Displaying Health Data

Cases, Techniques, Solutions

Colloquium + Live-Webcast + Recording Medical Sciences Building (MBS) 160 University of Victoria

November 28 - 30, 1 - 3 pm PST















Displaying Health Data

Cases, Techniques, Solutions

Health Data

DAY 1 2018-11-28Wednesday

- 13:00 Transactional data of Island Health: How patients vote with their feet Dr. Ken Moselle (Island Health) and Dr. Andriy Koval (BC Observatory, UCF)
- 14:00 Visualizing logistic regression with the "coloring book" technique:
 A study in ggplot2

 Dr. Andriy Koval (BC Observatory for Population and Public Health, UCF)

Substance Use

DAY 2 2018-11-29 Thursday

- Nuances of information sharing and data display in a mobile application for students with substance use disorder

 Dr. Barbara (Basia) Andraka-Christou (University of Central Florida)
- Optimizing public health surveillance through reproducible reporting: Response to opioid crisis on Vancouver Island Shannon Tracey (University of Victoria) and Maritia Gully (Island Health)

Pipelines & Dashboards

DAY 3 2018-11-30 Friday Building pipelines and dashboards for practitioners: Mobilizing knowledge with reproducible reporting

Dr. Will Beasley (University of Oklahoma Health Sciences Center)

4:00 Constructing workflows for reproducible analytics: Suppressing small counts for provincial chronic disease dashboard Dr. Andriy Koval (BC Observatory, UCF) and Anthony Leamon (Island Health)















Building pipelines and dashboards for practitioners

Knowledge mobilization with reproducible reporting



Will Beasley



Please email questions to aging@uvic.ca

Access lecture slides from

github.com/dss-ialh/displaying-health-data

Clone the repo to reproduce examples

Building Pipelines and Dashboards for Practitioners:

Mobilizing knowledge with reproducible reporting

Will Beasley, Geneva Marshall, & David Bard University of Oklahoma Health Science Center Biomedical and Behavioral Health Core

