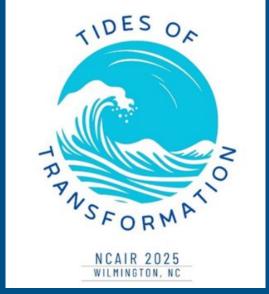
Putting the R in Institutional Research

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Two IR Departments, One Common Tool



Private Institution

Students: 2,000

IR Staff: 2 FTE

Primary Responsibilities: Compliance Reporting,

Survey Research, Analytics



Students: 32,400

IR Staff: ~14 FTE

Primary Responsibilities:

Accreditation, Analytics,

Assessment, Survey

Research, Reporting



Session Goals

- Attendees should be able to:
 - articulate the benefits of using R for Institutional Research work.
 - describe practical examples of how adopting R has improved upon legacy processes for reporting.
 - understand how R can be leveraged in different IR office settings (at both small and large institutions).

Benefits of Using R

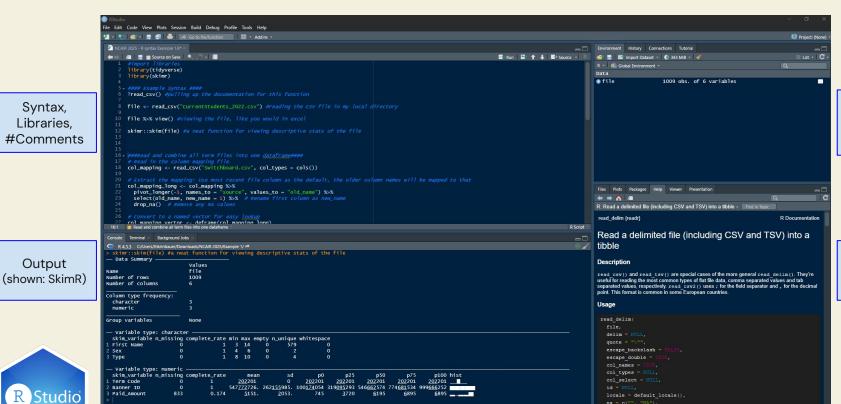
- Open Source Free with a large community
- Reproducibility Readable syntax for consistent data management and reporting
- Flexibility Ability to ingest, or export to, other data formats (e.g., SQL, Excel, SPSS)
- Complexity Easily handle large datasets

Practical examples to demonstrate why we love using R in IR

Ex. 1: Census Reporting & Cohort Tracking

Ex. 2: Survey Codebooks & Reporting

A brief look at R...



View Data. Connections. Values, etc.

View R Docs, Files, Graphs



Syntax,

Libraries,

#Comments

Output

Note: We will not be showing R Syntax during the presentation, but it will be made available after the session

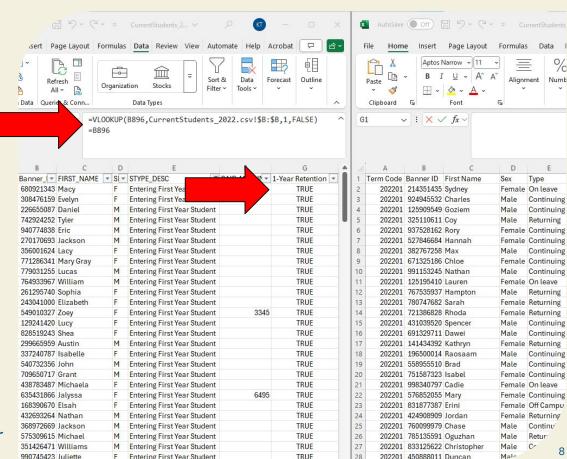
Example 1: Census Enrollment & Cohort Tracking

Ex 1: Legacy process for calculating retention/graduation

Legacy process relied heavily on VLOOKUPs. The process was...

- Inefficient/slow to do calculations
- No documentation difficult to replicate with staff turnover
- Tedious to calculate values across cohorts (1-YR Retention, 4-YR Grad, 6-YR grad, etc.)
- Incompatible with BI Tools

Note: Data included are "dummy data" to illustrate the analysis and insights for this conference session



Ex 1: Modernizing the process

Goal: Create a systemized, efficient approach to calculate enrollment, retention and graduation rates, across multiple cohorts

Challenges with a *messy data* across 20+ term files:

- Different column names (ID vs Banner_ID)
- Different values (Female vs. F)
- Different formatting (\$3,000 vs 3000.00)
- Different missing values (NA, " ", 0)

Table: Fall 2019		
Column	Example	
FILE_TERM	201901	
ID	107XXXXXX	
Sex	Female, Male	
Paid	\$6195.00, NA	

Table: Fall 2020		
Column	Example	
Term	202001	
ID	107XXXXXX	
SEX	F, M	
PAID AMT	\$6,345, \$0	

