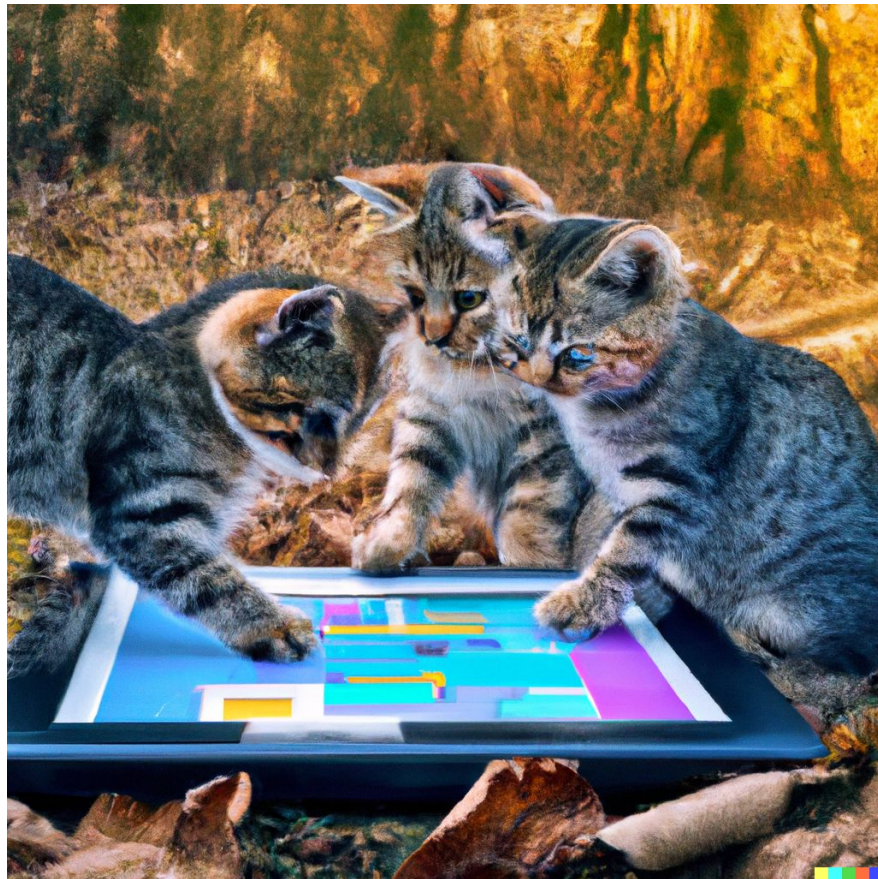


FALL 2023
DSE 12700
VISUAL ANALYTICS

Professor
Madeline Blount
she/her

Week 2



Dall-E2, tabby kittens creating colorful digital charts in a forest, photorealistic style

