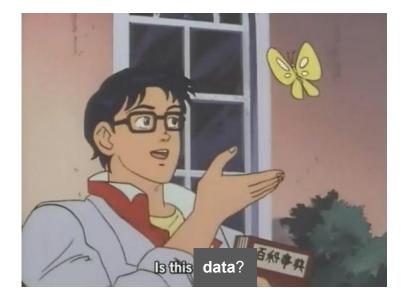
Data Literacy 101: For Water Conservation Professionals

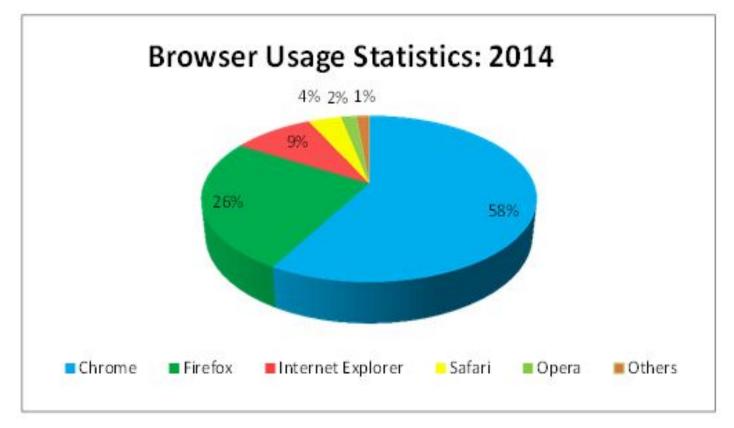




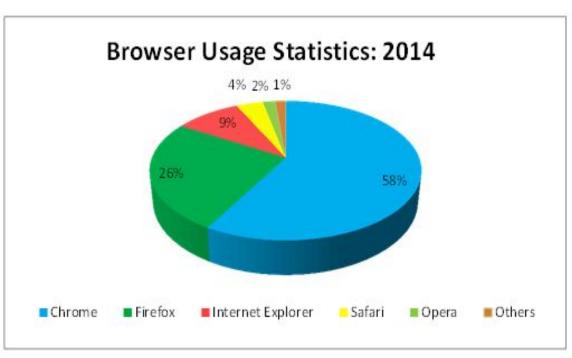
Warm-up: Is it data?



- 4 examples
- In the chat reply "Yes" or "No" and any justification
- Think about the reason **why** you selected what you did



NO: This is a data visualization (pie chart)



- What is the total that each percentage is derived from?
- How many "Others" are there, and what are they?

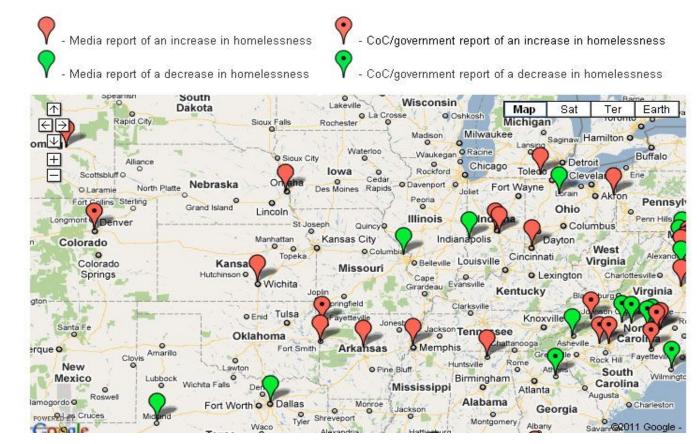
Monthly	Cash Flow		
	Actual	Budget	Variance
Cash received			
Fees	\$21,571	\$20,000	\$1,571
Salary grants	10,005	11,000	(995)
Other	76	8490 S.C.P.	76
	31,652	31,000	652
Cash paid out			
Salaries and benefits	21,575	20,000	(1,575)
Food	2,350	2,000	(350)
Play supplies	335	500	165
Other	3,270	1,500	(1,770)
-	27,530	24,000	(3,530)
Excess of cash received over cash			
paid out	\$4,122	\$7,000	\$(2,878)

Is it data? NO: This is a report

Monthly	Cash Flow		
	Actual	Budget	Variance
Cash received			
Fees	\$21,571	\$20,000	\$1,571
Salary grants	10,005	11,000	(995)
Other	76		76
	31,652	31,000	652
Cash paid out			
Salaries and benefits	21,575	20,000	(1,575)
Food	2,350	2,000	(350)
Play supplies	335	500	165
Other	3,270	1,500	(1,770)
_	27,530	24,000	(3,530)
Excess of cash received over cash			
paid out	\$4,122	\$7,000	\$(2,878)

• How does this month compare to last month?

• Which category is the most over-budget?



Is it data? NO: This is a visualization/map

Y - Media report of an increase in homelessness

 ${f Y}\,$ - Media report of a decrease in homelessness

- CoC/government report of an increase in homelessness

- CoC/government report of a decrease in homelessness



Which regions/states are seeing the most change?

What does a "report" mean? Are there numbers to back it up?