Table: Fall 2021		
Column	Example	
Term_Code	202101	
Banner_ID	107XXXXXX	
SEX	F, M	
PAID AMOUNT	6495, NA	

Table: Fall 2022		
Column	Example	
Term Code	202201	
Banner ID	107XXXXXX	
Gender	F, M	
PAID AMOUNT	6895, \$0	

Ex 1: Master **Enrollment Table**

Process includes:

- Ingesting data in bulk
- Dynamically renaming columns
- Bind Rows (i.e., Data Union)
- Data Cleaning on existing info
- Creating/Joining new data



Table: Fall 2019	Table: Fall 2020
FILE_TERM	Term
ID	ID
Sex	SEX

Table: Fall 2020		
Term		
ID		
SEX		

Table: Fall 2021	Table: Fall 2022	
Term Code	Term Code	
Banner_ID	Banner ID	
SEX	Gender	





Cohort Term	Student ID	Sex	Student Type	Paid Amount
2019	XXX102	Female	First-Year	\$6,195
2020	XXX102	Female	Continuing	
2021	XXX107	Male	Transfer	\$6,495

Ex 1: Cohort Tracking Output

Process Includes:

- Leveraging master enrollment file
- Dynamically calculating ret/grad rates
 - Lots of data joins!

Benefits:

- Reporting data across cohorts
- O/1 values make it easy for calculating sums and averages
- Import into BI Tool (i.e., PowerBI)



Table: Fall 2019		
FILE_TERM		
ID		
S Type		

Table: Fall 2020		
Term		
ID		
S Type		

Table: Fall 2021	Table: Fall 2022	
Term Code	Term Code	
Banner_ID	Banner ID	
STYPE_DESC	Туре	
_	Samor is	





Cohort Term	Student ID	Туре	1-Yr Retention	2-Yr Retention
2019	XXX119	First-Year	1	0
2020	XXX120	First-Year	1	1
2021	XXX121	Transfer	1	0

- 1. Generate graphs to see data across an entire dataframe
- Create a codebook with variable, label, data type, value labels
 & descriptives into a concise format & export to excel, csv,
 etc.

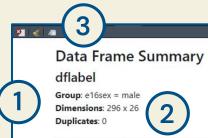
Benefit: Consistent format allows combining codebooks to create a repository that can be built-out in Tableau

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label	data_type	value_labels	n_missing	complete_rate	min	median	max	mean	sd h	hist
Average number of hours of care per week	. numeric		f	6 0.99	4	, 20	168	3 42	51	
Relationship to elder	numeric	1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-lav	7	7 0.99	1	. 2	. 8	3 2.9	2.1	
Elder's gender	numeric	1. male, 2. female	7	7 0.99	1	. 2	2	1.7	0.47	
Elder' age	numeric		17	7 0.98	65	5 79	103	3 79	8.1	_
Elder's dependency	numeric	1. independent, 2. slightly dependent, 3. moderately dependent	£ 7	7 0.99	1	3	4	2.9	0.94	
Do you feel you cope well as caregiver?	numeric	1. never, 2. sometimes, 3. often, 4. always	7	7 0.99	1	. 3	4	3.1	0.58 _	
	Average number of hours of care per week Relationship to elder Elder's gender Elder' age Elder's dependency	Average number of hours of care per week numeric Relationship to elder numeric Elder's gender numeric Elder' age numeric Elder's dependency numeric	Average number of hours of care per week Relationship to elder Elder's gender Elder' age Elder's dependency numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 1. male, 2. female 1. independent, 2. slightly dependent, 3. moderately dependent	Average number of hours of care per week numeric Relationship to elder numeric Elder's gender numeric Elder'age numeric Elder's dependency numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-lav 7 I. male, 2. female 7 I. independent, 2. slightly dependent, 3. moderately depende 7	Average number of hours of care per week Relationship to elder numeric numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 7 0.99 Elder's gender numeric 1. male, 2. female 7 0.99 Elder'age numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99	Average number of hours of care per week Relationship to elder numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 7 0.99 1 Elder's gender numeric 1. male, 2. female 7 0.99 1 Elder' age numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99 1 0.99 1	Average number of hours of care per week numeric Relationship to elder numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 7 0.99 1 2 Elder's gender numeric 1. male, 2. female 7 0.99 1 2 Elder' age numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99 1 3	Average number of hours of care per week numeric 6 0.99 4 20 168 Relationship to elder numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 7 0.99 1 2 8 Elder's gender numeric 1. male, 2. female 7 0.99 1 2 2 Elder' age numeric 17 0.98 65 79 103 Elder's dependency numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99 1 3 4	Average number of hours of care per week numeric 6 0.99 4 20 168 42 Relationship to elder numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son - in-law 7 0.99 1 2 8 2.9 Elder's gender numeric 1. male, 2. female 7 0.99 1 2 2 1.7 Elder'age numeric 17 0.98 65 79 103 79 Elder's dependency numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99 1 3 4 2.9	Average number of hours of care per week numeric 6 0.99 4 20 168 42 51 Relationship to elder numeric 1. spouse/partner, 2. child, 3. sibling, 4. daughter or son -in-law 7 0.99 1 2 8 2.9 2.1 Elder's gender numeric 1. male, 2. female 7 0.99 1 2 2 1.7 0.47 Elder' age numeric 17 0.98 65 79 103 79 8.1 Elder's dependency numeric 1. independent, 2. slightly dependent, 3. moderately depende 7 0.99 1 3 4 2.9 0.94