Ben Fry
MIT, Fathom



MoMA

Membershi

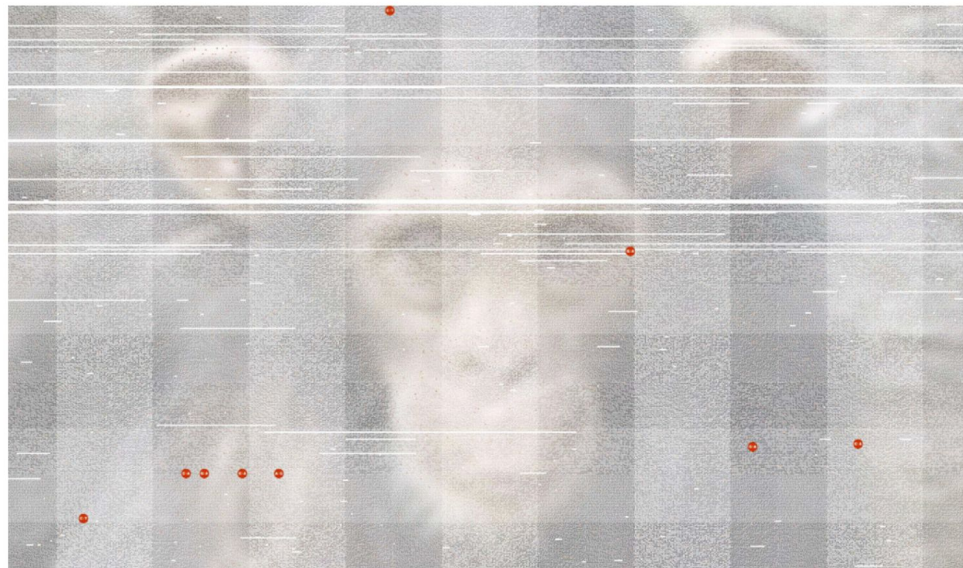
Visit

What's on

Art and artists

Store

Q



Ben Fry
Humans vs. Chimps
2005

<https://www.moma.org/collection/works/110354>

acquire

live or
changing data
sources

parse

modular
parsers for
new data
sources

filter

automation
of tedious
manual
processes
modify filter
in real-time

mine

modify
parameters
of statistical
methods in
real-time

represent

rapid prototyping
and iteration
juxtapose large
amounts of data
try multiple
representations

refine

change
design rules
without
manual
redesign
computation
as its own
“medium”

interact

smooth
transition
between states
to maintain
context
additional
information as
viewpoint
shifts

COMPUTER SCIENCE

MATHEMATICS, STATISTICS,
AND DATA MINING

GRAPHIC DESIGN

INFOVIS
AND HCI

acquire

parse

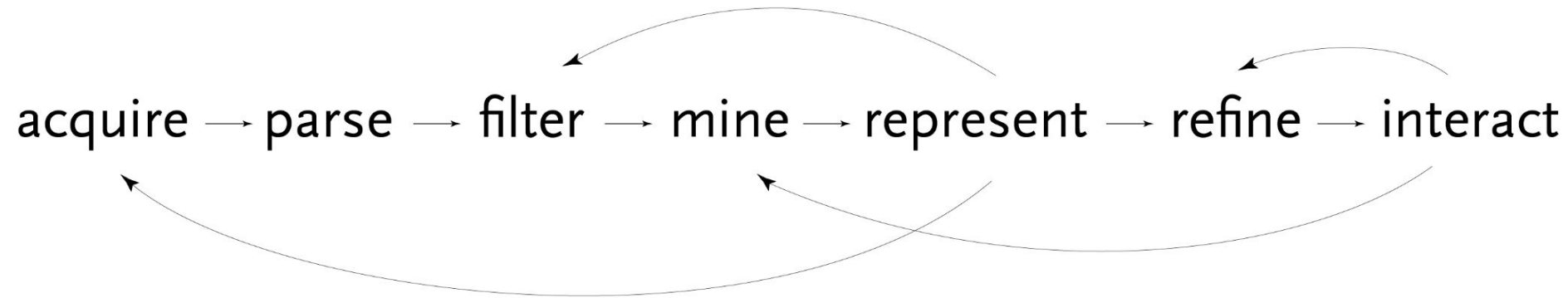
filter

mine

represent

refine

interact

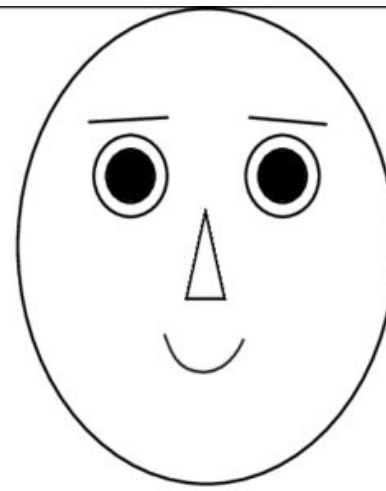


"More data is not implicitly better, and often serves to simply confuse the situation ... A focus on the question helps define what that minimum requirements are." - Fry, Ch. 5

"Knowledge of the audience is essential, for knowing what might be appropriate at each step." - Fry, Ch. 5



New York, NY



Los Angeles, CA

Two examples of multidimensional data on the pace of life and incidence of heart disease using Chernoff faces.

Walking speed = angle of eyebrows

Talking speed = width of mouth

Frequency of watch wearing = height of eyes

Speed of bank transactions = diameter of pupils

Death rate from heart disease = curvature of mouth

(After Levine 1990)

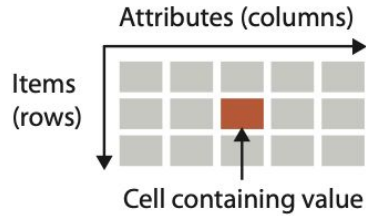
Tamara Munzner
University of British Columbia

