

From R to SAS and Back Again

Jasmine Sadler, Keane Cooke, Isabelle Coates

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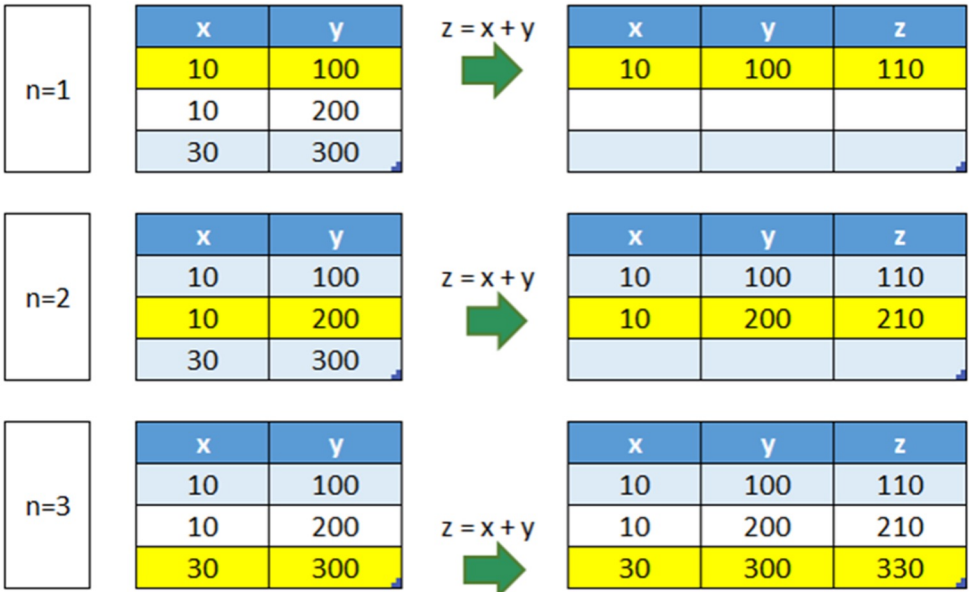


Learning SAS as an R user

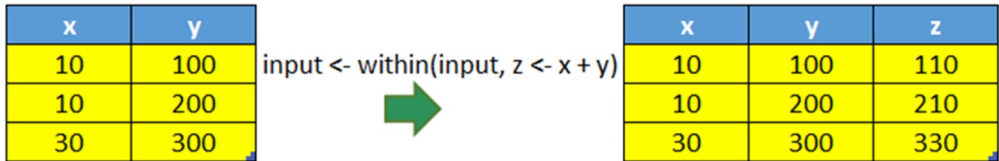


Stage 1: understanding the fundamentals

SAS	R
Procedural Language	Functional Language with support for object-oriented programming
Compiles and then executes	No compilation just execution
Data step works by looping over a dataset row by row	R functions are vectorized
Well documented	Variation in documentation standards
Technical support offered	Community based support



data in memory



data in memory



Stage 2: Basic functionality and syntax

Syntax:

- R functions are often more intuitive and show more similarity to other languages than SAS functions



The MEANS Procedure

Variable	N	Mean	Std Dev	Minimum	Maximum
Weight	9	70.4444444	3.1169340	65.0000000	73.5000000
Height	9	154.8888889	17.2514090	135.0000000	190.0000000



```
> summary(bmi)
```

Weight		Height	
Min.	:65.00	Min.	:135.0
1st Qu.:	69.00	1st Qu.:	140.0
Median	:71.50	Median	:155.0
Mean	:70.44	Mean	:154.9
3rd Qu.:	73.00	3rd Qu.:	164.0
Max.	:73.50	Max.	:190.0

Functionalities:

- SAS shows less flexibility to achieve certain outcomes
- R's extensive packages give multiple ways to achieve an outcome



