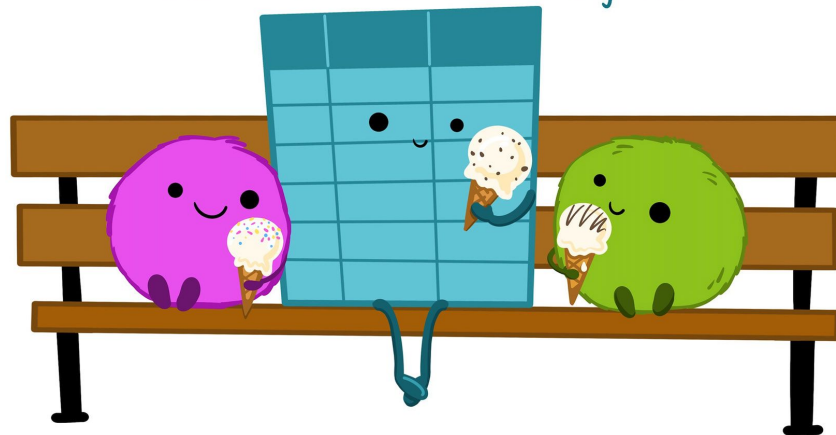


Ch. 8 Intro to Tidy Data

NCEAS Learning Hub
Arctic Data Center

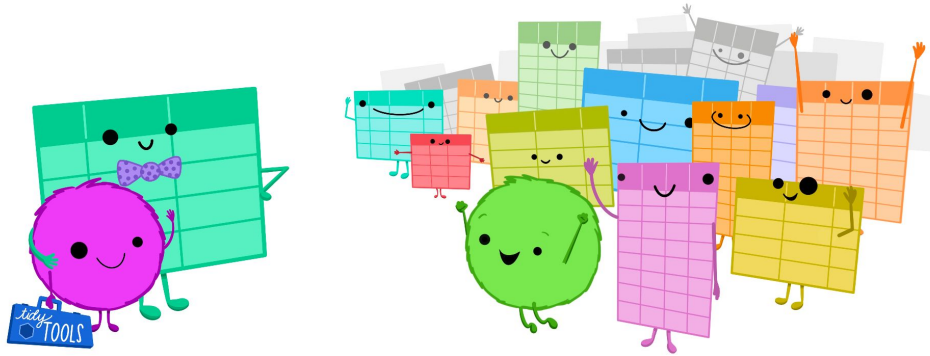
January 2024

make friends with tidy data.



Learning Objectives

- Understand basics of relational data models, aka **tidy data**
- Learn how to design and create effective data tables



artwork by [@allison_horst](https://twitter.com/allison_horst)

Introduction

The organizational structure that allows for relating data tables

A type of database that contains data tables that are related to one another

An organized collection of information (i.e. data)

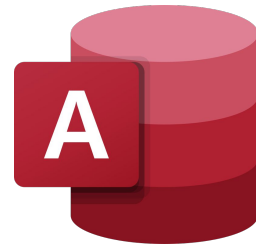
book sections here, in case you want to cross reference!

Relational data **models** are what relational **databases** use to organize tables.

However, you don't need to be using a relational database (e.g. mySQL, MariaDB, Oracle, Microsoft Access) to benefit from using a relational data model.

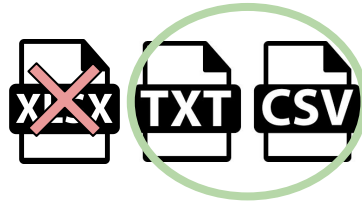
Benefits:

- Powerful search & filtering
- Handle large, complex datasets
- Enforce data integrity
- Decrease errors from redundant updates



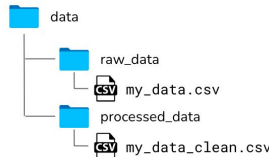
Simple Guidelines for Data Management (Borer et al. 2009)

Use a scripted program



Nonproprietary file formats

Keep a raw version of the data



mooredCTD_site1_2020-2023.txt
 mooredCTD_site2_2020-2023.txt
 mooredCTD_site3_2020-2023.txt

Descriptive names

Header line

date	site_name	temp_c
2023-01-01	site1	16.3
2023-02-01	site2	15.9
2023-03-01	site3	16.1

A B C / 1 2 3

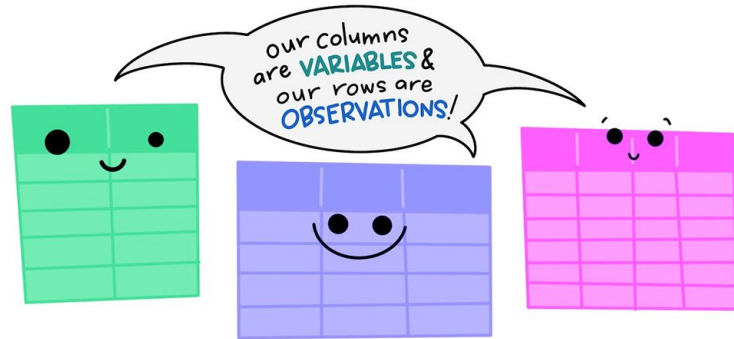
Plain ASCII text

Simple Guidelines for Data Management (Borer et al. 2009)

- Design your tables to add rows, not columns
- Each column should contain only one type of information
- Record a single piece of data only once; separate information collected at different scales into different tables -- in other words, create a *relational database*