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	Result	s In Order of Roll Numbers	100 XTX0 XX02 017 017										S21.
	ROLL	NAME	CCC DED	UCC DED	DED TOT	ENC	DALL L		E ENO	1100	O TOTAL	I amount	1
					PER_TOT			G_KNW		MCQ	G_TOTAL	MERIT	
	700002	SAIMA KASHEM BARSA	21.36	37.00	58.36	18.30	13.80	21.00	0.00	53.10	111.46	3996	
	700002 700008	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM		37.00	58.36			21.00 40.50	0.00		111.46	3996 365	
	700002 700008 700009	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM	21.36 27.00 20.28	37.00 43.00 37.00	58.36 70.00 57.28	18.30 15.30 21.00	13.80 13.80 15.00	21.00 40.50 30.60	0.00 0.00 0.00	53.10 69.60 66.60	111.46 139.60 123.88	3996 365 1893	
	700002 700008 700009	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM	21.36 27.00	37.00 43.00 37.00	58.36 70.00 57.28	18.30 15.30	13.80 13.80	21.00 40.50 30.60 41.40	0.00	53.10 69.60	111.46 139.60	3996 365 1893 266	
L	700002 700008 700009 700010	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM	21.36 27.00 20.28	37.00 43.00 37.00	58.36 70.00 57.28 73.00	18.30 15.30 21.00	13.80 13.80 15.00	21.00 40.50 30.60	0.00 0.00 0.00	53.10 69.60 66.60	111.46 139.60 123.88 142.00	3996 365 1893 266	
L	700002 700008 700009 700010 700020	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI	21.36 27.00 20.28 25.50	37.00 43.00 37.00 47.50	58.36 70.00 57.28 73.00 56.14	18.30 15.30 21.00 15.00	13.80 13.80 15.00 12.60	21.00 40.50 30.60 41.40	0.00 0.00 0.00 0.00	53.10 69.60 66.60 69.00	111.46 139.60 123.88 142.00 114.64	3996 365 1893 266 3490	
L	700002 700008 700009 700010 700020 700021	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN	21.36 27.00 20.28 25.50 19.14	37.00 43.00 37.00 47.50 37.00	58.36 70.00 57.28 73.00 56.14 63.00	18.30 15.30 21.00 15.00 9.00	13.80 13.80 15.00 12.60 16.50	21.00 40.50 30.60 41.40 33.00	0.00 0.00 0.00 0.00 0.00	53.10 69.60 66.60 69.00 58.50	111.46 139.60 123.88 142.00 114.64 120.60	3996 365 1893 266 3490 2433	
L	700002 700008 700009 700010 700020 700021 700028	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS	21.36 27.00 20.28 25.50 19.14 24.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00	18.30 15.30 21.00 15.00 9.00 12.30	13.80 13.80 15.00 12.60 16.50 13.20	21.00 40.50 30.60 41.40 33.00 32.10	0.00 0.00 0.00 0.00 0.00 0.00	53.10 69.60 66.60 69.00 58.50 57.60	111.46 139.60 123.88 142.00 114.64 120.60 140.60	3996 365 1893 266 3490 2433 321	
L	700002 700008 700009 700010 700020 700021 700028 700029	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN	21.36 27.00 20.28 25.50 19.14 24.00 24.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86	18.30 15.30 21.00 15.00 9.00 12.30 22.50	13.80 13.80 15.00 12.60 16.50 13.20 17.40	21.00 40.50 30.60 41.40 33.00 32.10 32.70	0.00 0.00 0.00 0.00 0.00 0.00 0.00	53.10 69.60 66.60 69.00 58.50 57.60 72.60 60.00	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86	3996 365 1893 266 3490 2433 321 3615	
l	700002 700008 700009 700010 700020 700021 700028 700029 700044	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86	37.00 43.00 37.00 37.00 39.00 44.00 34.00 46.70	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72	18.30 15.30 21.00 15.00 9.00 12.30 22.50 19.50	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 66.60 69.00 58.50 57.60 72.60 60.00	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86 127.22	3996 365 1893 266 3490 2433 321 3615 1412	
l	700002 700008 700009 700010 700020 700021 700028 700029 700044 700046	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR MD. SAJJAD HOSEN	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86 28.02	37.00 43.00 37.00 47.50 37.00 39.00 44.00 34.00 46.70 41.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72 62.00	18.30 15.30 21.00 15.00 9.00 12.30 22.50 19.50 9.00	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00 15.00	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50 28.50	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 66.60 69.00 58.50 57.60 72.60 60.00 52.50	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86 127.22 129.50	3996 365 1893 266 3490 2433 321 3615 1412 1151	
l	700002 700008 700009 700020 700021 700028 700029 700029 700044 700046 700053	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR MD. SAJJAD HOSEN MD.JAHANGIR HOSSAIN	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86 28.02 21.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00 34.00 46.70 41.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72 62.00 64.00	18.30 15.30 21.00 9.00 12.30 22.50 19.50 9.00 19.50	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00 15.00 12.00	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50 28.50 36.00	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 69.00 58.50 57.60 72.60 60.00 52.50 67.50	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86 127.22 129.50 124.00	3996 365 1893 266 3490 2433 321 3615 1412 1151 1876	
l	700002 700008 700010 700020 700021 700028 700029 700024 700044 700046 700053 700062	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR MD. SAJJAD HOSEN MD.JAHANGIR HOSSAIN SHIRIN AKTER SONIA	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86 28.02 21.00 24.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00 34.00 46.70 41.00 40.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72 62.00 64.00 50.00	18.30 15.30 21.00 9.00 12.30 22.50 19.50 9.00 19.50 19.50 16.50	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00 15.00 12.00 16.50	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50 28.50 28.50 36.00 27.00	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 66.60 58.50 57.60 72.60 60.00 52.50 67.50 60.00	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86 127.22 129.50 124.00 102.50	3996 365 1893 266 3490 2433 321 3615 1412 1151 1876 4736	
	700002 700008 700009 700010 700020 700028 700028 700028 700044 700046 700053 700046 700053	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR MD. SAJJAD HOSEN MD.JAHANGIR HOSSAIN SHIRIN AKTER SONIA MD. BELAL HOSSAIN	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86 28.02 21.00 24.00 18.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00 44.00 46.70 41.00 40.00 32.00 35.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72 62.00 64.00 50.00 56.00	18.30 15.30 21.00 15.00 9.00 12.30 22.50 19.50 9.00 19.50 16.50 15.00	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00 15.00 12.00 16.50 10.50	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50 28.50 36.00 27.00 27.00	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 66.60 58.50 57.60 72.60 60.00 52.50 67.50 60.00 52.50	111.46 139.60 123.88 142.00 114.64 120.60 140.60 140.60 113.86 127.22 129.50 124.00 102.50 106.70	3996 365 1893 266 3490 2433 321 3615 1412 1151 1876 4736 4513	
	700002 700008 700009 700020 700021 700028 700028 700029 700044 700046 700046 700053 700062 700069 700069	SAIMA KASHEM BARSA MOHAMMAD M.S ALAM MOHAMMAD SHAHIDUL ISLAM MOHAMMAD NUR NABI ABDUL BATEN KARTICK CHANDRA DAS URMY RAHMAN MEHER NIGAR MD. SAJJAD HOSEN MD.JAHANGIR HOSSAIN SHIRIN AKTER SONIA MD. BELAL HOSSAIN SANJIDA SANJANA NOVA	21.36 27.00 20.28 25.50 19.14 24.00 24.00 19.86 28.02 21.00 24.00 18.00 24.00 21.00	37.00 43.00 37.00 47.50 37.00 39.00 44.00 44.00 46.70 41.00 40.00 32.00 35.00 39.00	58.36 70.00 57.28 73.00 56.14 63.00 68.00 53.86 74.72 62.00 64.00 50.00 56.28	18.30 15.30 21.00 15.00 9.00 12.30 22.50 19.50 9.00 19.50 16.50 15.00 12.00	13.80 13.80 15.00 12.60 16.50 13.20 17.40 12.00 15.00 12.00 16.50 10.50 10.50	21.00 40.50 30.60 41.40 33.00 32.10 32.70 28.50 28.50 36.00 27.00 27.00 23.10	0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	53.10 69.60 66.60 58.50 57.60 72.60 60.00 52.50 67.50 60.00 52.50 50.70	111.46 139.60 123.88 142.00 114.64 120.60 140.60 113.86 127.22 129.50 124.00 102.50 106.70 115.98	3996 365 1893 266 3490 2433 321 3615 1412 1151 1876 4736 4513 3278	