Common Roles for the BBMC

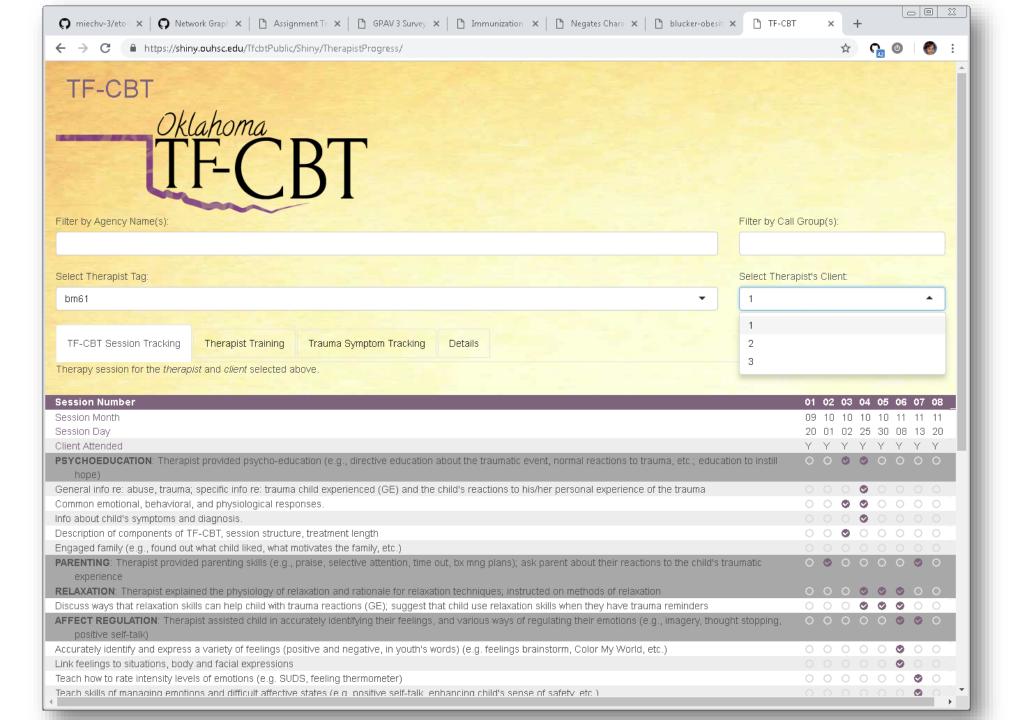
domain expertise in a few areas

incorporation of data sources from multiple agencies

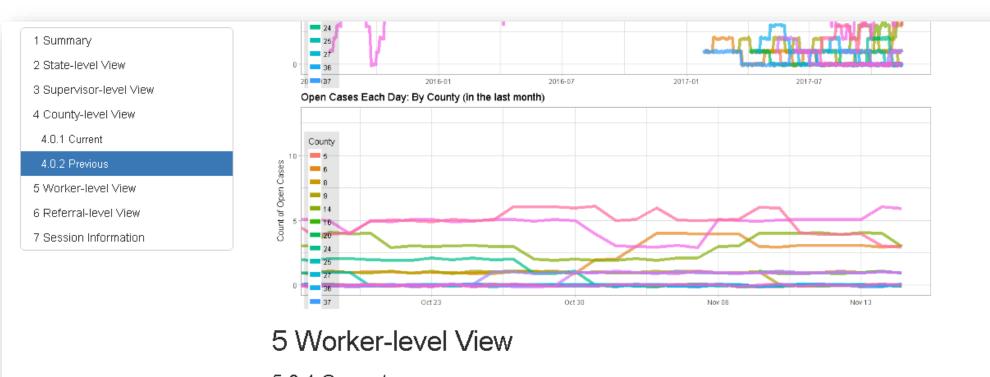
statistical analysis for academic publications

construct information displays for nontechnical audiences

Example 1: history of therapist training



Example 2: availability of specialized CPS workers



5.0.1 Current

Each row represents one ISS worker. The 'capacity by region' field squishes together the capacities for the five regions, separated by commas.

iss worker	region(s)	county(ies)		active worker	•	capacity max	assignments open [total]	•	has capacity
95 (Blaine/Canadian/Custer)	1;	6;9;20	-	Yes	1,5555	1	1 [43]	1658	no
488 (Canadian)	1;	9	-	Yes	0,-,-,-,-	0	0 [2]	-	no
749	1;	6;9;24;27;36;37;42;52;60	-	Yes	3,-,-,-	3	3 [15]	1622; 1625;	no

aanaaitu

Example 3: data collection quality

3 Smells

A smell test won't validate a given record (like the rules above), but it will make sure the dataset overall smells correct.

3.1 Smell Summary

Smell Name	Priority	Proportion of	Legal Boundaries	Value	Pass
female majority	1	female respondents.	[0.80, 1.00]	0.0000	FALSE
age proportion infant	2	index children under 2 months old.	[0.05, 1.00]	0.1660	TRUE
age proportion toddler	2	index children between 2 & 12 months old.	[0.20, 1.00]	0.2510	TRUE
proportion black	2	Around 4th of the survey population is black	[0.00, 0.30]	0.3166	FALSE
proportion children at home	2	Most participants have children living at home most of the time.	[0.50, 1.00]	0.8803	TRUE
proportion currently pregnant	2	More than 2/3rd of the participants are not currently pregnant.	[0.80, 1.00]	0.8672	TRUE
proportion developmental disabilities	2	Less than 10% of the children have developmental disabilities.	[0.00, 0.10]	0.0714	TRUE
proportion food stamps services	2	A little less than half the population is on food stamps.	[0.35, 0.50]	0.4981	TRUE
proportion hispanic	2	A little less than half the survey population is Hispanic.	[0.30, 0.40]	0.2857	FALSE
proportion household children	2	Around half of our survey population has households with one child.	[0.40, 0.50]	0.4942	TRUE

GPAV 3 Survey Validation

Date: 2018-11-06 10:46:50

This report identifies violations in the GPAV 3 Community Survey dataset, aka (Giving Parents A Voice -MIECHV 3).

1 Set Up

1.1 Instructions

1.2 Execute Rules

1.3 Execute Smells 1.4 Inactive Rules

- 1. Work through each line in the table to discover & correct entry mistakes in the database. Be aware that a single mistake may manifest in multiple rows; fixing one value may clear several rows.
- 2. Click the record ad value in a row to be taken to the participant's Event Grid in REDCap (i.e., the 'stop light page').
- Double-check that your manual corrections in REDCap are reflected in data-public/derived/gpaw/gpaw-3-violation-rule.csv. Be careful not to move this file to somewhere unsafe.
- 4. To help discover the dataflow and problematic location, use the
 - project's codebook in REDCap,
 - gpav-3-arch file, and
 - validation-gpav-3 report.
- 5. If you create a new rule or smell check, make sure the error message won't reveal any PHI.

2 Rules

A rule is very exact. Each record is examined, and determined if it passes each specific rule.