Domain-independent vocabulary for data
visualization

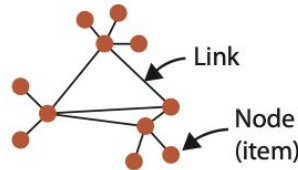


➔ Dataset Types

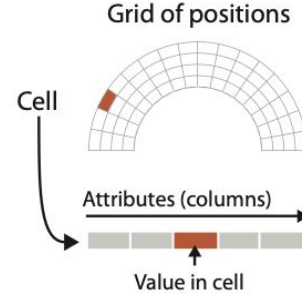
➔ Tables



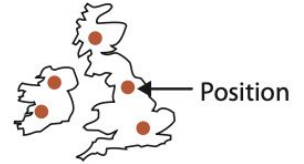
➔ Networks



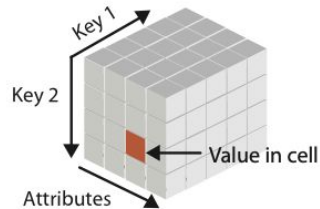
➔ Fields (Continuous)



➔ Geometry (Spatial)



➔ Multidimensional Table



➔ Trees



Figure 2.4. The detailed structure of the four basic dataset types.

A	B	C	S	T	U
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
3	10/14/06	5-Low	Large Box	0.8	10/21/06
6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
32	7/16/07	2-High	Small Pack	0.79	7/17/07
32	7/16/07	2-High	Jumbo Box		7/17/07
32	7/16/07	2-High	Medium Box		7/18/07
32	7/16/07	2-High	Medium Box	0.63	7/18/07
35	10/23/07	4-Not Specified	Wrap Bag	0.52	10/24/07
35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
36	11/3/07	1-Urgent	Small Box	0.55	11/3/07
65	3/18/07	1-Urgent	Small Pack	0.49	3/19/07
66	1/20/05	5-Low	Wrap Bag	0.56	1/20/05
69	5	4-Not Specified	Small Pack	0.44	6/6/05
69	5	4-Not Specified	Wrap Bag	0.6	6/6/05
70	12/18/06	5-Low	Small Box	0.59	12/23/06
70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
96	4/17/05	2-High	Small Box	0.55	4/19/05
97	1/29/06	3-Medium	Small Box	0.38	1/30/06
129	11/19/08	5-Low	Small Box	0.37	11/28/08
130	5/8/08	2-High	Small Box	0.37	5/9/08
130	5/8/08	2-High	Medium Box	0.38	5/10/08
130	5/8/08	2-High	Small Box	0.6	5/11/08
132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
166	9/12/07	2-High	Small Box	0.55	9/14/07
193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

attribute

item

cell

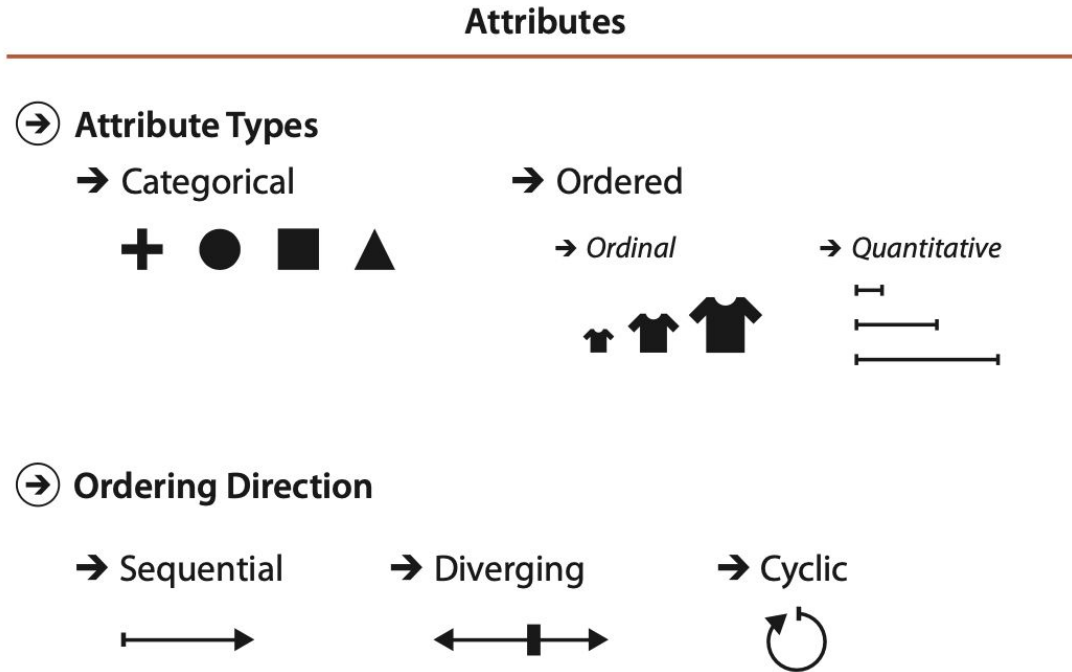


Figure 2.7. Attribute types are categorical, ordinal, or quantitative. The direction of attribute ordering can be sequential, diverging, or cyclic.