Is it data? NO: This is a PDF

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		s In Order of Roll Numbers											
		NAME	SSC_PER					G_KNW			G_TOTAL		
		SAIMA KASHEM BARSA	21.36	37.00	58.36	18.30		21.00	0.00	53.10		3996	
		MOHAMMAD M.S ALAM	27.00	43.00	70.00	15.30	13.80	40.50	0.00	69.60		365	
		MOHAMMAD SHAHIDUL ISLAM	20.28	37.00	57.28	21.00	15.00	30.60	0.00	66.60	123.88	1893	
		MOHAMMAD NUR NABI	25.50	47.50	73.00	15.00	12.60	41.40	0.00	69.00	142.00	266	
		ABDUL BATEN	19.14	37.00	56.14	9.00	16.50	33.00	0.00	58.50		3490	
		KARTICK CHANDRA DAS	24.00	39.00	63.00	12.30	13.20	32.10	0.00	57.60		2433	
		URMY RAHMAN	24.00	44.00	68.00	22.50	17.40	32.70	0.00	72.60	140.60	321	
		MEHER NIGAR	19.86	34.00	53.86	19.50	12.00	28.50	0.00	60.00	113.86	3615	
		MD. SAJJAD HOSEN	28.02	46.70	74.72	9.00	15.00	28.50	0.00	52.50		1412	
		MD.JAHANGIR HOSSAIN	21.00	41.00	62.00	19.50	12.00	36.00	0.00	67.50		1151	
		SHIRIN AKTER SONIA	24.00	40.00	64.00	16.50	16.50	27.00	0.00	60.00	124.00	1876	
		MD. BELAL HOSSAIN	18.00	32.00	50.00	15.00	10.50	27.00	0.00	52.50		4736	
		SANJIDA SANJANA NOVA	21.00	35.00	56.00	12.00	15.60	23.10	0.00	50.70		4513	
		ROBIUL AWALL	23.28	39.00	62.28	15.00	9.00	29.70	0.00	53.70		3278	
		MOHAMMAD MONIRUL ISLAM	24.78	40.00	64.78	14.70	12.30	24.90	0.00	51.90		3139	
		S.M.SAIFUL ISLAM	24.78	38.00	62.78	15.60	13.80	19.80	0.00	49.20	111.98	3923	~

 What is the average MERIT score?

• Who had the highest overall performance?



Takeaways

- All of these examples are created from raw material a bunch of letters and numbers in a file somewhere. That's the data.
- The real data should be easy to slice and dice to ask new questions
- Think about the data behind the chart, map, or report.
 What does the rest of it look like? Why are you only being shown this subset?

Some data about me

- Christopher Tull
- Born in Ventura, CA. Live in Oxnard CA.
- Studied Computer Science and Mathematics at CSU Channel Islands
- Spent a year in NYC learning how to use data to help cities operate more efficiently



• Now project manager for the California Data Collaborative

About the California Data Collaborative

- Formed in 2016 as a nonprofit coalition of water agencies and has grown into a trusted presence in California's water sector
- Staffed by a team of data and design experts, and governed by a steering committee of water managers

CALIFORNIA DATA COLLABORATIVE

Today we will talk about:

- The Universe of Water Agency Data
- Data Types and Formats
- Accessing and Sharing Data
- Metadata
- Data Terms Kahoot!
- Potential and Pitfalls
- An Applied Example

The Universe of Water Agency Data

Agency Budgets Account Balances Employee Detail	s Emails IT System Logs Contracts
Interaction between Agency and Customer	Interaction between Agency and System
BillsCall NotesRebatesField VisitsBill AdjustmentsFeesShutoffsBill InsertsMarketing ProgramsClasses	Meter Reads New Connections Capital Improvements Repairs Replacements Inspections Maintenance
Customers	System
Name Address Household Size Landscape Billing Preferences Customer Class Dates of Service	Area ServedMeter CharacteristicsPipe DetailsClimateEvapotranspirationFacility DetailsOperating Specs

The Universe of Water Agency Data

Agency Budgets Account Balances Employee Detail	s Emails IT System Logs Contracts
Interaction between Agency and CustomerBillsCall NotesRebatesField VisitsBill AdjustmentsFeesShutoffsBill InsertsMarketing ProgramsClasses	Interaction between Agency and SystemMeter ReadsNew ConnectionsCapital ImprovementsRepairsReplacementsInspectionsMaintenance
Customers Name Address Household Size Landscape Billing Preferences Customer Class Dates of Service	SystemArea Served Pipe Details Facility DetailsMeter Characteristics Climate Operating Specs

Water Efficiency Data

Name

Billed Use

Call Notes



Customer Class

Year Built

Rebates

Landscape Area

Evapotranspiration

Quantitative Data

Name

Billed Use (7 CCF)

Call Notes



Customer Class

Year Built (1960)

Rebate Amount (\$250)

Landscape Area (2000 Sq.Ft)

Evapotranspiration (57 in.)