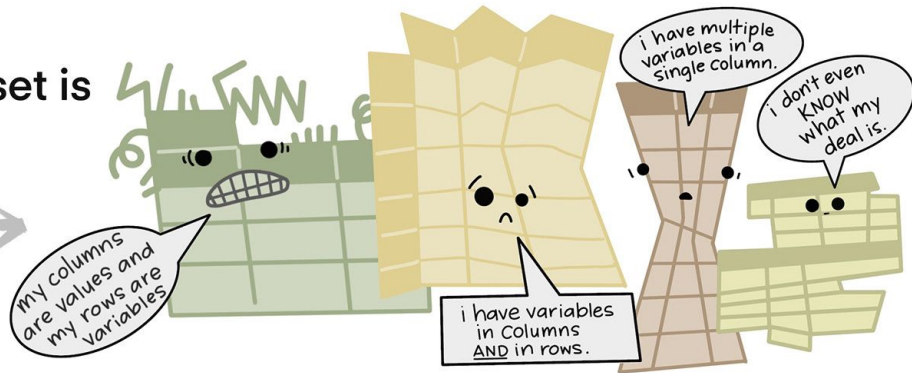
Recognizing “untidy” data

The standard structure of tidy data means that “tidy datasets are all alike...”



“...but every messy dataset is messy in its own way.”

—HADLEY WICKHAM



artwork by [@allison_horst](#)

Recognizing “untidy” data

A not-so-tidy spreadsheet received by NCEAS....

AtlasGroveCOMPLETE.xls

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
species	tree	main trunks kg	reiterated trunks kg	limbs kg	branches kg	leaves kg		type	species	main trunk	reiteration	dry masses (kg)		leaf	TOTAL	% total
												limb	branch			
SESE	Atlas	255144.9	46020.6	5477.7	13433.2	1101.2		tree	SESE	3569312	213247	53714	230945	17192	4084409	95.3491
SESE	Ballantine	221966.4	7651.6	5922.9	11210.0	1084.8		tree	PSME	135815	0	0	8338	961	145114	3.3876
SESE	Bell	253246.4	5454.3	5792.6	48500.7	1043.4		tree	THSE	31799	0	0	6343	864	39006	0.9105
SESE	Broken Top	130928.9	4805.2	1608.1	5137.4	729.9		tree	ACMA	4444	0	0	925	264	5634	0.1315
SESE	Buena Vista	128833.0	3486.5	0.0	8552.1	518.4		tree	UMCA	2921	0	0	937	273	4131	0.0964
SESE	Demeter	155896.0	11085.6	3204.3	10054.1	768.7		shrub	RUSP	0	0	0	1974	686	2660	0.0620
SESE	Epimetheus	226987.0	12915.7	1797.2	13585.2	1029.4		fern	POMU	0	0	0	0	1271	1271	0.0296
SESE	Iluvatar	349586.6	65003.9	12315.6	13987.0	1461.8		shrub	VAOV	0	0	0	526	26	552	0.0129
SESE	Kronos	134154.1	12204.4	7232.7	5036.1	597.3		shrub	COCO	0	0	0	284	6	289	0.0067
SESE	Pleiades I	182385.2	3735.0	1935.2	10846.6	762.2		fern	POSC	0	0	0	107	89	196	0.0045
SESE	Pleiades II	235838.8	11183.4	4306.0	11306.5	877.7		tree	RHPU	100	0	0	44	18	162	0.0037
SESE	Prometheus	239414.0	25228.9	1612.6	12458.2	1086.0		herb	OXOR	0	0	0	0	112	112	0.0026
SESE	Rhea	143710.4	487.8	730.1	5524.2	691.2		shrub	VAPA	0	0	0	94	4	99	0.0023
SESE	Zeus	243365.7	2885.5	1620.4	19104.7	954.3		tree	PISI	0	0	0	1	0	1	0.0000
SESE	3	1761.3	0.0	0.0	87.6	41.4		tree	CHLA	0	0	0	1	0	0	
SESE	4	6312.0	356.0	73.5	214.1	43.8		shrub	GASH	0	0	0	0	0	0	
SESE	5	206.0	0.0	0.0	8.7	2.5		shrub	SACA	0	0	0	0	0	0	
SESE	6E	18697.4	0.0	0.0	1055.2	66.3				3744390	213247	53714	250519	21767	42838	
SESE	6W	14651.5	7.7	0.0	626.3	49.6										
SESE	11	614.4	0.0	0.0	28.1	17.0										
SESE	12	232.1	0.0	0.0	11.2	10.3										
SESE	18	15632.0	0.0	0.0	946.3	106.8										
SESE	19	11805.5	0.0	0.0	770.1	80.3			SESE geo	3569312	213247	53714	230945	17192	4084409	
SESE	20	309.5	0.0	0.0	12.5	5.9			SESE epi	0	0	0	0	0	0	
SESE	22	25618.3	0.0	0.0	1504.0	120.2			PSME geo	135815	0	0	8338	961	145114	
SESE	23	463.7	0.0	0.0	18.9	4.5			PSME epi	0	0	0	0	0	0	
SESE	25	87.7	0.0	0.0	4.1	1.3			TSHE geo	31740	0	0	6332	860	38906	
SFSF	30	512.1	1.8	0.0	18.7	8.7			TSHE epi	59	0	0	12	4	65	
									ACMA geo	4444	0	0	925	264	5634	
									ACMA epi	0	0	0	0	0	0	

es in a column.

I don't even know what my deal is.

