data cleaning principles

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Tidy data are all alike, but every messy dataset is messy in its own way.

Hadley Wickham

If I clean up [Medicare] data ...
does any of the knowledge I gain ...
apply to the processing of RNA-seq data?

Roger Peng

Data Mishaps Night

Join us for the first inaugural Data Mishaps Night! We will feature a lineup of data mistake stories with a focus on the human aspect of data work and lessons learned the hard way.



Caitlin Hudon & Laura Ellis dataMishapsNight.com

Data cleaning

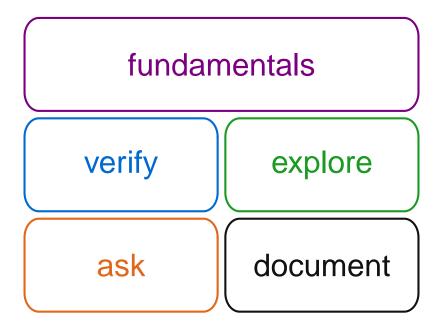
- ▶ tedious
- embarrassing
- needs context
- ► doesn't feel like progress

Data cleaning

- tedious
- embarrassing
- needs context
- ▶ doesn't feel like progress

- requires creativity
- ► requires coding prowess
- ► source of many problems

Data cleaning principles



1. Don't clean data when you're tired or hungry.

(paraphrasing Ghazal Gulati)

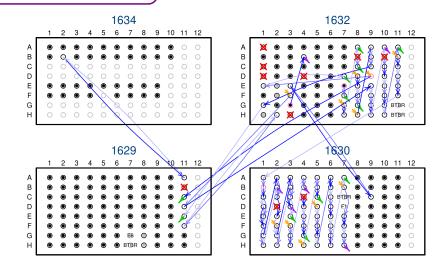
2. Don't trust anyone (even yourself)

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"my motto is 'trust no one' ...except maybe @kwbroman?"

Jenny Bryan

3. Think about what might have gone wrong and how it might be revealed



4. Use care in merging

	А	В	С	D	E	F	G		
1	id	glucose.0	glucose.5	glucose.15	glucose.30	insulin.0	insulin.5		
2	DO-221	145.742786	206.452638	216.640608	299.55501	0.74455	2.0264		
3	DO-222		A	В	С	D	Е	F	G
4	DO-223	1	id	glucose.0	insulin.0	glucose.5	insulin.5	glucose.15	insulin.15
5	DO-224	2	DO-321	66.839405	0.04	246.685995	0.04	305.26214	0.04
6	DO-225	3	DO-322	98.12509	0.51185	246.25574	1.4062	301.8201	2.828
7	DO-226	4	DO-323	94.68305	1.7812	448.1068	1.0248	521.61894	1.02725
8	DO-227	5	DO-324	121.051535	0.0882	407.355505	0.63475	470.541525	0.8195
9	DO-228	6	DO-325	122.95695	0.19155	298.193665	0.6467	323.148455	0.40515
10	DO-229	7	DO-326	201.447755	0.7454	386.51887	0.6081	654.99799	1.07225
11	DO-230	8	DO-327	130.025425	0.0509	477.302675	0.166	610.49733	0.4842
		9	DO-328	143.60919	0.23435	438.88705	0.70505	406.249135	0.2498
		10	DO-329	125.29262	0.04	543.74634	1.7366	520.205245	0.8498
		11	DO-330	135.61874	0.91275	393.03416	3.73095	454.62209	1.7325

