a2-4	Catherine M	I Isabel mwiha	736598202	-99			
	19				29cd7d67-1	Dickson Mal	Ann Wambui	710366151	-99			II.
	20				ba1ac56a-1d	Dickson Mal	Eunice Ann N	726788961	787570073			
	21				69d6f86e-8	Dickson Mal	Gladys Waru	727654170	N/A	h 🕒 i		1 1
	22				3c488e02-e	Jackline Kie	r James mwau	708663928	-98	_		
	23				3fdff8dc-ce	Jackline Kie	r George gikor	721995377	731334166			
	24						Peter Kamau		-99			
	25						r Stephen gita		-99			
	26						Margaret Njo		7268182322			
	27				f747e864-9	Githu Anne	James kimar	701828752	720067721			

Alternative tools

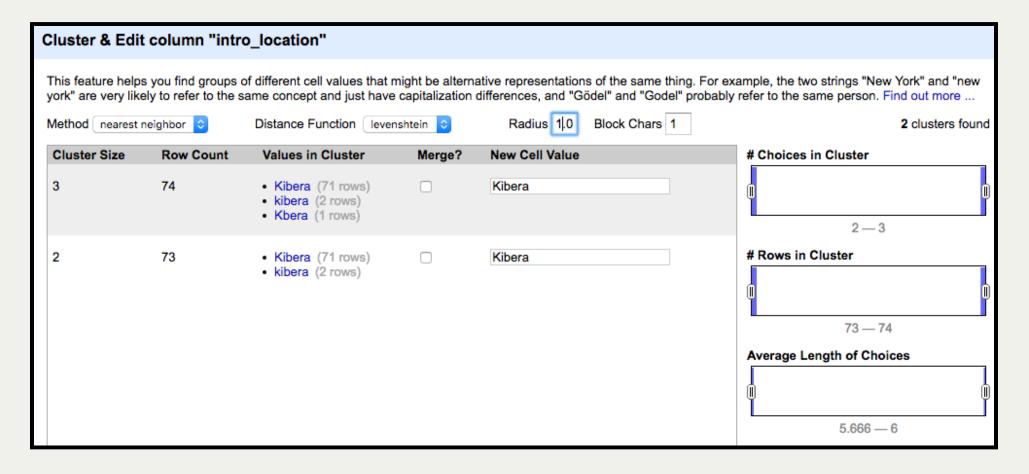
OpenRefine openrefine.org

Formerly Google Refine

OpenRefine Facets



OpenRefine Clusters



OpenRefine: Strengths

- Open source software
- Extensible custom checks/filters can be written in Clojure, Python/Jython, or Refine's own expression language

OpenRefine: Weaknesses

- Can't handle large files (> 500 KB)
- Feels slow
- No easy way to track changes (not very reproducible)