Recognizing “untidy” data - multiple tables

Easy for humans to interpret (sort of?), hard for computer programs (e.g. R)

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
species	tree	main trunks kg	reiterated trunks kg	limbs kg	branches kg	leaves kg		type	species	main trunk	reiteration	dry masses (kg)		leaf	TOTAL	% total
SESE	Atlas	255144.9	46020.6	5477.7	13433.2	1101.2		tree	SESE	3569312	213247	53714	230945	17192	4084409	95.3491
SESE	Ballantine	221966.4	7651.6	5922.9	11210.0	1084.8		tree	PSME	135815	0	0	8338	961	145114	3.3876
SESE	Bell	253246.4	5454.3	5792.6	48500.7	1043.4		tree	THSE	31799	0	0	6343	864	39006	0.9105
SESE	Broken Top	130928.9	4805.2	1608.1	5137.4	729.9		tree	ACMA	4444	0	0	925	264	5634	0.1315
SESE	Buena Vista	128833.0	3486.5	0.0	8552.1	518.4		tree	UMCA	2921	0	0	937	273	4131	0.0964
SESE	Demeter	155896.0	11085.6	3204.3	10054.1	768.7		shrub	RUSP	0	0	0	1974	686	2660	0.0620
SESE	Epimetheus	226987.0	12915.7	1797.2	13585.2	1029.4		fern	POMU	0	0	0	0	1271	1271	0.0296
SESE	Iluvatar	349586.6	65003.9	12315.6	13987.0	1461.8		shrub	VAOV	0	0	0	26	552	0.0129	
SESE	Kronos	134154.1	12204.4	7232.7	5036.1	597.3		shrub	COCO	0	0	0	6	289	0.0067	
SESE	Pleiades I	182385.2	3735.0	1935.2	10846.6	762.2		fern	POSC	0	0	0	107	89	196	0.0045
SESE	Pleiades II	235838.8	11183.4	4306.0	11306.5	877.7		tree	RHPU	100	0	0	44	18	162	0.0037
SESE	Prometheus	239414.0	25228.9	1612.6	12458.2	1086.0		herb	OXOR	0	0	0	0	112	112	0.0026
SESE	Rhea	14777.5	487.5	730.1	5524.2	691.2		shrub	VAPA	0	0	0	94	4	99	0.0023
SESE	Zeus	24339.0	885.5	1620.4	19104.7	954.3		tree	PISI	0	0	0	1	0	1	0.0000
SESE	3	1761.3	0.0	0.0	87.6	41.4		tree	CHLA	0	0	0	1	0	1	0.0000
SESE	4	6312.0	356.0	73.5	214.1	43.8		shrub	GASH	0	0	0	0	0	0	0.0000
SESE	5	206.0	0.0	0.0	8.7	2.5		shrub	SACA	0	0	0	0	0	0	0.0000
SESE	6E	18697.4	0.0	0.0	1055.2	66.3				3744390	213247	53714	250519	21767	4283636	
SESE	6W	14651.5	7.7	0.0	626.3	49.6										proportion
SESE	11	614.4	0.0	0.0	28.1	17.0										total
SESE	12	232.1	0.0	0.0	11.2	10.3										geophytic
SESE	18	15632.0	0.0	0.0	946.3	106.8		SESE geo		3569312	213247	53714	230945	17192	4084409	1.00
SESE	19	11805.5	0.0	0.0	770.1	80.3		SESE epi		0	0	0	0	0	0	
SESE	20	309.5	0.0	0.0	12.5	5.9		PSME geo		135815	0	0	8338	961	145114	1.00
SESE	22	25618.3	0.0	0.0	1504.0	120.2		PSME epi		0	0	0	0	0	0	
SESE	23	463.7	0.0	0.0	18.9	4.5		THSE geo		31740	0	0	6343	860	38932	0.99
SESE	25	87.7	0.0	0.0	4.1	1.3		THSE epi		59	0	0	12	4	74	
SESE	30	512.1	1.8	0.0	18.7	8.7		ACMA geo		4444	0	0	925	264	5634	1.00
								ACMA epi		0	0	0	0	0	0	

INSTEAD: create separate tables/files for each entity measured

Recognizing “untidy” data - inconsistent observations

Each row corresponds to more than one observation

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
species	tree	main trunks kg	reiterated trunks kg	limbs kg	branches kg	leaves kg		type	species	main trunk	reiteration	limb	branch	leaf	TOTAL	% total
SESE	Atlas	255144.9	46020.6	5477.7	13433.2	1101.2		tree	SESE	3569312	213247	53714	230945	17192	4084409	95.3491
SESE	Ballantine	221966.4	7651.6	5922.9	11210.0	1084.8		tree	PSME	135815	0	0	8338	961	145114	3.3876
SESE	Bell	253246.4	5454.3	5792.6	48500.7	1043.4		tree	THSE	31799	0	0	6343	864	39006	0.9105
SESE	Broken Top	130928.9	4805.2	1608.1	5137.4	729.9		tree	ACMA	4444	0	0	925	264	5634	0.1315
SESE	Buena Vista	128833.0	3486.5	0.0	8552.1	518.4		tree	UMCA	2921	0	0	937	273	4131	0.0964
SESE	Demeter	155896.0	1100.0	3204.3	10054.1	768.7		shrub	RUSP	0	0	0	1974	686	2660	0.0620
SESE	Epimetheus	226987.0	12915.7	1797.2	13585.2					0	0	0	1271	1271	0.0296	
SESE	Iluvatar	349586.6	65003.9	15015.6	13987.0					0	0	0	526	26	552	0.0129
SESE	Kronos	134154.1	12204.4	723.7	5036.1					0	0	0	284	6	289	0.0067
SESE	Pleiades I	182385.2	3735.0	1935.2	10846.6					0	0	0	107	89	196	0.0045
SESE	Pleiades II	235838.8	11183.4	4306.0	1306.5					0	0	0	44	18	162	0.0037
SESE	Prometheus	239414.0	25228.9	1612.6	12409.2					0	0	0	0	112	112	0.0026
SESE	Rhea	143710.4	487.8	730.1	5524.2					0	0	0	94	4	99	0.0023
SESE	Zeus	243365.7	2885.5	1620.4	19104.7					0	0	0	1	0	1	0.0000
SESE	3	1761.3	0.0	0.0	87.6					0	0	0	1	0	1	0.0000
SESE	4	6312.0	356.0	73.5	214.1					0	0	0	0	0	0	0.0000
SESE	5	206.0	0.0	0.0	8.7					0	0	0	0	0	0	0.0000
SESE	6E	18697.4	0.0	0.0	1055.2					247	53714	250519	21767	4283636		
SESE	6W	14651.5	7.7	0.0	626.3	49.8										proportion
SESE	11	614.4	0.0	0.0	28.1	17.0				main trunk	reiteration	limb	branch	leaf	total	geophytic
SESE	12	232.1	0.0	0.0	11.2	10.3			SESE geo	3569312	213247	53714	230945	17192	4084409	1.00
SESE	18	15632.0	0.0	0.0	946.3	106.8			SESE epi	0	0	0	0	0	0	
SESE	19	11805.5	0.0	0.0	770.1	80.3			PSME geo	135815	0	0	8338	961	145114	1.00
SESE	20	309.5	0.0	0.0	12.5	5.9			PSME epi	0	0	0	0	0	0	
SESE	22	25618.3	0.0	0.0	1504.0	120.2			THSE geo	31740	0	0	6332	860	38932	0.99
SESE	23	463.7	0.0	0.0	18.9	4.5			THSE epi	59	0	0	12	4	74	
SESE	25	87.7	0.0	0.0	4.1	1.3			ACMA geo	4444	0	0	925	264	5634	1.00
SESE	30	512.1	1.8	0.0	18.7	8.7			ACMA epi	0	0	0	0	0	0	