2.1 Rule Detail 2.2 Rule Summary

Violations at 2018-11-06 10:47:04

	check name	record id	♦ data collector ♦	error message 🛊	priority \$	instrument \$	interview date
			-	All		Al	F
1	parent pro services	255-24	8	ParentPRO home-based services status is missing. (The question is 'Are you currently participating in parentPRO home-based services?')	1	participant demographics	2018-09-05
2	parent pro services	255-63	8	ParentPRO home-based services status is missing. (The question is 'Are you currently participating in parentPRO home-based services?')	1	participant demographics	2018-10-09
3	Parentpro service	255-24	8	ParentPRO service status is missing when recruit source is present. If recruit source	1	participant	2018-09-05

Example 4: MCMC HPDs are updated as collection progresses

1. Histogram Overlay

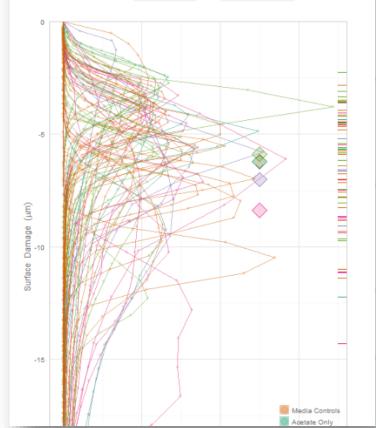
The **first graph** represents the probe heights, as a distance from the coupon's value is the depth of the probe, while the *x* indicates how much of the coupon *treatement*'s mean depth. The ticks on the right side indicate a *coupon*'s mean

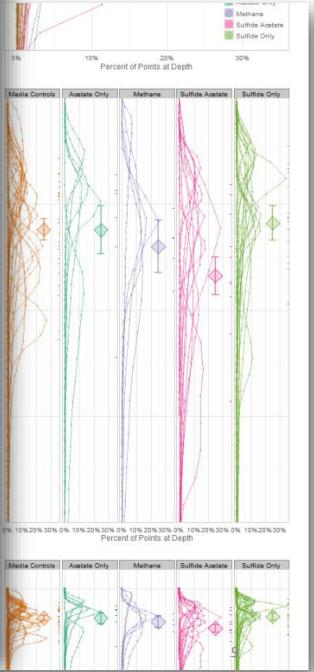
The **second graph** is almost identical to the first, but with two differences. First standard errors are shown around each treatment mean; the means and errors shown below. They bands mark the 16% and 84% quantiles of the posterior dithe 68% coverage of a +/-1 parametric SE band.

The **third graph** is identical, yet loosens the *y*-axis range so that the full depth maximum pit depth is a variable worth including in a formal analysis.

The five outlier coupons are *excluded* from these two graphs (*ie*, the four procususpicious control coupon).

The diamonds/means for AcetateOnly and MediaControls are on top of eac



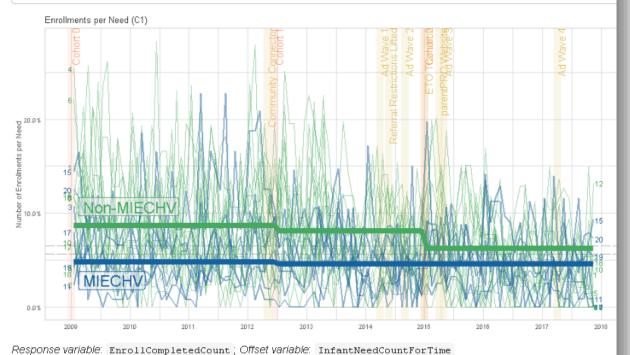


Example 5: semi-annual reports with multilevel longitudinal models

completed enrollments (not significant for the drop between 0 and 1, but significant for the drop between 0 and 2. Coefficients 1d and 1e are greater than 1.00, but not significant; a significant coefficient would have indicated that the six MIECHV regions are comparatively outgaining the fourteen comparison regions between cohorts 0 and 2. However, $1b \times 1d$ produces a value under 1.00 (and $1c \times 1e$ do too), indicating MIECHV regions observed a decline in overall enrollments (but a much less steep decline than comparison regions). This pattern is evident in Figure zzz, notice how the green line descends sharply during the nine year period, while the blue line is essentially flat. The groups' enrollment rates are almost equal by 2015.

The esimates from Models 2 and 3 corresponded closely to Model 1, suggesting that the three operationalizations of the MIECHV treatment are comparable for completed enrollments. The only notable difference involves coefficient g. The Model 2 and 3 estimates are about 5%, at which point coefficients become significant; this suggests MIECHV regions did relatively better than non-MIECHV regions between cohorts 1 and 2.