Qualitative Data

Name (Kamala Harris)

Billed Use

Call Notes ("Asked about a water audit...") Landscape Area



Customer Class (Single Family)

Year Built

Rebate **Type** (WBIC)

Evapotranspiration

Categorical (Qualitative) Data

Name

Billed Use

Call Notes



Customer Class (Single Family)

Year Built

Rebate **Type** (WBIC)

Landscape Area

Evapotranspiration

Ordinal (Qualitative) Data

Name

Billed Use

Call Notes



Customer Class

Year Built

Rebate Date (4/22/2021) *

Landscape Area

Evapotranspiration

*(sort of...)

Data formats

Easy for people to understand vs. easy for computers to understand ("machine readable")

Common machine readable data types:

FORMAT	OFTEN USED FOR	EXAMPLE
Shapefile (.shp), Geojson (.json)	Spatial data (points, lines, polygons)	Parcel boundaries
Comma Separated Values (.csv), Excel (.xlsx)	Tabular data, spatial data in point form (lat/lng)	Records of rebate participation
Text (.txt)	Textual data	Call notes of customer interactions
Raster (.jpg, .png, .tif)	Images	Aerial photograph
Semi-structured (.json, .yaml)	Sending information over the internet	CIMIS API

customers.csv



account, class, use_ccf, landscape_area

1,Single Family,6,1500

2,Single Family,17,2300

3,Multi Family,274,5000

4,Commercial,153,1300

5, Irrigation, 55, 9000

6, Irrigation, 825, 23000

customers.json



"account": 1, "class": "Single Family", "use_ccf": 6, "landscape_area": 1500 , "account": 2, "class": "Single Family",

customers.shp



Spatial files encode the coordinates of points, lines, and polygons.

Shapefiles are a binary format - need the right software to make sense of it or it ends up like

this:

NUL	NOL
NUL	NUTNUTNUTNUTNUTNUTNUTNUTNUTNUTNUTNUTNUTN
NUL	NULNULNULNULNULNULNULNULNULNULNULNULNULN
SOH	NULNULNULݵ SUBÖ0WAÁÊ-QI9>ANULNULNULSTXNULNULNUL
SOH	NULNUL;
ETY:	3WA-FUCDD>ANULNULNULETXNULNULNUL
SOH	NULNULÀ
US	Ú2WAýü; c-E>ANULNUL EOT NULNUL
SOH	NULNULNUL•11ªã1WAmY"aSD>ANULNULNULENQNULNULNUL
SOH	NULNULNUL (-;ÿ×0WAÕ•, Ûþ@>ANULNULNULACKNULNULNUL
SOH	NULNULNULÐ1UŞÚOWAD)ØB÷@>ANULNULNULBELNULNULNUL
SOH	NULNULNUL-EÍÈïOWAÏ~;âSOHA>ANULNULNULBSNULNULNUL
SOH	NULNULNULÒÁ2±%1WAFFWCAN9-A>ANULNULNUL NULNULNUL
SOH	NULNULKÇ/'1WAoék8±a>aNULNULNUL
NUL	NULNUL
SOH	NULNULNULÖ-ZIK1WA Yã>, B>ANULNULNULVTNULNULNUL
SOH	NULNULNUL°4G1WA/ESCfïSIC>ANULNULNULFENULNULNUL
SOH	NULNUL-3u 1WAŸIYÁDLEC>ANULNULNUL
NUL	NULNUL
SOH	NULNULNUL (c-RSÊ3WAÀ*Œ +?>ANULNULNULSONULNULNUL

customers.geojson



Spatial files encode the coordinates of points, lines, and polygons.

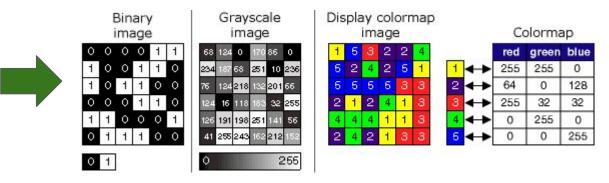
Geojson is an open format based on JSON

"type": "Feature", "geometry": { "type": "Point", "coordinates": [125.6, 10.1] }, "properties": { "name": "Dinagat Islands" }

customers.tif

Raster files encode data in a grid of values



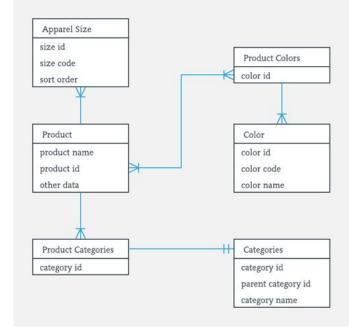


A note on databases

"an organized collection of data, generally stored and accessed electronically from a computer system."

A relational database is often used for more permanent data storage. Has nice properties to ensure data consistency and allows querying data with Structured Query Language (SQL)

SELECT * FROM product WHERE product_name = 'TShirt'





Open data vs. closed data

Data availability is a spectrum!



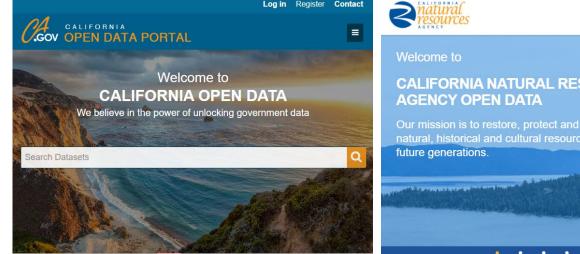
- Complete datasets
- Easily and permanently available in electronic format
- No unreasonable barriers to access or restrictions on re-use
- Machine readable, non-proprietary format
- Reliable, complete metadata
- Does NOT contain sensitive personal information



Open Data™

Log in Register Contact

Open Data™





CALIFORNIA NATURAL RESOURCES AGENCY OPEN DATA

future generations.