INSTEAD: each row should represent a single observed entity

Recognizing “untidy” data - inconsistent variables

Each column contains more than one variable type

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
species	tree	main trunks kg	reiterated trunks kg	limbs kg	branches kg	leaves kg		type	species	main trunk	reiteration	limb	branch	leaf	TOTAL	% total
SESE	Atlas	255144.9	46020.6	5477.7	13433.2	1101.2		tree	SESE	3569312	213247	53714	230945	17192	4084409	95.3491
SESE	Ballantine	221966.4	7651.6	5922.9	11210.0	1084.8		tree	PSME	135815	0	0	8338	961	145114	3.3876
SESE	Bell	253246.4	5454.3	5792.6	48500.7	1043.4		tree	THSE	31799	0	0	6343	864	39006	0.9105
SESE	Broken Top	130928.9	4805.2	1608.1	5137.4	729.9		tree	ACMA	4444	0	0	925	264	5634	0.1315
SESE	Buena Vista	128833.0	3486.5	0.0	8552.1	518.4		tree	UMCA	2921	0	0	937	273	4131	0.0964
SESE	Demeter	155896.0	11085.6	3204.3	10054.1	768.7		shrub	RUSP	0	0	0	1974	686	2660	0.0620
SESE	Epimetheus	226987.0	12915.7	1797.2	13585.2	1029.4		fern	POMU	0	0	0	0	1271	1271	0.0296
SESE	Iluvatar	349586.6	65003.9	12315.6	13987.0	1461.8		shrub	VAOV	0	0	0	526	26	552	0.0129
SESE	Kronos	134154.1	12204.4	7232.7	5036.0	1000.0				0	0	0	284	6	289	0.0067
SESE	Pleiades I	182385.2	3735.0	1935.2	10846.0	1000.0				0	0	0	107	89	196	0.0045
SESE	Pleiades II	235838.8	11183.4	4306.0	11306.0	1000.0				0	0	0	44	18	162	0.0037
SESE	Prometheus	239414.0	25228.9	1612.6	12458.0	1000.0				0	0	0	0	112	112	0.0026
SESE	Rhea	143710.4	487.8	730.1	5524.0	1000.0				0	0	0	94	4	99	0.0023
SESE	Zeus	243365.7	2885.5	1620.4	19104.0	1000.0				0	0	0	1	0	1	0.0000
SESE	3	1761.3	0.0	0.0	87.0	1000.0				0	0	0	1	0	1	0.0000
SESE	4	6312.0	356.0	73.5	214.0	1000.0				0	0	0	0	0	0	0.0000
SESE	5	206.0	0.0	0.0	8.0	1000.0				0	0	0	0	0	0	0.0000
SESE	6E	18697.4	0.0	0.0	1054.0	1000.0				213247	53714	250519	21767	4283636		
SESE	6W	14651.5	7.7	0.0	626.0	1000.0										proportion geophytic
SESE	11	614.4	0.0	0.0	24.0	1000.0				reiteration	limb	branch	leaf	total		
SESE	12	232.1	0.0	0.0	11.2	10.3		SESE	geo	3569312	213247	53714	230945	17192	4084409	1.00
SESE	18	15632.0	0.0	0.0	946.3	106.8		SESE	epi	0	0	0	0	0	0	
SESE	19	11805.5	0.0	0.0	770.1	80.3		PSME	geo	135815	0	0	8338	961	145114	1.00
SESE	20	309.5	0.0	0.0	12.5	5.9		PSME	epi	0	0	0	0	0	0	
SESE	22	25618.3	0.0	0.0	1504.0	120.2		THSE	geo	31740	0	0	6332	860	38932	0.99
SESE	23	463.7	0.0	0.0	18.9	4.5		THSE	epi	59	0	0	12	4	74	
SESE	25	87.7	0.0	0.0	4.1	1.3		ACMA	geo	4444	0	0	925	264	5634	1.00
SESE	30	512.1	1.8	0.0	18.7	8.7		ACMA	epi	0	0	0	0	0	0	