Stage 3: Linking steps for data transformation

In R, the package tidyverse is the go-to for data transformation

- SAS programmers find the tidyverse syntax more familiar to SAS's proc steps

sex	wt_lb	ht_cm
M	145.2	140
F	158.4	145
M	161.7	160
M	160.6	190
F	151.8	155



1. Keep
2. Data step
3. Drop
4. Proc sort



sex	weight	unit
M	66	kg
F	72	kg
M	73.5	kg
M	73	kg
F	69	kg

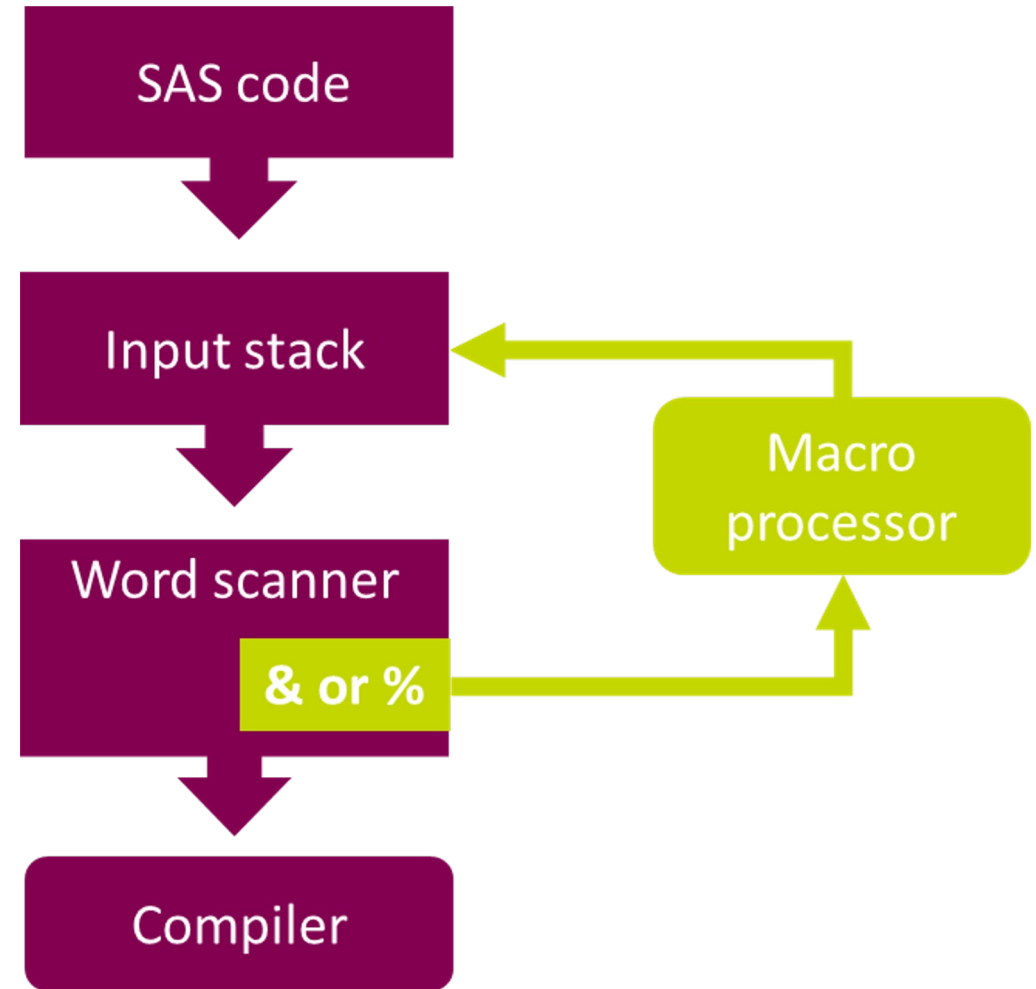


1. Mutate
2. Select
3. Arrange



Stage 4: Applying learnings to new contexts

SAS Macro	R Function
Macro is compiled differently to normal SAS code	R functions are executed like normal R code
Has different syntax to base SAS	Same syntax as R
SAS macro variables are just character strings	R functions arguments can take any object type



R for Statistical Programming



General benefits to using R in Pharma

1. Open source and free
2. Updated frequently
3. Validation process in place



Opportunities – Statistical Capabilities

1. Includes more recent cutting-edge statistical techniques than SAS
2. Multiple packages and methods available to achieve the same type of analysis



Packages capable of fighting linear models



Opportunities – The Tidyverse

- The Tidyverse – Perhaps the greatest strength of R
- A set of tools for data transformation and visualisation
- Focused on the tidy data principles
 - Each variable must have its own column
 - Each observation must have its own row
 - Each value must have its own cell



country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	666	20095360
Brazil	1999	3737	172006362
Brazil	2000	8488	174504898
China	1999	21258	1272015272
China	2000	21676	128012583

variables

country	year	cases	population
Afghanistan	1999	745	19987071
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observations

country	year	cases	population
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values