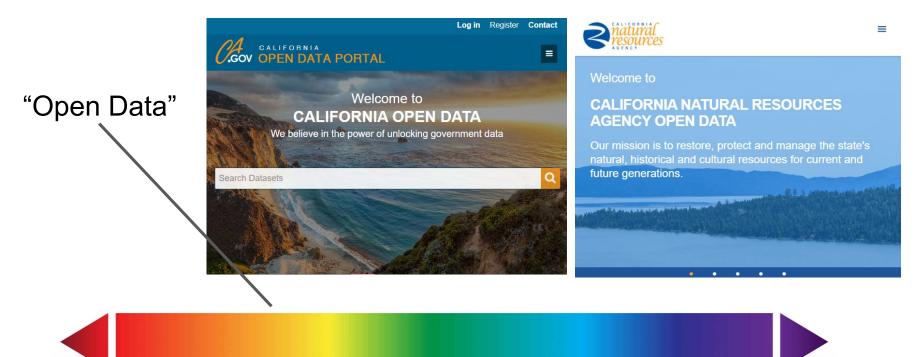


• Complete datasets

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- Complete datasets
- Easily and permanently available in electronic format
- No unreasonable barriers to access or restrictions on

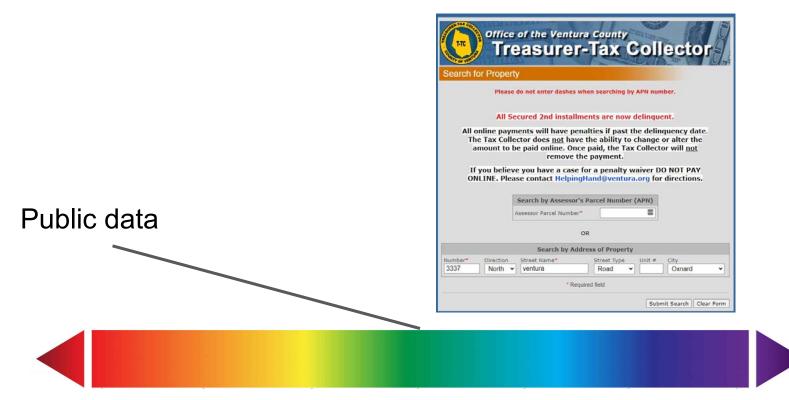
re-use

- Machine readable, non-proprietary format
- Reliable, complete metadata

Public data

Does NOT contain sensitive personal information





- No way for the general public to access unless requested, e.g. by public records act request
- Does NOT contain sensitive personal information

Example: Internal communications of a public agency

Non-public data

Data Openness

- No way for the general public to access.
- Accessible by others under the right circumstances and with the right precautions
- **CAN CONTAIN** sensitive personal information

Example: Water meter reads and billing data



Private data

Even this can be viewed on a spectrum!

How would you rank the following customer data from most to least sensitive? Why?

- Account balance
- Meter geolocation
- Credit card number
- Water use
- Name or email address



Data Openness

Private data

Data Openness

- Security-critical data
- No way for the general public to access.
- Accessible by others only when absolutely necessary and under the highest security precautions
- CAN CONTAIN sensitive personal information or information that threatens public safety

Example: Data that could allow access to or manipulation of critical infrastructure

Data Access

- What is an API?
- Application Programming Interface
- Basically a way for external programmers to interact with a software system
- Common example retrieving data from a REST API

https://webservices.mwdsc.org/wins/Pub lic/api/MeterInterval/OC-81/2019-10-03/ 2019-10-03?type=json



[{"MeterDate":"2019-10-03T00:00:00","MeterID":"OC-81","Flow":1.92,"Start Date":"1995-07-01T00:00:00","EndDate":"2045-12-31T00:00:00","BillCustID ":"OC","IntervalNum":1.0,"EndMeterReading":26778939.0,"Volume":0.04132 2,"ProcessedFlag":"2"},{"MeterDate":"2019-10-03T00:00:00","MeterID":"OC-81","Flow":1.91,"StartDate":"1995-07-01T00:00:00","EndDate":"2045-12-31 T00:00:00","BillCustID":"OC","IntervalNum":2.0,"EndMeterReading":267789 56.0,"Volume":0.039027,"ProcessedFlag":"2"},{"MeterDate":"2019-10-03T00:00:00","EndDate":"2019-10-03T00:00:00","BillCustID":"OC","IntervalNum":2.0,"EndMeterReading":267789 56.0,"Volume":0.039027,"ProcessedFlag":"2"},{"MeterDate":"2019-10-03T00:00:00","EndDate":"2019-10-03T00:00:00","EndDate":"2019-10-03T00:00:00","MeterID":"OC","IntervalNum":2.0,"EndMeterReading":267789 56.0,"Volume":0.039027,"ProcessedFlag":"2"},{"MeterDate":"2019-10-03T00:00:00","EndDate":"2019-10-03T00:00:00","MeterID":"OC-81","Flow":1.91,"StartDate":"1995-07-01T00:00:00","EndDate":"2019-10-03T00:00:00","MeterID":"OC-81","Flow":1.91,"StartDate":"1995-07-01T00:00:00","EndDate":"2045-12-31T00:00:00","BillCustID":"OC","IntervalNum":3.0,"EndMeterReading":26778973.0,"Volume":0.039027,"ProcessedFlag":"2"},...

Data About Data

Who took this photo?

When was this photo taken?

What spectral bands were collected?



Can you use this photo?

How was the photo processed?

How frequently will it be revised?