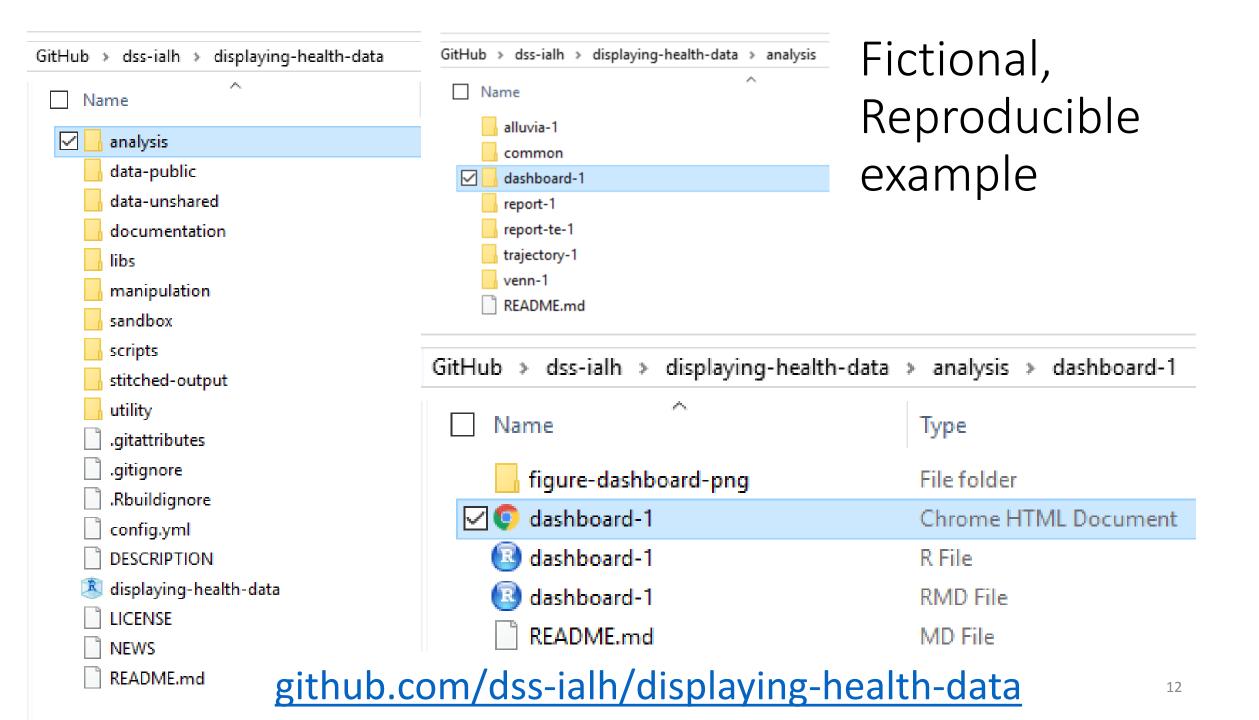
	MIEC	HV Conti	acted	M	ECHV St	affed	I	Has Con	nector
		Change	р		Change	p		Change	р
0 - NvM	: 1a	0.57	<.001	2a	0.62	<.001	3a	0.68	<.001
0v1 - N	: 1b	0.93	.235	2b	0.94	.181	3b	0.97	.533
0v2 - N	: 1c	0.71	<.001	2c	0.76	<.001	3c	0.76	<.001
0v1 - NvM	: 1d	1.01	.925	2d	1.00	.978	3 d	0.89	.088
0v2 - NvM	: 1e	1.29	.103	2e	1.17	.003	3e	1.18	.002
1v2 - N	: 1f	1.05	.287	2 f	0.99	.744	3 f	0.96	.276
1v2 - NvM	: 1g	1.23	<.001	2g	1.26	<.001	3 g	1.31	<.001



dashboards vs reports

Blur the distinction

- Try to make your reports more like dashboards, so they
 - 1. Are automated,
 - 2. Are frequently updated & consumed (especially early in the project),
 - 3. Actively flag errors, and
 - 4. Are easy and fun to jump into
- Try to make your dashboards more like reports, so they
 - 1. Provide context and are self-explanatory,
 - 2. Are portable and archivable, and
 - 3. Contain statistical analyses that encourage sophisticated reasoning.



Pipeline (available at github.com/dss-ialh/displaying-health-data)

Raw CSVs

The external data's first appearance in our system.

`data-public/raw/`

Central Database

The inspected & loaded data into tables.