- Add grouping variables
- 2. Choose variables or the whole df
- Exported to word, excel, pdf

Benefit: quick to see what's going on or little time for a task.



No	Variable	Label	Stats / Values	Freqs (% of Valid)	Graph	Valid	Missing
1	c12hour [numeric]	Average number of hours of care per week	Mean (sd) : 47.3 (55.1) min \leq med \leq max: $4 \leq 20 \leq 168$ IQR (CV) : 40 (1.2)	50 distinct values		296 (100.0%)	0 (0.0%)
2	e15relat [factor]	Relationship to elder	1. spouse/partner 2. child 3. sibling 4. daughter or son -in-law 5. ancle/aunt 6. nephew/niece 7. cousin 8. other, specify	123 (41.6%) 110 (37.2%) 13 (4.4%) 15 (5.1%) 10 (3.4%) 2 (0.7%) 1 (0.3%) 22 (7.4%)		296 (100.0%)	0 (0.0%)



- 1. Customize headers & footnotes
- 2. Two grouping variables
- 3. Descriptives & statistical analyses

Benefit: answers more specific client questions.

		Table o	f Items by Sex 8	Educatio	n					
	Male				Female					
Characteristic	low level of education N = 41 ¹	intermediate level education N = 113 ²	th level of ducation N = 47 ²	p- value ²	low level of education N = 138 ²	intermediate level of education N = 393 ¹	high lev edu N : 3	p- value		
Average number of hours of care per week	41 (44)	33 (47)	39 (55)	0.6	53 (55)	44 (53)	35 (42)	0.033		
Do you find caregiving too demanding?				0.14				0.047		
Never	13 (32%)	30 (27%)	9 (19%)		26 (19%)	73 (19%)	20 (18%)			
Sometimes	20 (49%)	70 (62%)	26 (55%)		81 (59%)	253 (64%)	61 (56%)			
Often	5 (12%)	12 (11%)	10 (21%)		19 (14%)	51 (13%)	25 (23%)			
Always	3 (7.3%)	1 (0.9%)	2 (4.3%)		12 (8.7%)	16 (4.1%)	3 (2.8%)			
¹ Mean (SD); n (%)										
² One-way analysis of means; Pearson's C	hi-squared test									
-Data: sjlabelled::efc										
-Missing data listwise removed										



Perspectives on R Journey

Reflecting on our R Journey

- Increased efficiency in daily workflow. Working smarter, not harder!
- Improved collaboration with colleagues
- Building robust documentation for projects (Syntax & Codebooks)
- Potential for advanced statistical analysis to further our work
- Remembering Roche's Maxim: "Data should be transformed as far upstream as possible, and as far downstream as necessary."

Challenges

Colleagues may not use R and institutions (or other departments)
 may prefer another tool or be tool agnostic

 Training staff on using R - There can be a steep initial learning curve coming from Excel or SPSS (point-and-click)

Replacing legacy processes can be more tedious than anticipated.
 It's usually not ideal to replicate an old process.

Free Resources

- <u>RforIR.com</u>: IR-specific resources from colleagues at Furman University.
 *** Highly Recommended ***
- RStudio: Beginners resources
- R for Data Science: The seminal resource for learning R, written by the creator of R
- R for Excel Users: A useful aid for converting Excel users
- <u>Data Transformation Cheat Sheet</u>: Print it out & Hang it!

Questions?

 presentation, data, & code available on: github.com/kfassett/NCAIR_2025

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Bonus Slide - Conversation Prompts

- 1. Do you have any notable case examples of how you use R at your institution?
- 2. Do you have any concerns or barriers preventing you from using R at your institution?
- 3. How can institutional research pros advocate for improving data processes at your institution?

Bonus Slide - Our Favorite R functions!

- SkimR::skim() Provides descriptives of all variables (missing, mean, std dev, quartiles, histogram)
- Janitor::clean_names() Cleans column names
- Clipr::write_clip() Copies your dataframe to your clipboard
- dbplyr::show_query() Converts syntax into SQL query
- styler:style_file() Cleans script to make code format consistent