Tidy Data & CDISC Standards

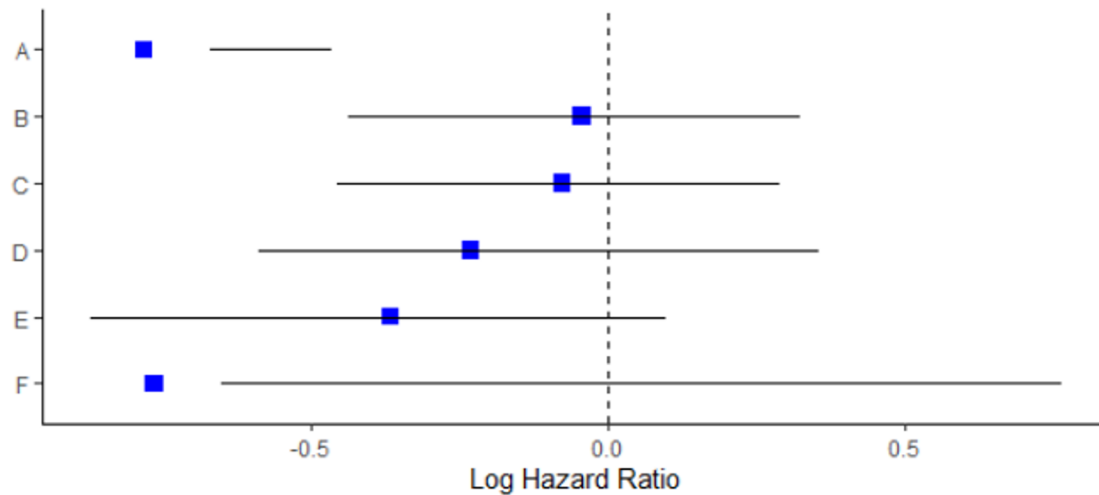
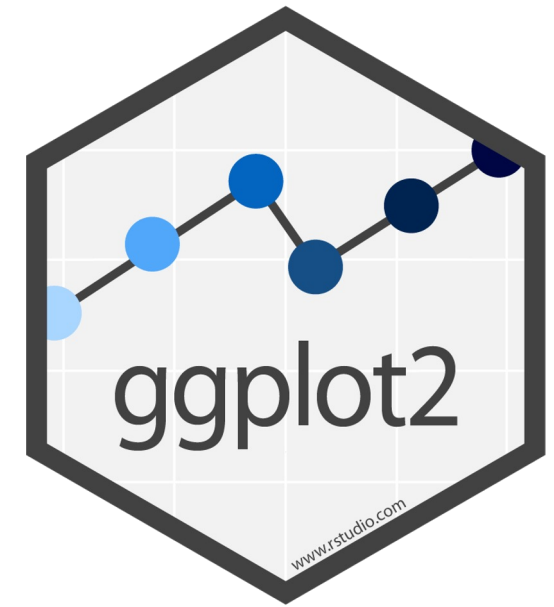
- The tidy data principles are compatible with CDISC standards

Row	STUDYID	DOMAIN	USUBJID	SUBJID	RFSTDTC	RFENDTC	RFXSTDTC	RFXENDTC
1	ABC123	DM	ABC12301001	01001	2006-01-12	2006-02-10	2006-01-12	2006-03-10
2	ABC123	DM	ABC12301002	01002	2006-01-15	2006-02-28	2006-01-15	2006-02-28
3	ABC123	DM	ABC12301003	01003	2006-01-16	2006-03-19	2006-01-16	2006-03-19
4	ABC123	DM	ABC12301004	01004				
5	ABC123	DM	ABC12302001	02001	2006-02-02	2006-03-31	2006-02-02	2006-03-31
6	ABC123	DM	ABC12302002	02002	2006-02-03	2006-04-05	2006-02-03	2006-04-05



Opportunities – Ggplot2

- Aside from SDTMs and ADaMs, the real ‘endgame’ of statistical programming is generating TLFs
- This is where R could shine – ggplot allows a user to create publication-ready plots with ease and flexibility



```
library(tidyverse)

##Create forest plot
ggplot(data=df, aes(y = fct_rev(model))) +
  theme_classic() +
  geom_point(aes(x=log_est), shape=15, size=3, colour='blue') +
  geom_linerange(aes(xmin=log_conf_l, xmax=log_conf_h)) +
  geom_vline(xintercept = 0, linetype="dashed") +
  labs(x="Log Hazard Ratio", y="")
```



Opportunities – Harnessing Shiny



- Use of R opens the door to leverage powerful tools such as shiny
- Shiny allows a user to create interactive web apps
- Potential use cases:
 - ❑ Interactive exploratory data analysis
 - ❑ Automation
 - ❑ More possibilities are likely to arise in the future!