2. What: Data Abstraction

A	B	C	S	T	U
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
3	10/14/06	5-Low	Large Box	0.8	10/21/06
6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
32	7/16/07	2-High	Small Pack	0.79	7/17/07
32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
32	7/16/07	2-High	Medium Box	0.6	7/18/07
32	7/16/07	2-High	Medium Box	0.65	7/18/07
35	10/23/07	4-Not Specified	Wrap Bag	0.52	10/24/07
35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
36	11/3/07	1-Urgent	Small Box	0.55	11/3/07
65	3/18/07	1-Urgent	Small Pack	0.49	3/19/07
66	1/20/05	5-Low	Wrap Bag	0.56	1/20/05
69	6/4/05	4-Not Specified	Small Pack	0.44	6/6/05
69	6/4/05	4-Not Specified		0.6	6/6/05
70	12/18/06	5-Low		0.59	12/23/06
70	12/18/06	5-Low		0.82	12/23/06
96	4/17/05	2-High		0.55	4/19/05
97	1/29/06	3-Medium		0.38	1/30/06
129	11/19/08	5-Low		0.37	11/28/08
130	5/8/08	2-High	Small Box	0.37	5/9/08
130	5/8/08	2-High	Medium Box	0.38	5/10/08
130	5/8/08	2-High	Small Box	0.6	5/11/08
132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
166	9/12/07	2-High	Small Box	0.55	9/14/07
193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

quantitative
ordinal
categorical

Figure 2.9. The order table with the attribute columns colored by their type; none of them is a key.

Why?

Actions

Analyze

→ Consume

→ Discover



→ Present



→ Enjoy



→ Produce

→ Annotate



→ Record



→ Derive



Search

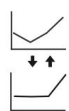
	Target known	Target unknown
Location known	••• Lookup	••• Browse
Location unknown	<•••> Locate	<•••> Explore

Query

→ Identify



→ Compare



→ Summarize



Targets

All Data

→ Trends



→ Outliers



→ Features



Attributes

→ One

→ Distribution



→ Extremes



→ Many

→ Dependency



→ Correlation



→ Similarity



Network Data

→ Topology



→ Paths



Spatial Data

→ Shape



What?

Why?

How?

Figure 3.1. Why people are using vis in terms of actions and targets.

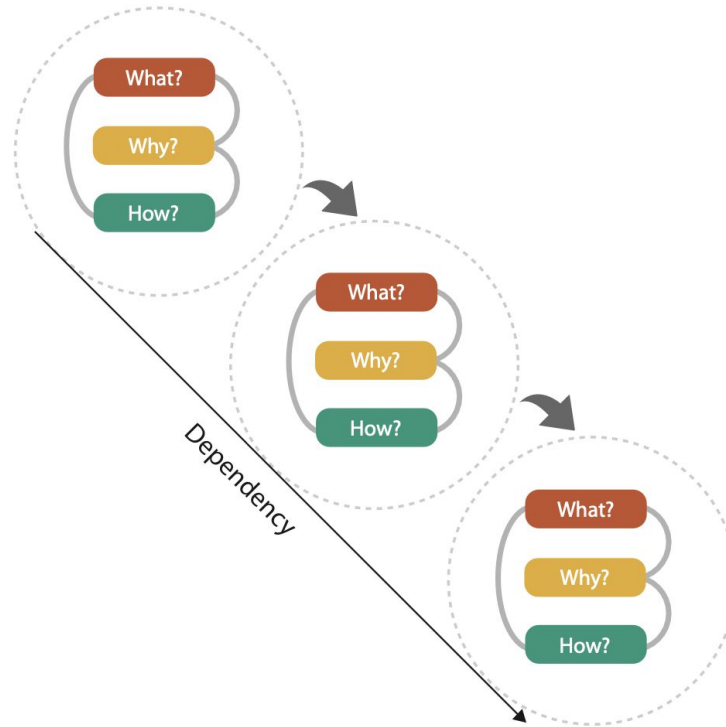


Figure 1.8. Analyzing vis usage as chained sequences of instances, where the output of one instance is the input to another.