5. Dates & categories suck

Principle:

a fundamental truth that guides our thinking

5. Dates & categories suck

6. Check that distinct things are distinct

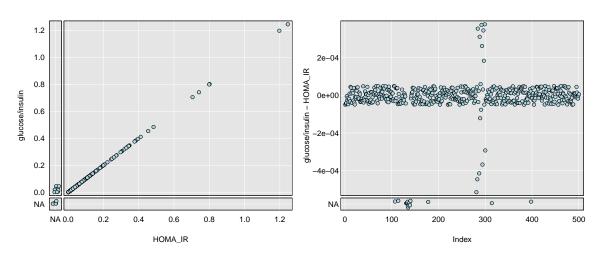
	Α	В	С	D	Е	F	G
1	WiscID	ID	NEOID	Fem_CA	Fem_Imax	Fem_Imin	Fem_J
2	F2.C1W.F.1248	1248	NEO183	0.7524	0.1427	0.1006	0.2433
3	F2.C1W.M.1250	1250	NEO184	0.7669	0.1556	0.09652	0.2521
4	F2.C1W.F.1251	1251	NEO185	0.7613	0.1549	0.09659	0.2515
5	F2.C1W.F.1254	1254	NEO186	0.7475	0.1503	0.08603	0.2363
6	F2.C1W.M.1257	1257	NEO187	0.8197	0.1849	0.1056	0.2905
7	F2F.715	715	NEO764	0.6017	0.09662	0.05969	0.1563
8	F2F.751	751	NEO765	0.7273	0.1304	0.08735	0.2178
9	F2F.1251	1251	NEO766	0.6675	0.1157	0.07814	0.1938
10	F2M.1340	1340	NEO768	0.6656	0.1387	0.08122	0.2199
11	F2.C1W.M.739	739	NEO779	0.9336	0.2828	0.1628	0.4456

7. Check that matching things match

	А	В	С	D
1	id	sex	n_gen	age_days
2	F20.25	М	20	75
3	F21.30	М	21	75
4	F21.68	М	21	71
5	F22.52	М	22	73
6	F21.71	F	22	63
7	F22.116	F	22	57
8	F21.F20.9.M5	М	20	82
9	F21.F20.18.M5	М	20	77
10	F20.26	М	20	75
11	F21.62	М	21	72

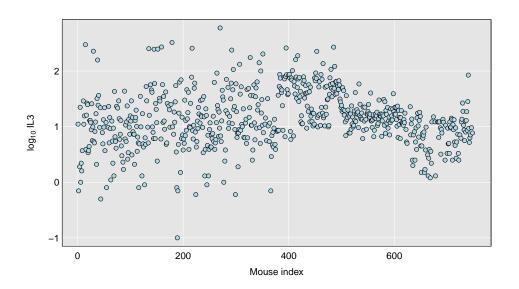
	А	В	С	D
1	id	sex	age_at_dosing	n_gen
2	F22.69	F	67	22
3	F22.106	F	69	22
4	F22.70	F	67	22
5	F22.107	F	69	22
6	F21.71	F	65	21
7	F22.116	F	62	22
8	F22.73	F	65	22
9	F22.117	F	62	22
10	F21.108	F	62	21
11	F22.118	F	59	22