`data-public/derived/`

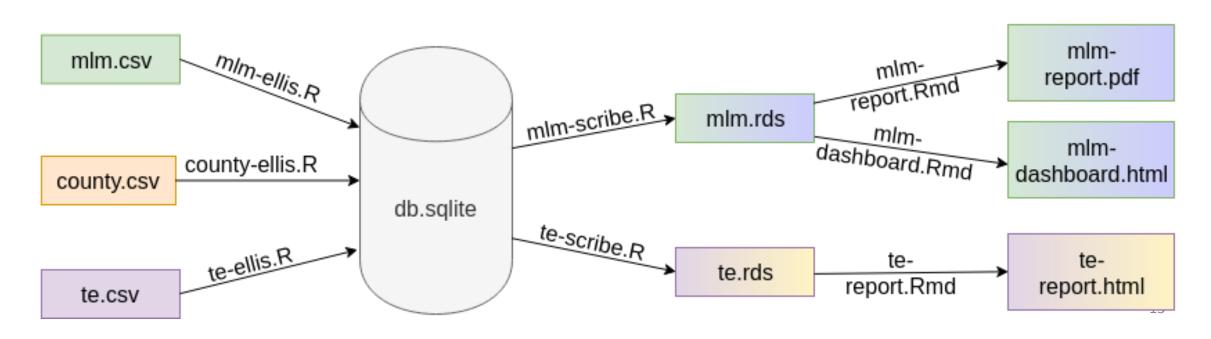
Analysis-Ready Rectangles

The joined & groomed datasets easily digestible.

`data-public/derived/`

Analysis

Products for internal & external audiences,



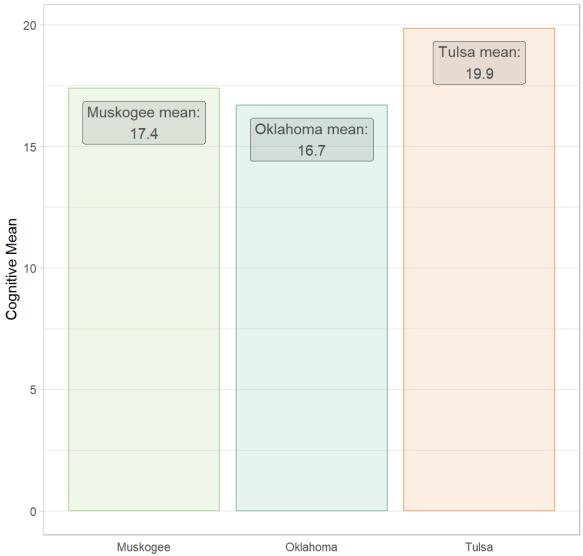
Fictional scenario

 We partner with the government to monitor the cognitive trajectories of patients over time.

 Recruited 200 seniors from three county programs.

 Dashboard is tailored to providers in Tulsa





In the real study...

The nurses create three 6-month mini-research studies.

They develop a manageable change to the services the provide, and see if their mini-intervention improves an outcome they've chosen.

One real intervention developed a thorough script of welcoming new clients. A second intervention was sending text messages to clients to reduce no-shows. A third intervention developed guidelines for discussing traumatic ACEs in their past.

Starts with 2-day workshop to develop the ideas.