Exploring R in Stat Programming



Exploratory Use-Cases

- Automation - generating specifications using R
- Fully automated process - calling a single function

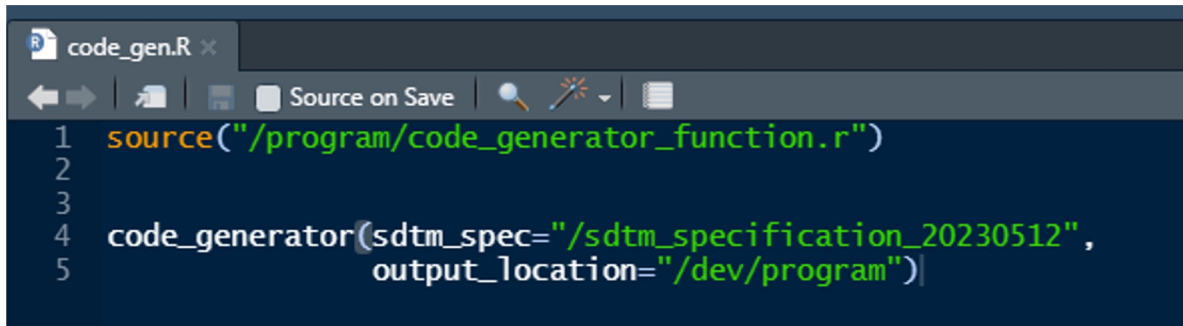
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
	Source Library	Source Dataset	Source Variable	SDTM Domain	SDTM Dataset	SDTM Variable Name	Variable Label	Variable Length	Seq	Key	Origin	Format / Controlled Terminology	Variable Type	Significant Digits	Core
1	RAW	AE	STUDY	AE	AE	STUDYID	Study Identifier	21	1	1	Protocol		C		Req
2	RAW	EVTLOG	STUDY	AE	AE	STUDYID	Study Identifier	21	1	1	Protocol		C		Req
3	RAW	AE		AE	AE	DOMAIN	Domain Abbreviation	8	2		Derived	(DOMAIN)	C		Req
4	RAW	EVTLOG		AE	AE	DOMAIN	Domain Abbreviation	8	2		Derived	(DOMAIN)	C		Req
5	RAW	AE	STUDYSUBJECT	AE	AE	USUBJID	Unique Subject Identifier	30	3	2	Derived		C		Req
6	RAW	EVTLOG	STUDYSUBJECT	AE	AE	USUBJID	Unique Subject Identifier	30	3	2	Derived		C		Req
7				AE	AE	AESQ	Sequence Number	8	4		Derived		N		Req
8				AE	AE	AESQ	Sequence Number	8	4		Derived		N		Req
9				AE	AE	AESQ	Sequence Number	8	4		Derived		N		Req
10	RAW	EVTLOG	LINEMODULE_O	AE	AE	AEREFID	Reference ID	40	6		Derived		C		Perm
11	RAW	AE		AE	AE	AESPID	Sponsor-Defined Identifier	40	7		CRF		C		Perm
12	RAW	EVTLOG	EVNO	AE	AE	AESPID	Sponsor-Defined Identifier	40	7		CRF		C		Perm
13				AE	AE	AESPID	Sponsor-Defined Identifier	40	7		CRF		C		Perm

```
spec_creation.R* x
Source on Save
1 source("/program/specification_creation_function.r")
2
3 specification_creation(standards_location="/standards",
4                       output_location="/output",
5                       output_file=gsub(" ", "-", paste("sdm_specification", format(Sys.Date(), "%Y%m%d"), ".xlsx")),
6                       study_file="/study_file.xlsx",
7                       previous_spec_location = NA,
8                       previous_spec_file = NA)
9
10
```



Exploratory Use Cases

- Automation - generating domain code using R
- Fully automated process - calling a single function



```
code_gen.R x
Source on Save
1 source("/program/code_generator_function.r")
2
3
4 code_generator(sdtm_spec="/sdtm_specification_20230512",
5               output_location="/dev/program")
```



	ae	SAS_Pgm	development	10.59 KB
	ce	SAS_Pgm	development	12.68 KB
	cm	SAS_Pgm	development	20.19 KB
	co	SAS_Pgm	development	10.27 KB



Advantages and disadvantages of switching languages



Advantages/Disadvantages of switching between languages

- Structure/Syntax/processing
 - Javascript, C++, Python syntax
 - SQL, MySQL syntax
- Compilation and Function logic
 - SAS step and macro compilation