Metadata

Who took this photo? (USDA)

When was this photo taken? (2018)

What spectral bands were collected? (RGB + NIR)



Can you use this photo? (Yes)

How was the photo processed? (Color-infrared)

How frequently will it be revised? (Every 2 years)

Metadata

Good metadata...

- Makes it easy to share your data with a consultant
- Enables someone else to take over a data set you maintain
- Helps you stay sane when you revisit something years later
- Reduces the chance that data is misinterpreted

Metadata Example

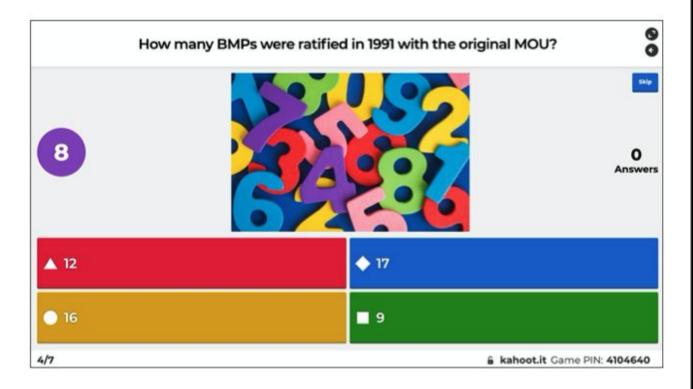
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17-DEV1-CM	VVIIal U			• V V I		Ultra Low Water Urinal	2	2		
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17-DEV1-HT	RES	13-AUG-18	20-AUG-1	18	28-JUN-18 CW	High Efficiency Clothes Washer		1	WHIRLPOOL	WFW7590FV
17-DEV1-HT	RES	13-AUG-18	20-AUG-1	18	02-JUN-18 CW	High Efficiency Clothes Washer		1	WHIRLPOOL	WFW7590FV
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17-DEV1-HT	RES	04-OCT-18)7-DEC-1	18	03-JUL-18 CW	High Efficiency Clothes Washer	1	1	WHIRLPOOL	WFW7590FV
17-DEV1-HT	RES	04-OCT-18)7-DEC-1	18	11-AUG-18 CW	High Efficiency Clothes Washer		1	WHIRLPOOL	WFW7590FV
17-DEV1-CM	СОММ	1			<u>IC</u>	Weather-Based Irrigation Controller	2	1 24	4	
17-DEV1-CM	СОММ	05-DEC-18	01-MAR-1	-19	27-NOV-18 FT	Toilet (tank and bowl)		7 /	0 NIAGARA	N7716R: N7
17-DEV1-CM	СОММ	Constant of the second	Contraction of the		FT	Toilet (tank and bowl)	36	1		
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		/			FT	Toilet (tank and bowl)		.6		
17-DEV1-CM	СОММ	/			FT	Toilet (tank and bowl)	2			
17-DEV1-CM		/			FT	Toilet (tank and bowl)		4		

Activity: Data Terms Jeopardy

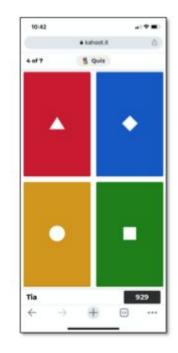
Introducing CalWEP's newest virtual addition.....



What you see on the webinar screen



What you see on your phone screen





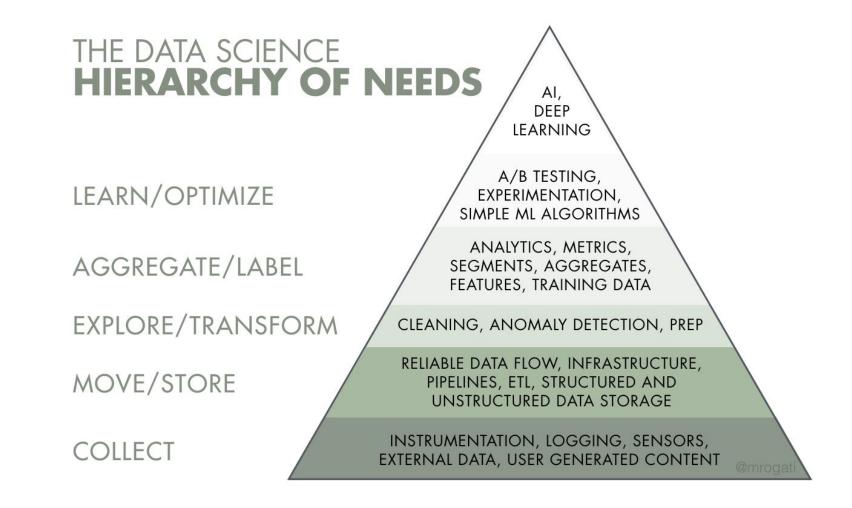
In your phone's browser open kahoot.it

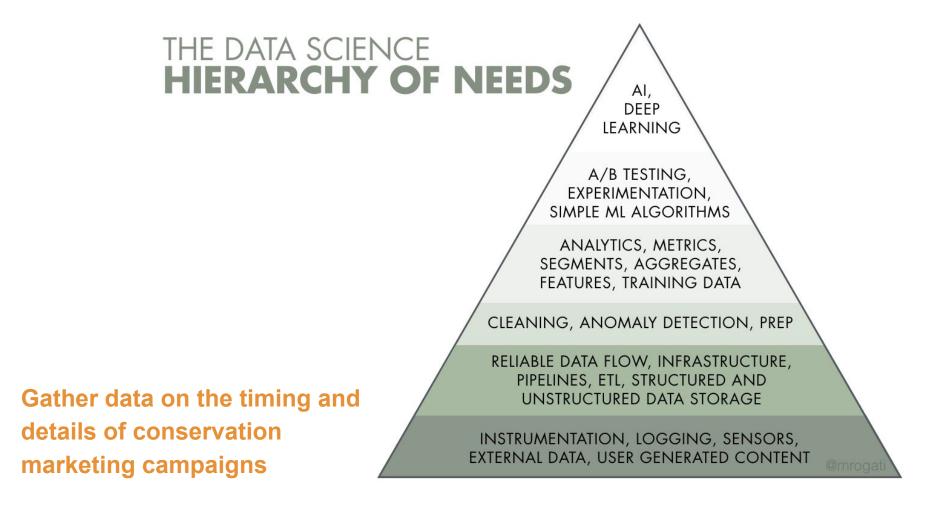
Once we launch the game you will see a pin and be asked to enter your name.