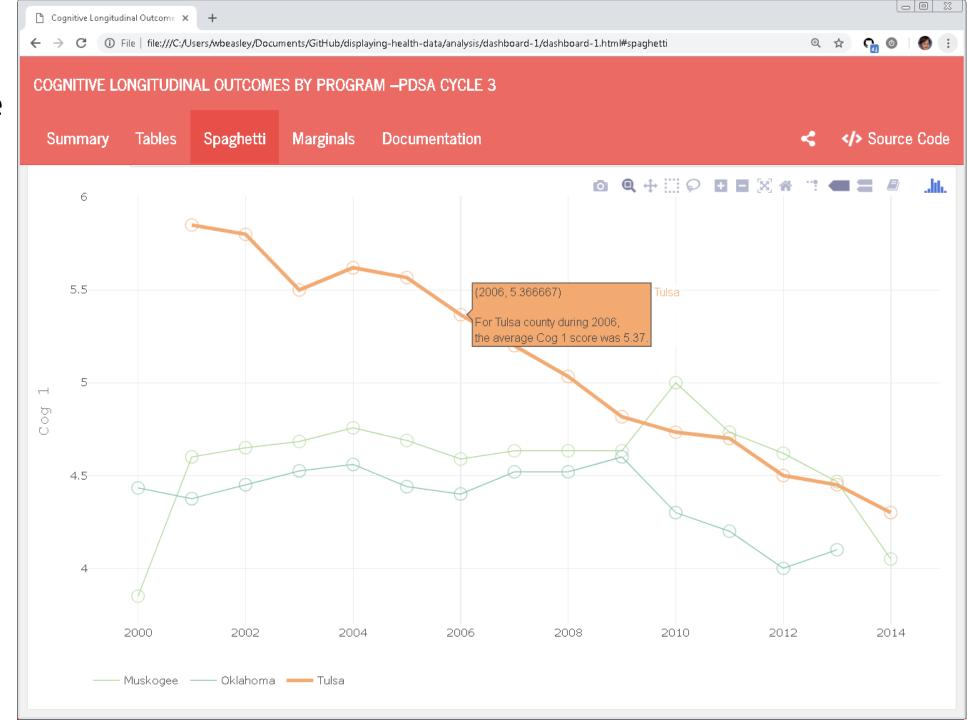
Data Glimpse

(start demo)

```
> ds %>% dplyr::glimpse(50)
Observations: 200
Variables: 21
$ subject_wave_id <int> 1, 2, 3, 4, 5, 6, 7, ...
$ subject_id
                  <int> 1001, 1001, 1001, 100...
$ county_id
                  <int> 51, 51, 51, 51, 51, 5...
$ gender_id
                  <db1> 2, 2, 2, 2, 2, 2, 2, ...
                  <chr> "Native Hawaiian or 0...
$ race
$ ethnicity
                  <chr> "Hispanic or Latino",...
                  <chr> "Muskogee", "Muskogee...
$ county
$ wave_id
                  <int> 1, 2, 3, 4, 5, 6, 7, ...
$ year
                  <int> 2000, 2001, 2002, 200...
                  <date> 2000-04-25, 2001-02-...
$ date_at_visit
                  <int> 67, 68, 69, 70, 71, 7...
$ age
                  <chr> "60s", "60s", "60s", ...
$ age_cut_4
$ age_80_plus
                  <int> 0, 0, 0, 0, 0, 0, 0, ...
                  <db1> 8.895, 8.895, 8.895, ...
$ int_factor_1
                  <db1> -0.029, -0.029, -0.02...
$ slope_factor_1
$ cog_1
                  <db1> 3.3, 3.5, 3.5, 3.4, 3...
                  <db1> 4.6, 4.6, 4.2, 4.7, 4...
$ cog_2
$ cog_3
                  <db1> 5.3, 5.3, 5.2, 5.2, 5...
$ phys_1
                  <db1> 2.8, 2.9, 2.8, 3.1, 3...
$ phys_2
                  <db1> 3.9, 3.5, 3.9, 3.9, 3...
$ phys_3
                  <db1> 0.5, 0.0, 0.7, 1.3, 2...
```

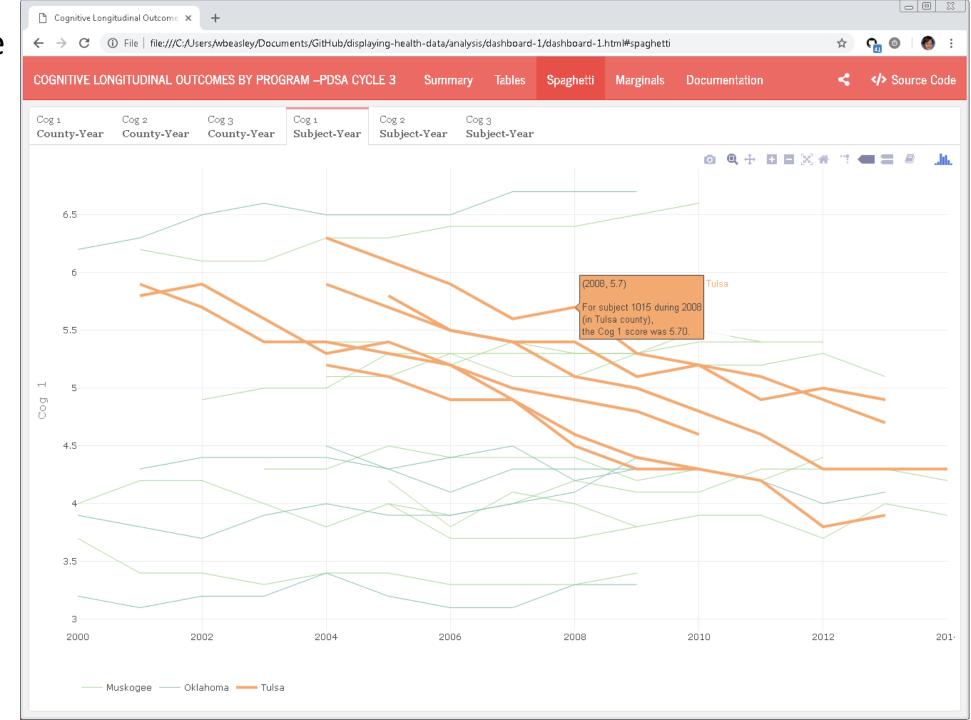
Program directors need to monitor the focal group over time.