Tidy Data

Hadley Wickham
RStudio

2.3. Tidy data

Tidy data is a standard way of mapping the meaning of a dataset to its structure. A dataset is messy or tidy depending on how rows, columns and tables are matched up with observations, variables and types. In *tidy data*:

1. Each variable forms a column.
2. Each observation forms a row.
3. Each type of observational unit forms a table.

religion	<\$10k	\$10–20k	\$20–30k	\$30–40k	\$40–50k	\$50–75k
Agnostic	27	34	60	81	76	137
Atheist	12	27	37	52	35	70
Buddhist	27	21	30	34	33	58
Catholic	418	617	732	670	638	1116
Don't know/refused	15	14	15	11	10	35
Evangelical Prot	575	869	1064	982	881	1486
Hindu	1	9	7	9	11	34
Historically Black Prot	228	244	236	238	197	223
Jehovah's Witness	20	27	24	24	21	30
Jewish	19	19	25	25	30	95

Table 4: The first ten rows of data on income and religion from the Pew Forum. Three columns, \$75–100k, \$100–150k and >150k, have been omitted.

religion	income	freq
Agnostic	<\$10k	27
Agnostic	\$10–20k	34
Agnostic	\$20–30k	60
Agnostic	\$30–40k	81
Agnostic	\$40–50k	76
Agnostic	\$50–75k	137
Agnostic	\$75–100k	122
Agnostic	\$100–150k	109
Agnostic	>150k	84
Agnostic	Don't know/refused	96

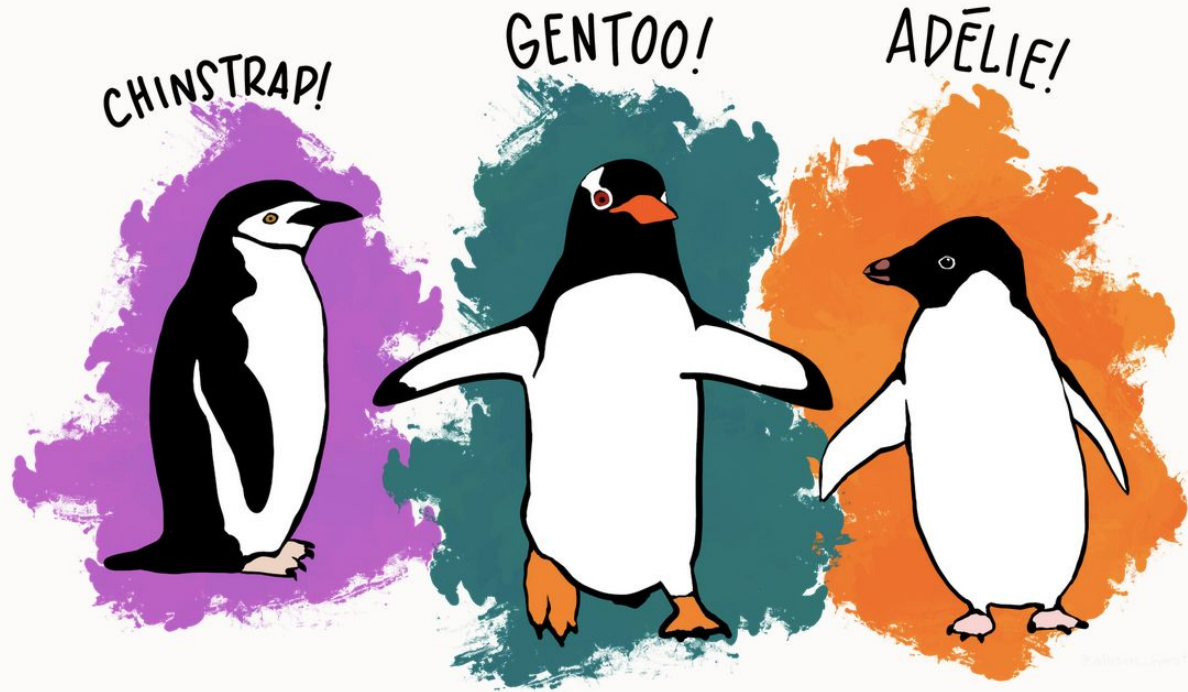
Table 6: The first ten rows of the tidied Pew survey dataset on income and religion. The `column` has been renamed to `income`, and `value` to `freq`.