8. Check calculations

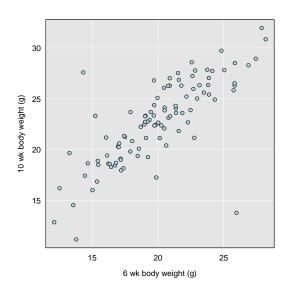


9. Look for other instances of a problem

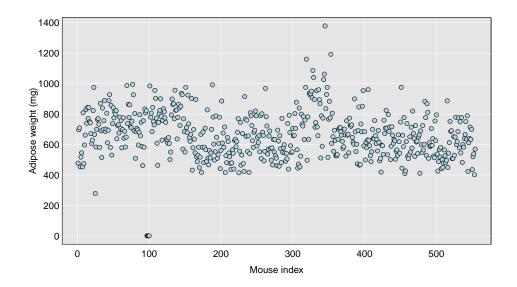
10. Make lots of plots



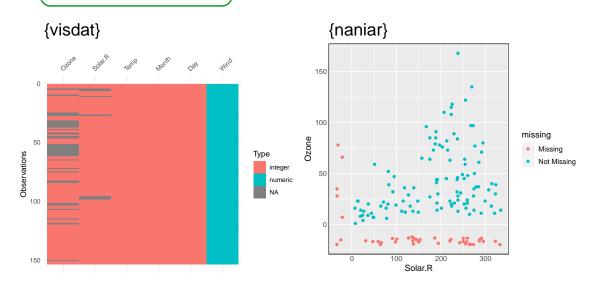
10. Make lots of plots



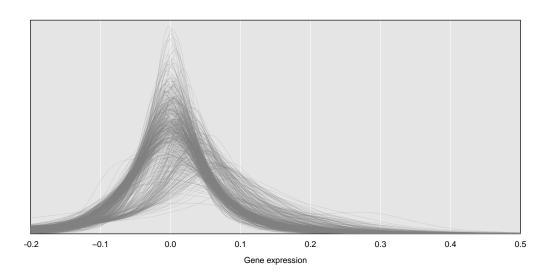
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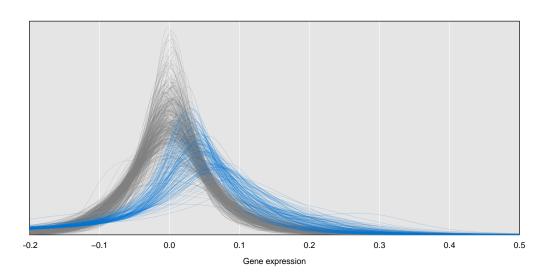
11. Look at missing value patterns



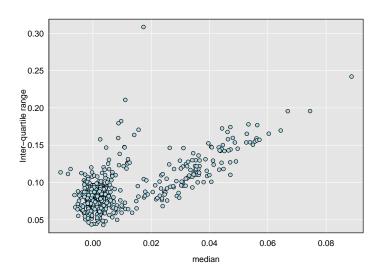
12. With massive data, make more plots not fewer



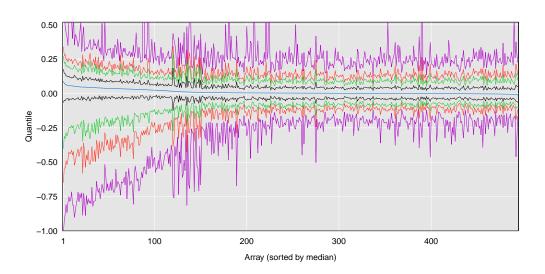
12. With massive data, make more plots not fewer



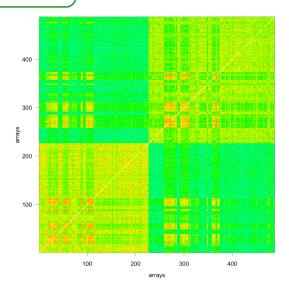
With massive data, make more plots not fewer



With massive data, make more plots not fewer



13. Follow up all artifacts



ask

- 14. Ask questions
- 15. Ask for the primary data
- 16. Ask for metadata
- 17. Ask why data are missing

document

- 18. Create checklists & pipelines
- 19. Document not just what but why
- 20. Expect to recheck

Data cleaning principles

fundamentals

- 1. Don't clean data when tired or hungry
- 2. Don't trust anyone (even yourself)
- 3. Think about what might have gone wrong
- 4. Use care in merging
- 5. Dates & categories suck

verify

- 6. Verify that distinct things are distinct
- 7. Verify that matching things match
- 8. Check calculations
- 9. Look for other instances of problems

explore

- 10. Make lots of plots
- 11. Look at missing value patterns
- 12. With big data make more plots
- 13. Follow up all artifacts

ask

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- 15. Ask for the primary data
- 16. Ask for metadata
- 17. Ask why data are missing

document

- 18. Create checklists & pipelines
- 19. Document not just what but why
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I will let the data speak for itself when it cleans itself.

Allison Reichel

Slides: kbroman.org/Talk_DataCleaning



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