De-emphasized programs provide meaningful reference points.



We also need to see the progress of individuals

Tulsa county trajectories are in orange. The others are primarily for context.



Some nurses were more comfortable starting with tables than graphs.



Histograms are particularly benefited from full-sentence hover text.



Each dashboard had several tabs of documentation.

Our quality varied across dashboards, and this is a representative example.

COGNITIVE LONGITUDINAL	_OUTCOMES BY PROGRAMPDS	A CYCLE 3	Summary	Tables	Spaghetti	Marginals	Documentation
Explanation -Current PDSA	Explanation -All CQI Dashboards	Glossary	Tips Config				

SMART Aim

By implementing plans for self-care activities with established clients, primary caregivers will see a mean score reduction of at least 4 points on the Parenting care activities by 12-1-18.

Measures

- Outcome: Parent Stress
 - · Pre- and Post- Parent Stress Scale score change
- Process: Completed Resiliency Plans
 - · Numerator: Count of clients who completed their designated self-care activity
 - · Denominator: Count of clients with a visist in a given week
- Disruptor: Cultural issues preventing participation, toxic stress, family catastrophe

Spaghetti Notes

- To be added:
 - · Post score plots
 - o Pre-post comparisons
 - · Splitting out new clients from old clients to see if differences are more or less prominent
- · Mean PSS Scores Pre:
 - · Each blue dot represents a single client's PSS score on the pre measure
 - The orange dots (connected by the red line) represent the mean PSS score for all clients on a given week.
 - Dot size is proportional to the number of clients who had a pre-PSS score.
- · Stress Activity Success:
 - · Numerator: Number of clients who indicated having completed their self-care activity
 - o Denominator: All clients who responded to the self-care question (Yes/No to self-care)
 - · Dot size: The dot size is proportional to the number of clients who responded to that question on a given week.

Resources

Current PDSA

Lessons Reinforced

1. The providers have incredible ideas for improving services delivered. It's worth our time to facilitate and translate these ideas.

2. Implementing their tx & measurements ideas increases their engagement.

3. Incorporate the providers in the dashboard design

4. Math phobia is real, and it might be hard for your developers to related to.

5. Writing the scribes is difficult (ie, grooms and reshapes multiple analysisagnostic data tables into a single analysis-specific rectangle).

Lessons Learned

1. This particular group of nurses preferred a top-down entry into their data. Most scientists prefer bottom-up. Here are some examples of how we changed our presentation.

2. Hovering text provides an alternative entry into understanding the graph.

3. Incorporate the providers in the dashboard design

4. Monthly meetings are not frequent enough at the beginning

Suggestions for dashboard improvements?

- Specific graphs
- Presentation order
- Verbal documentation
- My approach/attitude towards the information
- My approach/attitude towards the consumers
- Feature to emphasis or de-emphasize



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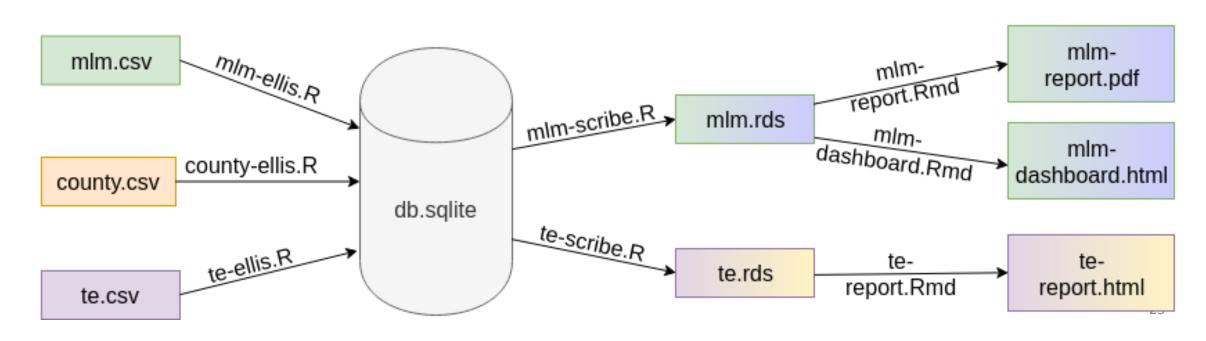
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Analysis