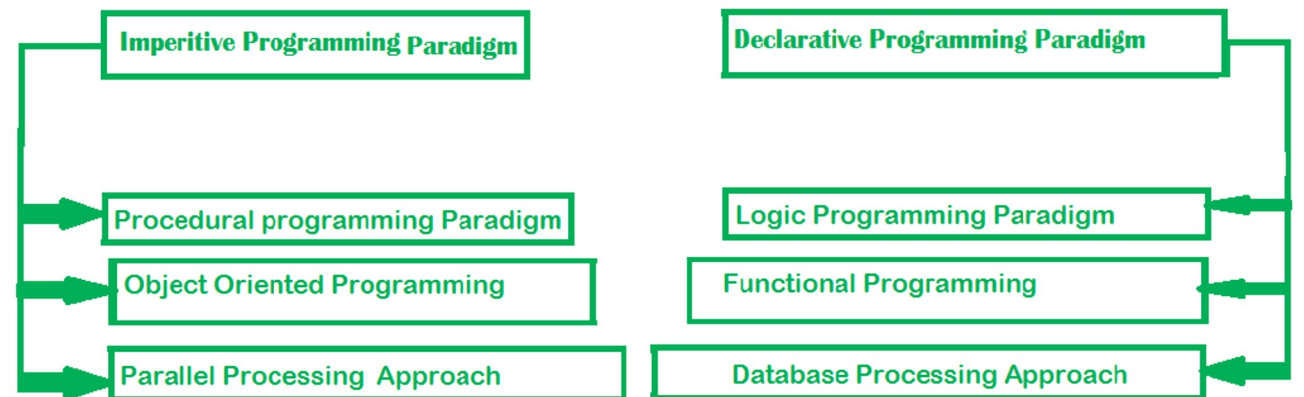


Advantages/Disadvantages of switching between languages

- Paradigms
 - No universal language
- Language specialisation
 - Python
 - Javascript



Programming Paradigms

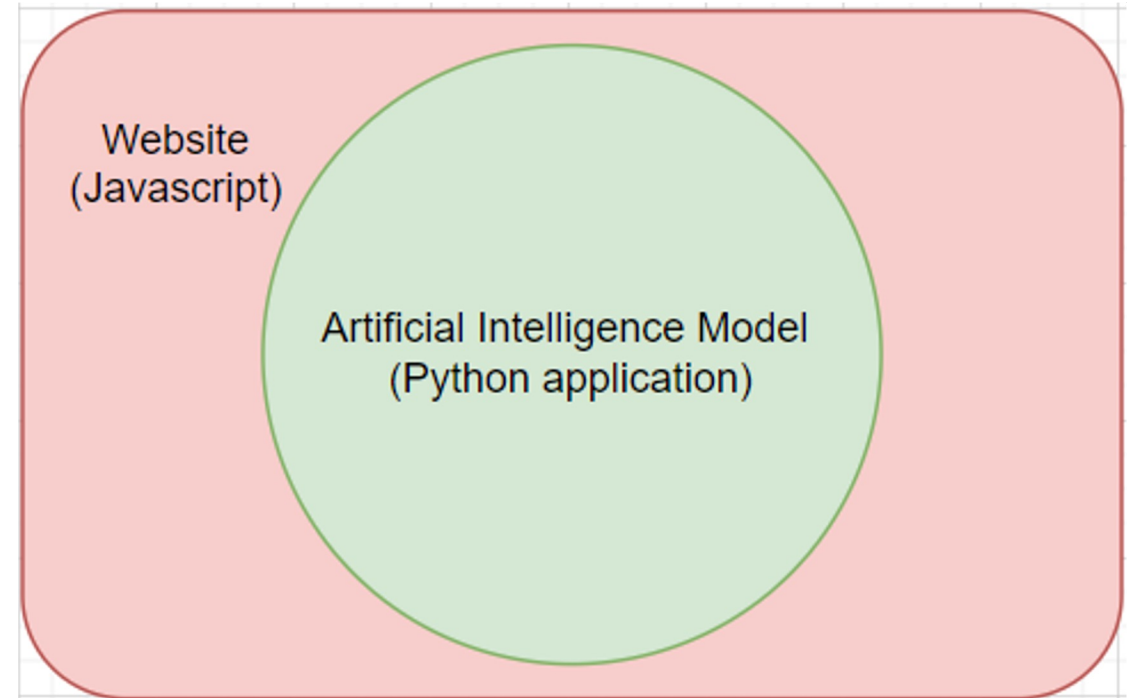


Recommendations of switching languages



Recommendations of switching languages

- Recommendations
 - Switching between languages
 - Exploiting languages for their specialisation.



Conclusion



Acknowledgements

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