- 🐍 python
- **P**anel **D**ata, from econometrics data, 2008 (Wes McKinney)
- open source, now run by nonprofit



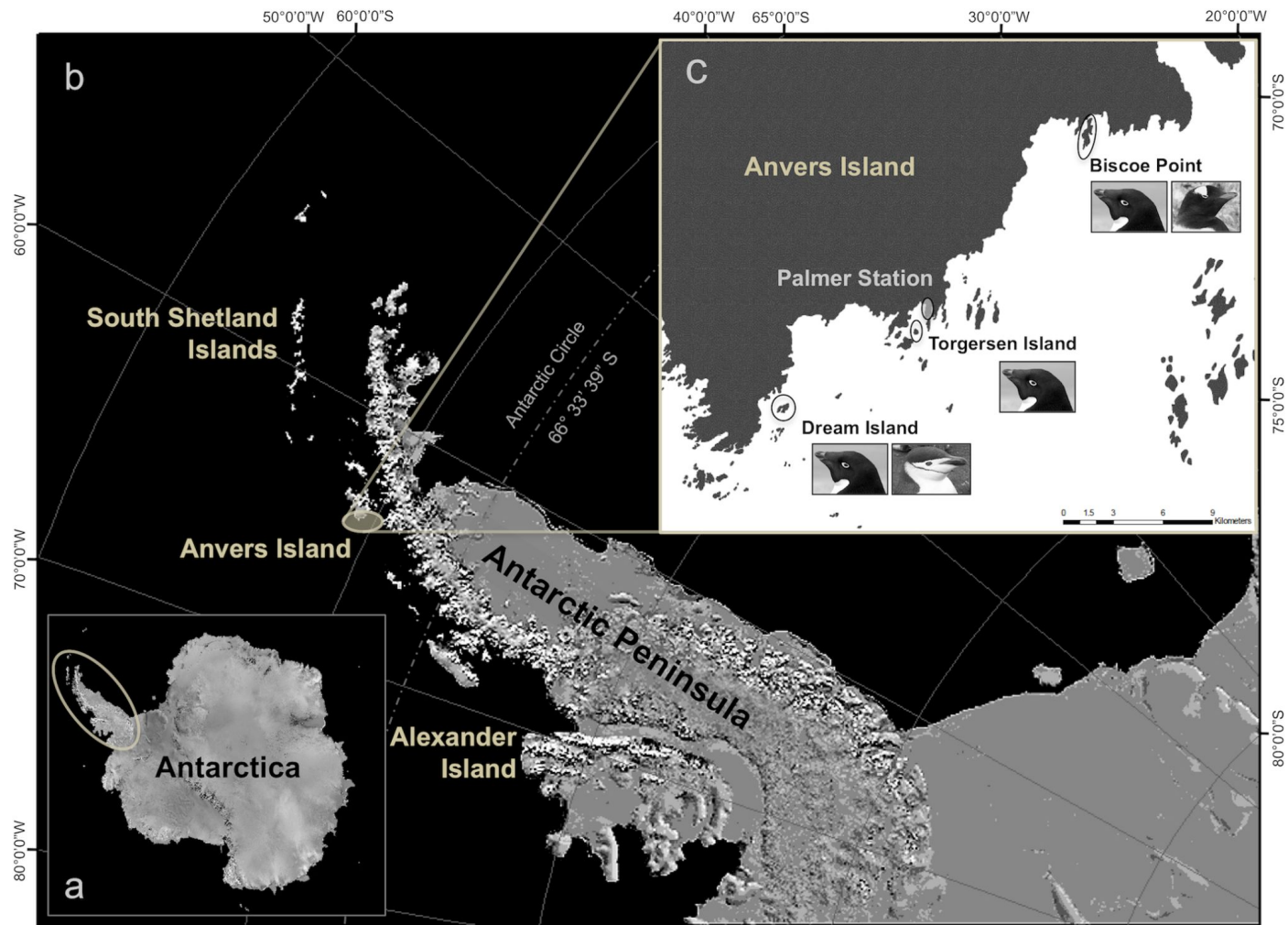
Meet the Palmer penguins





Data were collected from 2007-2009 by **Dr. Kristen Gorman** with the [Palmer Station Long Term Ecological Research Program](#)

Gorman et. al. made their data public - **Allison Horst and Allison Hill** turned into dataset package





Chinstrap Penguins, [Richard Sidley](#)