Then....we play.

(A few ways) data can be used

Check assumptions Tell stories Make (support) decisions Inform questions Enable new approaches





Put systems into place so data makes it to reliable long-term data storage (e.g. a database)

THE DATA SCIENCE

HIERARCHY OF NEEDS

A/B TESTING, EXPERIMENTATION, SIMPLE ML ALGORITHMS

AI, DEEP LEARNING

ANALYTICS, METRICS, SEGMENTS, AGGREGATES, FEATURES, TRAINING DATA

CLEANING, ANOMALY DETECTION, PREP

RELIABLE DATA FLOW, INFRASTRUCTURE, PIPELINES, ETL, STRUCTURED AND UNSTRUCTURED DATA STORAGE

INSTRUMENTATION, LOGGING, SENSORS, EXTERNAL DATA, USER GENERATED CONTENT



THE DATA SCIENCE

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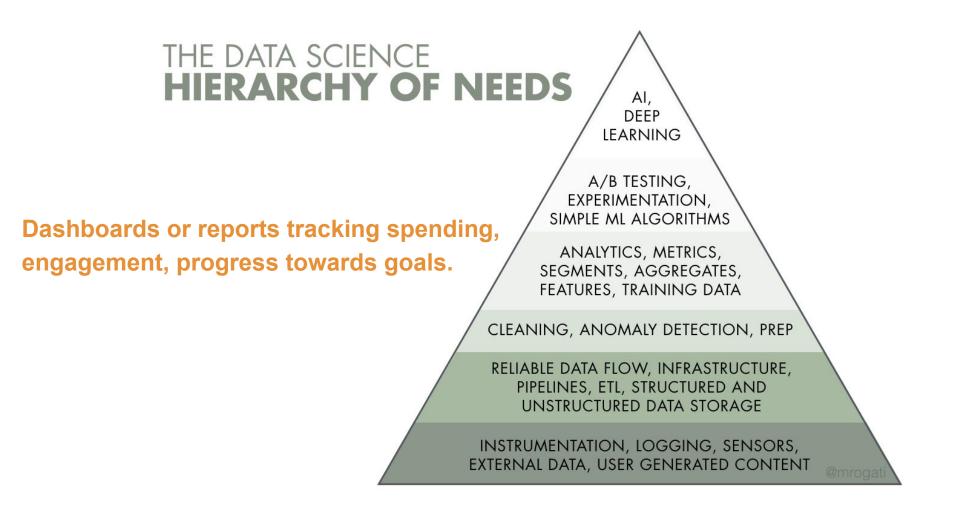
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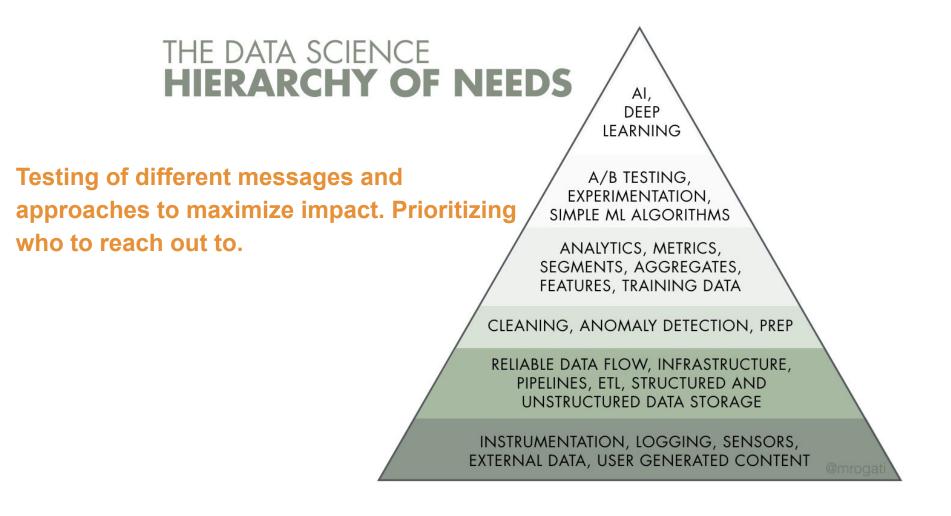
ANALYTICS, METRICS, SEGMENTS, AGGREGATES, FEATURES, TRAINING DATA