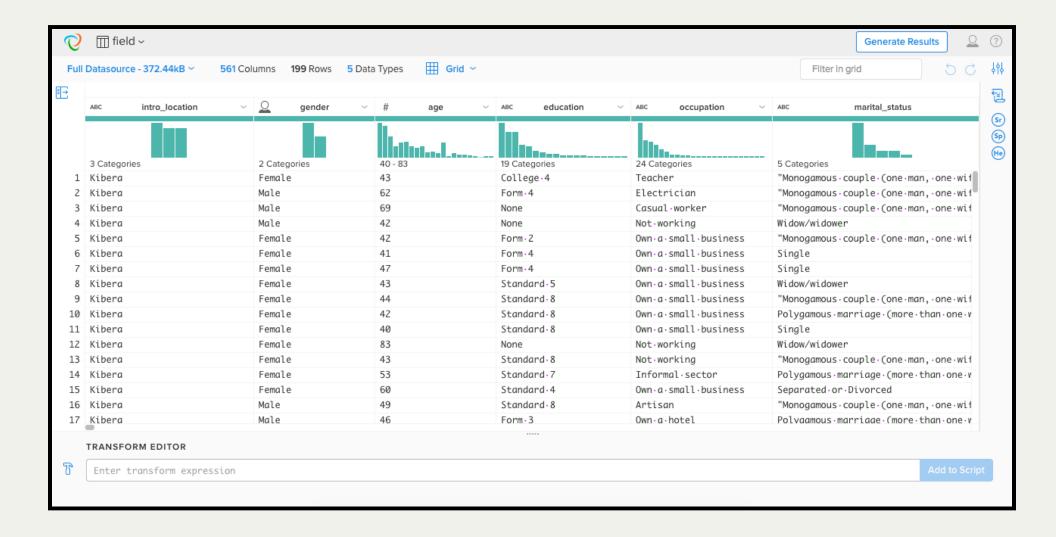
OpenRefine Demo

Trifacta Wrangler

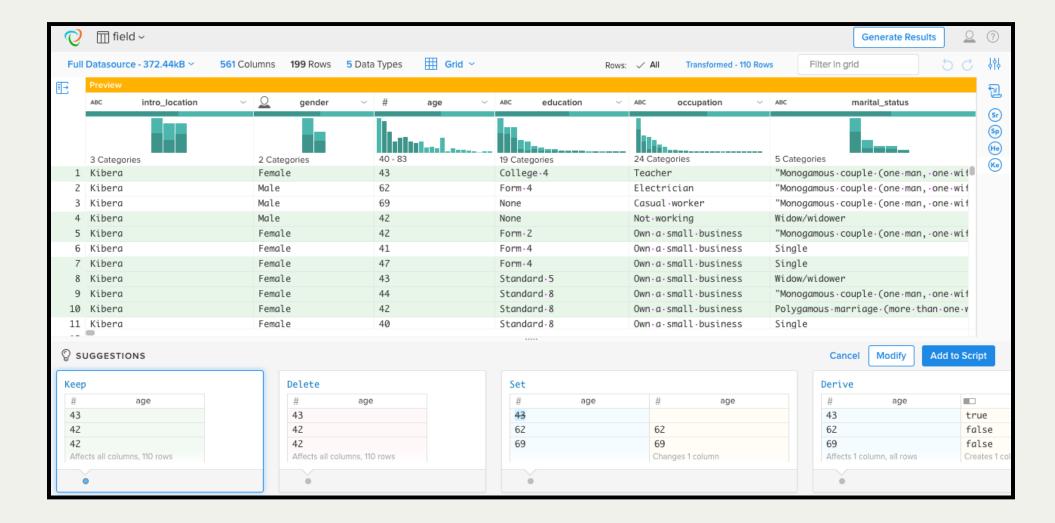
trifacta.com

Formerly Stanford Data Wrangler

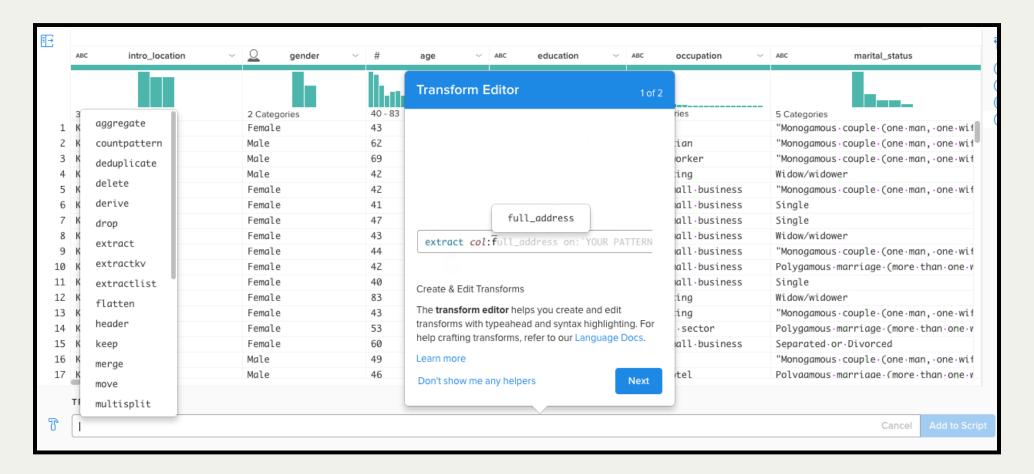
Trifacta: Overview



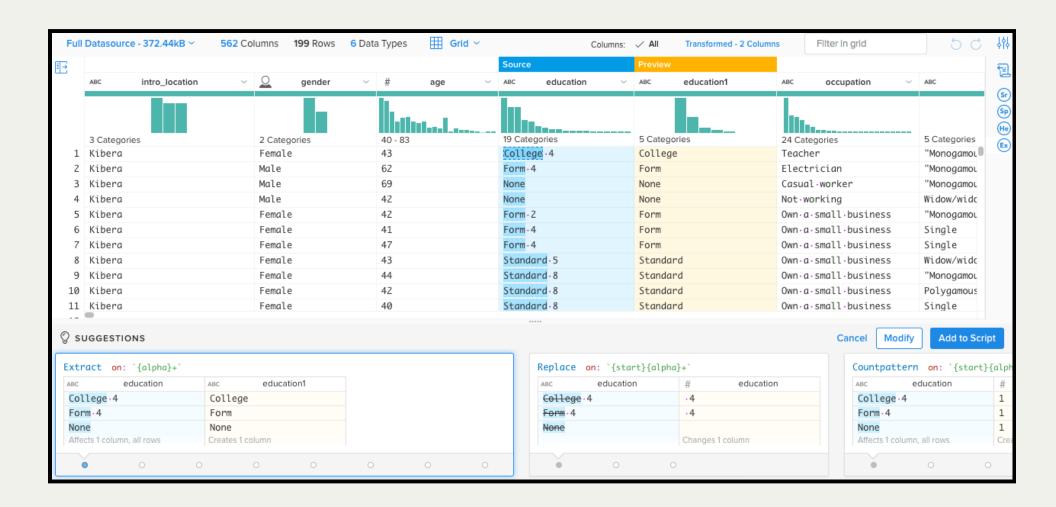
Trifacta: Filter



Trifacta: Scripts



Trifacta: Scripting Without Code



Trifacta: Strengths

- Interactive histograms
- Script builder
- Reproducible data cleaning
- Takes a 500 KB sample from large datasets

Trifacta: Weaknesses

- Requires an internet connection (may have to setup proxy settings)
- May not be appropriate for confidential data

Trifacta Demo

Conclusion

Use Dataproofer for

- Data quality checks for data entry
- Nice high-level overview of potential data errors
- First check after downloading or creating a new dataset

If you need more, consider OpenRefine and Trifacta Wrangler.