INSTEAD: all values in a column should be of the same type (tip: compare units)

Recognizing “untidy” data - marginal sums & stats

Marginal sums & statistics are combinations of observations

AtlasGroveCOMPLETE.xls

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
species	tree	main trunks kg	reiterated trunks kg	limbs kg	branches kg	leaves kg		type	species	main trunk	reiteration	limb	branch	leaf	TOTAL	% total
SESE	Atlas	255144.9	46020.6	5477.7	13433.2	1101.2		tree	SESE	3569312	213247	53714	230945	17192	4084409	95.3491
SESE	Ballantine	221966.4	7651.6	5922.9	11210.0	1084.8		tree	PSME	135815	0	0	8338	961	145114	3.3876
SESE	Bell	253246.4	5454.3	5792.6	48500.7	1043.4		tree	THSE	31799	0	0	6343	864	39006	0.9105
SESE	Broken Top	130928.9	4805.2	1608.1	5137.4	729.9		tree	ACMA	4444	0	0	925	264	5634	0.1315
SESE	Buena Vista	128833.0	3486.5	0.0	8552.1	518.4		tree	UMCA	2921	0	0	937	273	4131	0.0964
SESE	Demeter	155896.0	11085.6	3204.3	10054.1	768.7		shrub	RUSP	0	0	0	1974	686	2660	0.0620
SESE	Epimetheus	226987.0	12915.7	1797.2	13585.2	1029.4		fern	POMU	0	0	0	0	1271	1271	0.0296
SESE	Iluvatar	349586.6	65003.9	12315.6	13987.0	1461.8		shrub	VAOV	0	0	0	526	26	552	0.0129
SESE	Kronos	134154.1	12204.4	7232.7	5036.1	597.3		shrub	COCO	0	0	0	284	6	289	0.0067
SESE	Pleiades I	182385.2	3735.0	1935.2	10846.6	762.2		fern	POSC	0	0	0	107	89	196	0.0045
SESE	Pleiades II	235838.8	11183.4	4306.0	11306.5	877.7		tree	RHPU	100	0	0	44	18	162	0.0037
SESE	Prometheus	239414.0	25228.9	1612.6	12458.2	1086.0		herb	OXOR	0	0	0	0	112	112	0.0026
SESE	Rhea	143710.4	487.8	730.1	5524.2	691.2		shrub	VAPA	0	0	0	94	4	99	0.0023
SESE	Zeus	243365.7	2885.5	1620.4	19104.7	954.3		tree	PISI	0	0	0	1	0	1	0.0000
SESE	3	1761.3	0.0	0.0	87.6	41.4		tree	CHLA	0	0	0	1	0	1	0.0000
SESE	4	6312.0	356.0	73.5	214.1	43.8		shrub	GASH	0	0	0	0	0	0	0.0000
SESE	5	206.0	0.0	0.0	8.7	2.5		shrub	SACA	0	0	0	0	0	0	0.0000
SESE	6E	18697.4	0.0	0.0	1055.2	66.3				3744390	213247	53714	250519	21767	4283636	
SESE	6W	14651.5	7.7	0.0	626.3	49.6										proportion
SESE	11	614.4	0.0	0.0	28.1	17.0										geophytic
SESE	12	232.1	0.0	0.0	11.2	10.3										
SESE	18	15632.0						SESE	geo	3569312	213247	53714	230945	17192	4084409	1.00
SESE	19	11805.5						SE	epi	0	0	0	0	0	0	
SESE	20	309.5						ME	geo	135815	0	0	8338	961	145114	1.00
SESE	22	25618.3						ME	epi	0	0	0	0	0	0	
SESE	23	463.7						HE	geo	31740	0	0	6332	860	38932	0.99
SESE	25	87.7						HE	epi	59	0	0	12	4	74	
SFRF	30	512.1						MA	geo	4444	0	0	925	264	5634	1.00
								MA	epi	0	0	0	0	0	0	

Marginal sums

INSTEAD: only identifying or measured variables should exist here; use a scripted language to analyze data / calculate summary stats

Denormalized (untidy) data

Data are **denormalized** when observations about different entities are combined. For example, each row in the data table below has site characteristics & species observations:

id	date	site	name	temp	sp1code	sp1height	sp2code	sp2height
1	2017-10-10	1	Taku	23.7	DAPU	4.6	DAMA	4.5
2	2017-09-05	2	Lituya	19.9	DAMA	3.5	DAPU	3.9

site characteristics

species observations

Importantly, a new species observation would require us to add columns (not a row) -- this data table organization is also known as **wide format**

Normalizing (tidying) this data table

To normalize this data table, we want to organize observations about each type of entity in its own table

id	date	site	name	temp	sp1code	sp1height	sp2code	sp2height
1	2017-10-10	1	Taku	23.7	DAPU	4.6	DAMA	4.5
2	2017-09-05	2	Lituya	19.9	DAMA	3.5	DAPU	3.9

Observed entities:

- site characteristics
- plant species

Variables associated with those observations:

- temperature
- height

Normalized (tidy) data

denormalized / untidy / wide format

id	date	site	name	temp	sp1code	sp1height	sp2code	sp2height
1	2017-10-10	1	Taku	23.7	DAPU	4.6	DAMA	4.5
2	2017-09-05	2	Lituya	19.9	DAMA	3.5	DAPU	3.9



normalized / tidy / long format

We now have:

- Separate tables for each type of entity
- Each row represents a single observed entity
- Observations (rows) are all unique

Additionally:

- All values in a column are of the same type
- All columns pertain to the same observed entity
- Each column represents either an identifying variable or a measured variable (no summary stats)

date
species
height

plants

id	date	site	scode	height
1	2017-10-10	1	DAPU	4.6
2	2017-09-05	2	DAMA	3.5
3	2017-10-10	1	DAMA	4.5
4	2017-09-05	2	DAPU	3.9

sites

site	name	temp
1	Taku	23.7
2	Lituya	19.9

name
temperature

Normalized (tidy) data

Our normalized data now meet the guidelines set by Borner et al. 2009:

- Tables are designed to **add rows**, not columns
- Each **column** contains only **one type of information**
- A single piece of **data is recorded only once** & separated information collected at **different scales** into **different tables**

plants

id	date	site	scode	height
1	2017-10-10	1	DAPU	4.6
2	2017-09-05	2	DAMA	3.5
3	2017-10-10	1	DAMA	4.5
4	2017-09-05	2	DAPU	3.9

sites

site	name	temp
1	Taku	23.7
2	Lituya	19.9

date species height

name temperature

Normalized (tidy) data has lots of benefits!

denormalized / untidy / wide format

id	date	site	name	temp	sp1code	sp1height	sp2code	sp2height
1	2017-10-10	1	Taku	23.7	DAPU	4.6	DAMA	4.5
2	2017-09-05	2	Lituya	19.9	DAMA	3.5	DAPU	3.9



normalized / tidy / long format

date
species
height

id	date	site	scode	height
1	2017-10-10	1	DAPU	4.6
2	2017-09-05	2	DAMA	3.5
3	2017-10-10	1	DAMA	4.5
4	2017-09-05	2	DAPU	3.9

plants

More easily filter rows for observations of interest

```
dplyr::filter(data = plant_data, scode == "DAPU")
```

Describe columns more precisely

scode is the spp. identifier, but what exactly is **sp1code**, **sp2code**?

Optimize storage

not repeating data (e.g. date) reduces file size

sites

site	name	temp
1	Taku	23.7
2	Lituya	19.9

name
temperature

Decrease errors from redundant updates

e.g. only need to update site name in table 2

One more look at tidy data

“**TIDY DATA** is a standard way of mapping the meaning of a dataset to its structure.”

—HADLEY WICKHAM

In tidy data:

- each variable forms a column
- each observation forms a row
- each cell is a single measurement

each column a variable

id	name	color
1	floof	gray
2	max	black
3	cat	orange
4	donut	gray
5	merlin	black
6	panda	calico

each row an observation

Wickham, H. (2014). Tidy Data. Journal of Statistical Software 59 (10). DOI: 10.18637/jss.v059.i10

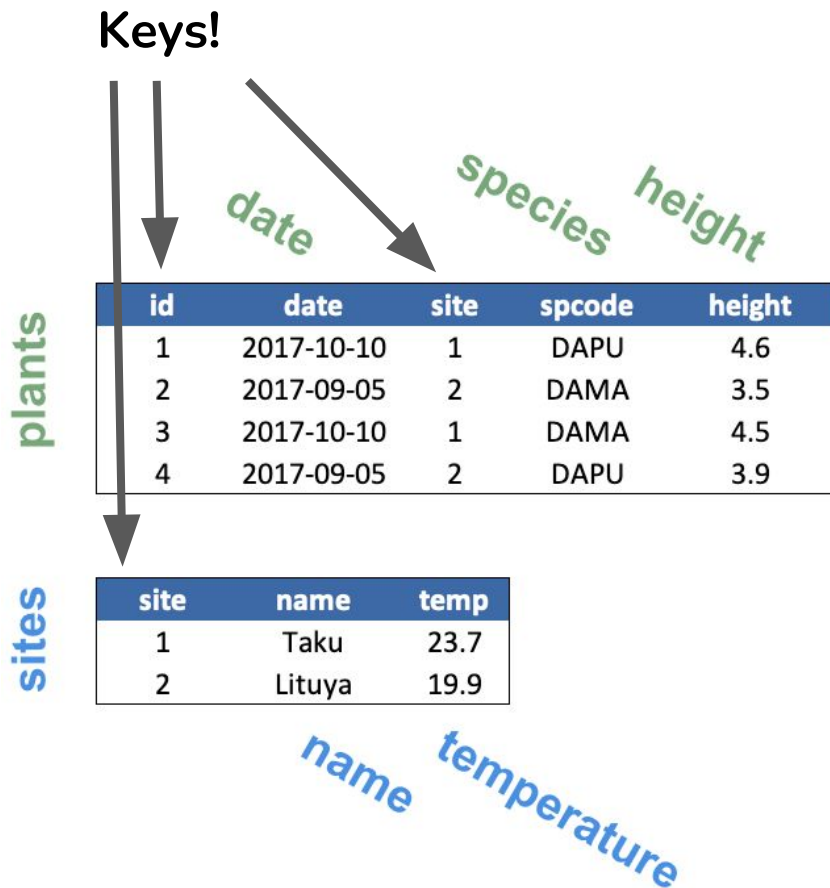
artwork by [@allison_horst](#)

Using normalized data

Two tables?!? Don't we want to analyze all these different measurements together??



(e.g. how will we use site temperature as a predictor variable for species composition?)