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Pipeline

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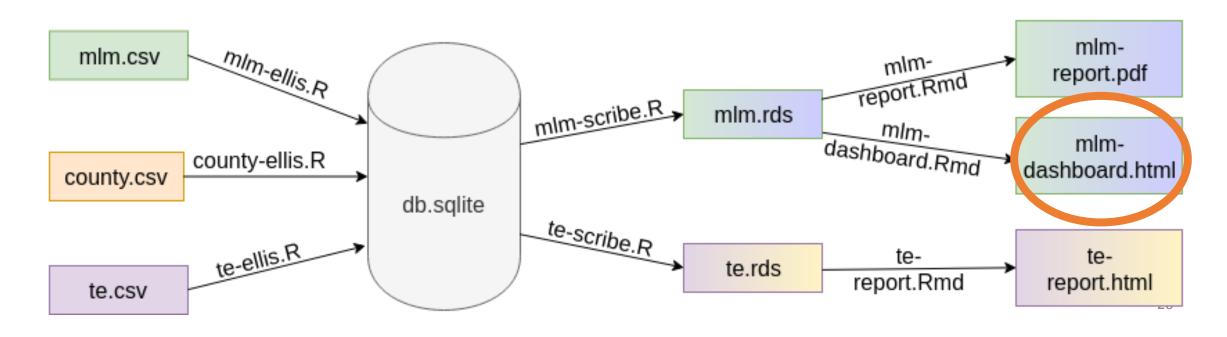
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Benefits

Scaling

- more data sources
- 2. multiple analysts
- 3. multiple reports

2. Hide PHI for as long as possible

3. Leverage database performance

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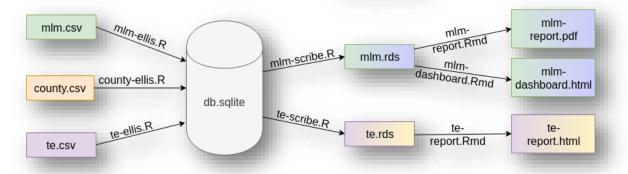
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Deployment Options

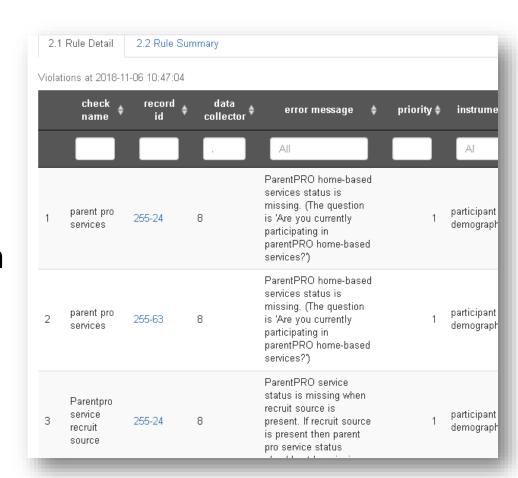
1. Server-rendered vs Client-rendered

2. Interactive vs static

3. Stage to redact PHI

4. Reconnect to PHI record using authentication

5. human vs cronjob



Providers in Oklahoma

- 1. Oklahoma State Department of Health—Persephone Starks, John Delara, Beth Martin
- Children First Tulsa LouAnn Beuke
- **3. LCDA** Patty Demoraes-Huffhine
- 4. Oklahoma City County Health Department Denise Howard, Diane Sammons, Sally Dixon
- **5. OKC Public Schools** Yolanda Lucero, Kethzia Njikam
- **6. PAT Bethany Public Schools** Mindy Turner, Jem Balderas
- 7. PAT Parent Promise Shawna Norman
- 8. Community Action Program Tulsa Dana James
- 9. Cherokee PARENTS Ben King, Jennifer Kirby, Amy Thilges
- 10. PCCT PAT Tulsa Sarah Neyman
- **11. SafeCare PCCT** Sheri Davis
- 12. SafeCare NorthCare Dwan McDonald

University of Oklahoma HSC Analysts for the MIECHV-3 CQI

David Bard, Will Beasley Geneva Marshall Thomas Wilson **Andrew Peters** Chris Aston Donna Wells Som Bohora Maleeha Shahid

Yutian Thompson

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