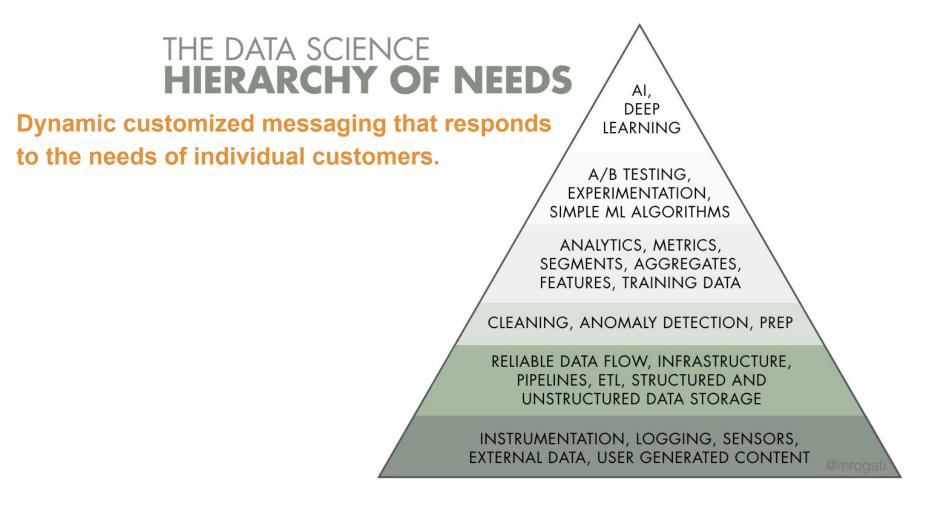
CLEANING, ANOMALY DETECTION, PREP

RELIABLE DATA FLOW, INFRASTRUCTURE, PIPELINES, ETL, STRUCTURED AND UNSTRUCTURED DATA STORAGE

INSTRUMENTATION, LOGGING, SENSORS, EXTERNAL DATA, USER GENERATED CONTENT







Two Purposes for Data

- Data is often originally collected for **administrative** purposes.
 - Billing customers
 - Providing a rebate check
- Administrative data is often repurposed for analysis after the fact.
 - Finding inefficient customers
 - Evaluating program effectiveness
- If analysis is planned for from the beginning, it usually goes more smoothly!
 - Know what you plan analyze
 - Collect the right data
 - Keep good metadata

Activity: Data Storytelling

Tell the story of this data set:

[Purple, Beach, Button, Coffee, Experience]

Data is not neutral

Humans make choices about:

- Data collection methods
- Defining terms
- Analyzing/applying data
- Visualizing data

Minimize bias when possible, identify biases that cannot be removed, and be clear about assumptions and methodology that may affect reading of data How many ways could you answer the question:

"How much water did your agency use last year?"

Production? Metered? Sales? Calendar or fiscal year? Pass-through? Potable?...

Pitfalls to watch out for



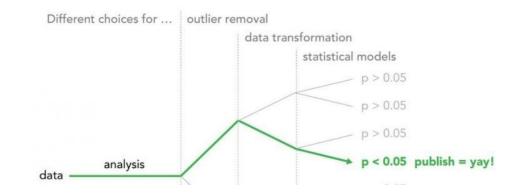
- Data isn't what you expected
- Not sure what the data means or how/why data was collected
- Data taken out of context

Pitfalls to watch out for

"How is the sausage made?"

- 1. Data collection
- 2. Data interpretation
- 3. Data transformation / Analysis
- 4. Presentation of results
- 5. Interpretation of results

Garden of forking paths [Gelman and Loken 2014]



Applied Example: LAM Data Assessment

Goals of LAM Assessment

 Quality control the Landscape Area Measurement (LAM) data that DWR has provided to agencies

2. Estimate residential portions of water use objective to quantify the impact for the agency

Where to start?

Let's break down the formula to see what data we need.

Residential Indoor

Residential Outdoor



Standards Data Constant

DWR's LAM data is derived from many different data sources with different considerations

- Imagery
- Classification Algorithm
- Agency Boundary
- Parcel Polygons



DWR's LAM data is derived from many different data sources with different considerations

- Imagery
- Classification Algorithm
- Agency Boundary
- Parcel Polygons

Raster data, 4-band imagery, year, time of year, resolution, errors in images?



DWR's LAM data is derived from many different data sources with different considerations

- Imagery
- Classification Algorithm
- Agency Boundary
- Parcel Polygons

Preprocessing steps, type of algorithm, accuracy, format of outputs



DWR's LAM data is derived from many different data sources with different considerations

- Imagery
- Classification Algorithm
- Agency Boundary
- Parcel Polygons

Does the boundary accurately represent

the area where you provide water?



(GPCD * Pop) + (ETAF * ETO * LA * C) Indoor standard Population Outdoor standard Reference ET Landscape Area Scaling Factor

DWR's LAM data is derived from many different data sources with different considerations

- Imagery
- Classification Algorithm
- Agency Boundary
- Parcel Polygons

Parcel georectification, alignment of land use codes with agency classes, overlap

of parcels with irrigated areas



(GPCD * Pop) + (ETAF * ETO * LA * C)

Indoor standard

Population

Outdoor standard

Reference ET

Landscape Area Scaling Factor

- Only way to check many of these things is to open up the files and look!
- Shapefiles opened in ArcGIS, QGIS or similar
- Compare parcel locations against residential meter locations



Zooming in on Evapotranspiration

"Data": {
 "Providers": [
 {
 "Name": "cimis",
 "Type": "station",
 "Owner": "water.ca.gov",
 "Records": []

"Dater: "2010-01-01", "Julian": "1", "Station": "2", "Standard": "english", "ZipCodes": "93624", "Scope": "daily", "DayAirTmpAvg": { "Value": "39", "Oct: " ".

"Unit": "(F)" }, "DayAirTmpMax": { "Value": "57.3", "Qc": " ", "Unit": "(F)" }, "DayAirTmpMin": {

"Value": "29.8", "Qc": " ", "Unit": "(F)"

... Results Truncated ...

3.

- Where to source data on reference evapotranspiration?
- CIMIS is standard. Spatial CIMIS vs. Individual Stations.
- CIMIS has a REST API!

http://et.water.ca.gov/api/data?appKey=YOUR-APP-KEY&targets=2,8,127&startDate= 2010-01-01&endDate=2010-01-05

(GPCD * Pop) + (ETAF * ETO Indoor standard Population Outdoor standard Reference ET * LA * C) Landscape Area Scaling Factor

Zooming in on Population

- How to calculate population?
 - Number of meters * Household Size
 - Census
 - American Community Survey
- Seasonal population?



Today we talked about:

- The Universe of Water Agency Data
- Data Types and Formats
- Accessing and Sharing Data
- Metadata
- Data Terms Kahoot!
- Potential and Pitfalls
- An Applied Example

Resources

- Data Terms
 - o <u>https://www.springboard.com/blog/data-science-terms/</u>
 - <u>https://www.dataquest.io/blog/data-science-glossary/</u>
- Data needs pyramid
 - <u>https://hackernoon.com/the-ai-hierarchy-of-needs-18f111fcc007</u>



More info: CalWEP.org

CA Water Data Summit August 19-20, 2021

More info: CaWaterDataSummit.org

Questions?