Keys allow us to link observations across tables

id uniquely identifies each row in the *plant* table

Primary Key: a unique identifier for each observed entity, one per row

Foreign Key: reference to a primary key in another table (linkage)

site uniquely identifies each row in the *site* table

site references the **primary key** in the *site* table -- this is our linkage

id	date	site	scode	height
1	2017-10-10	1	DAPU	4.6
2	2017-09-05	2	DAMA	3.5
3	2017-10-10	1	DAMA	4.5
4	2017-09-05	2	DAPU	3.9

site	name	temp
1	Taku	23.7
2	Lituya	19.9



primary key



foreign key

entity: plants

id	date	site	sp_code	sp_height
1	2017-10-10	1	DAPU	4.6
2	2017-10-10	1	DAMA	4.5
3	2017-09-05	2	DAMA	3.5
4	2017-09-05	2	DAPU	3.9

entity: sites

site	name	altitude
1	Taku	944
2	Lituya	525



primary key



surrogate
key



natural
key

entity: sites

site	name	altitude
1	Taku	944
2	Lituya	525



compound key

entity: plants

id	date	site	sp_code	sp_height
1	2017-10-10	1	DAPU	4.6
2	2017-10-10	1	DAMA	4.5
3	2017-09-05	2	DAMA	3.5
4	2017-09-05	2	DAPU	3.9

Keys allow us to link observations across tables

Joined the tables by **site**



id	date	site	scode	height	name	temp
1	2017-10-10	1	DAPU	4.6	Taku	23.7
2	2017-09-05	2	DAMA	3.5	Lituya	19.9
3	2017-10-10	1	DAMA	4.5	Taku	23.7
4	2017-09-05	2	DAPU	3.9	Lituya	19.9

Merging data (aka “joins”)

Merging (or joining) two related data tables based on key values is something you’ll probably do often during the data preparation (pre-analysis & visualization) stage. We’ll use these two tables to showcase how different types of joins work:

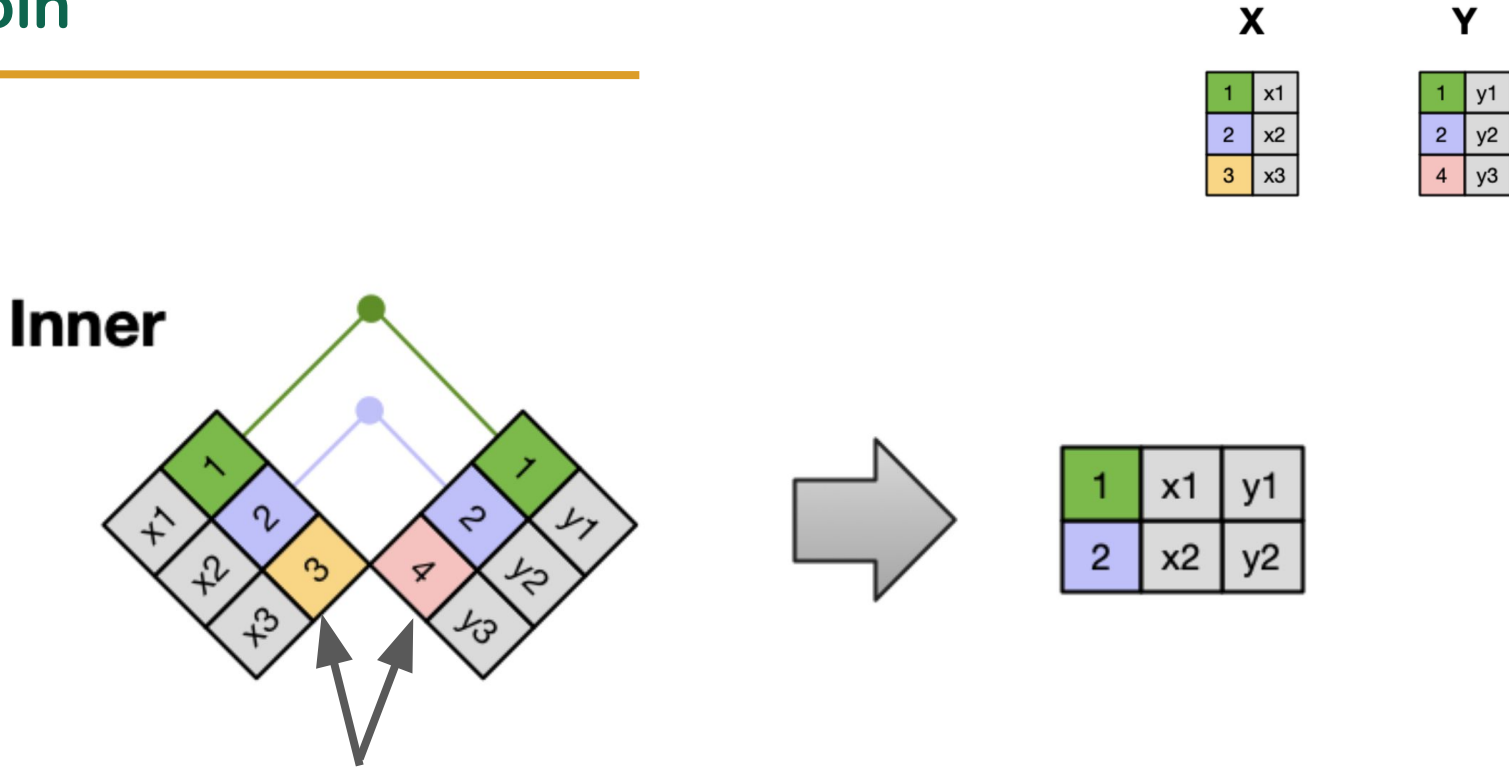
X

1	x1
2	x2
3	x3

Y

1	y1
2	y2
4	y3

Inner join

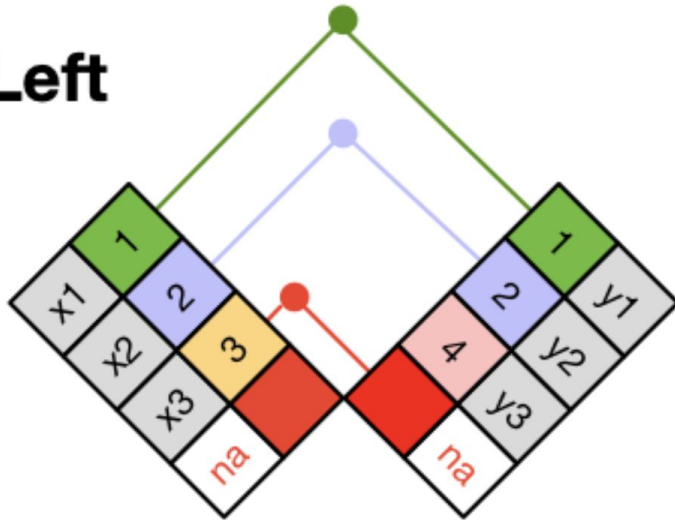


rows 3 (from left table) & 4 (from right table) are dropped because they have no matches

Merge (i.e. keep) the subset of rows that have matches in both the left and right tables

Left join

Left



X

1	x1
2	x2
3	x3

Y

1	y1
2	y2
4	y3

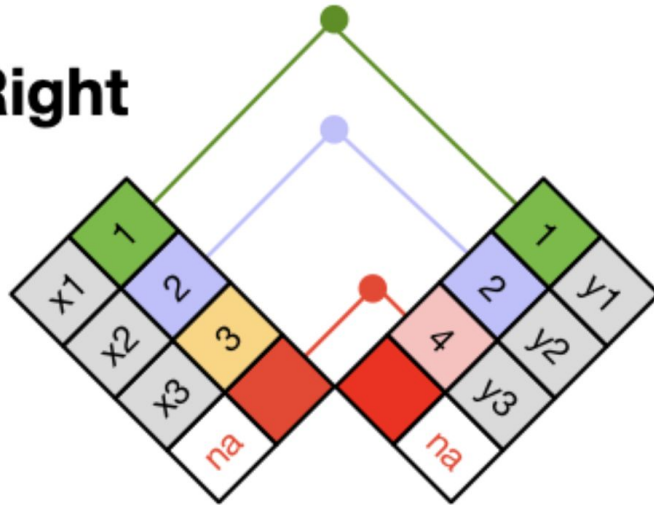
1	x1	y1
2	x2	y2
3	x3	na

rows 1 & 2 (left table) have matches in the right table and are kept;
 row 3 (left table) does not have a match in the right table, so it is kept and assigned an NA value

Take all rows from **left** table and merge on data from matching rows in right table

Right join

Right



X		Y	
1	x1	1	y1
2	x2	2	y2
3	x3	4	y3

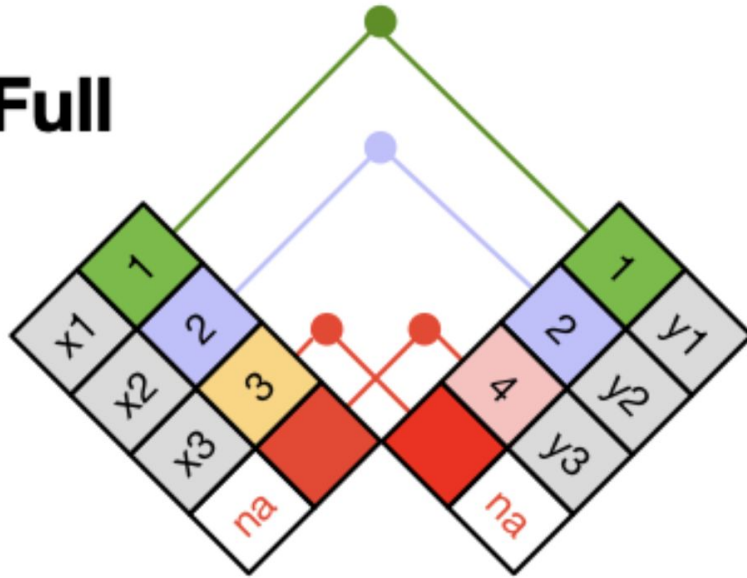
1	x1	y1
2	x2	y2
4	na	y3

rows 1 & 2 (right table) have matches in the left table and are kept;
 row 4 (right table) does not have a match in the left table, so it is kept and assigned an NA value

Take all rows from **right** table and merge on data from matching rows in left table

Full join

Full



	X	Y
1	x1	y1
2	x2	y2
3	x3	

1	x1	y1
2	x2	y2
3	x3	na
4	na	y3

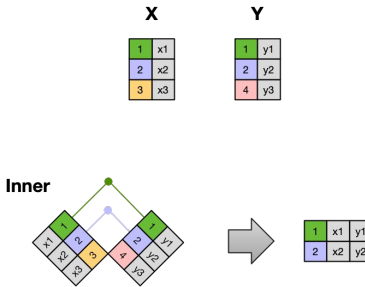
rows 1 & 2 are matched;

row 3 (left table) and row 4 (right table) are kept despite not having matches (assigned the value, NA)

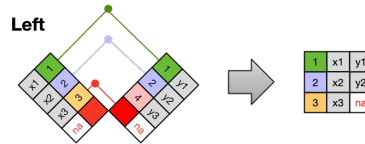
Includes all rows from both tables and adds missing values (NAs) where necessary

Spoiler: {dplyr} has super helpful functions for joining data

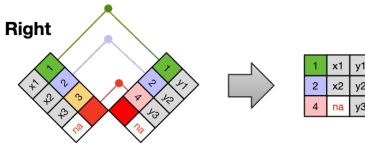
`inner_join(x, y)`



`left_join(x, y)`



`right_join(x, y)`



`full_join(x, y)`

