### R PACKAGES FOR SDG 6.2 DATA

Development and MSc Data Science project

Linda Karani

Lars Schöbitz

#### **AGENDA**

- 1. R Package Development
- 2. Master Thesis Project
  - Fit models for estimates
    - Prepare primary indicators estimates based on JMP estimation rules
    - Prepare secondary indicator service levels
  - Extract r squared and p-values from models
    - Identify those with low values < 0.20 and high values > 0.80
    - Plots exploring the OLS regression using two countries with low and high rsq values
  - Plots showing the model fit coefficients(r.squared and p value)
  - Fit alternative models and assess goodness of fit
  - Next steps
- 3. Questions for Discussion

### R PACKAGE DEVELOPMENT

### R DATA PACKAGE - BENEFITS

- Data accessible as a single table for any analysis tool
- Data can be imported to R using one command
- Public website with detailed documentation \_ e.g. washdata R Package https://katilingban.io/washdata/index.html

### R DATA PACKAGE - SANITATION

- Data in long format (19,528 rows)
- 9 variables

iso3	source	type	year	var_short	var_long	residence	san_service_
AFG	MICS	Survey	2003	s_imp_u	Improved	urban	user interfac
AFG	NRVS	Survey with microdata	2005	s_imp_u	Improved	urban	user interfac
AFG	NVRA	Survey with microdata	2008	s_imp_u	Improved	urban	user interfac
AFG	MICS	Survey with microdata	2011	s_imp_u	Improved	urban	user interfac

### R DATA PACKAGE - NEW VARIABLES

- residence: urban/rural/national
- san\_service\_chain (Sanitation Service Chain):

san_service_chain	n
open defecation	2774
sharing	1553
user interface	12663
containment	195
emptying	1356
transport	10
FS treatment	85
WW treatment	921
NA	3

#### R DATA PACKAGE - USE CASES

- 1. Using JMP methods to reproduce estimates and apply different models (Linda Karani MSc Data Science)
- 2. Writing an R Package with a function to produce estimates (and a function to produce service ladder plots)

#### R DATA PACKAGE - USE CASES

- 1. Using JMP methods to reproduce estimates and apply different models (Linda Karani MSc Data Science)
- 2. Writing an R Package with a function to produce estimates (and a function to produce service ladder plots)

```
1 estimate(iso3 = "AFG",  # default: all iso3 codes
2  year = 2010:2030,  # Single year or range of years
3  var_short = NULL,  # default: all variables (NULL)
4  residence = "national") # default: national
```

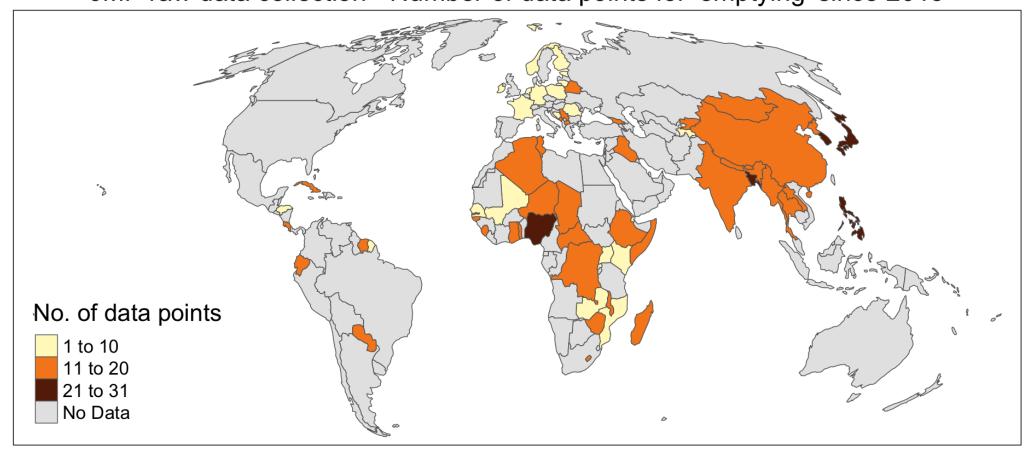
### R DATA PACKAGE - USE CASES

- 1. Using JMP methods to reproduce estimates and apply different models (Linda Karani MSc Data Science)
- 2. Writing an R Package with a function to produce estimates (and a function to produce service ladder plots)

```
1 estimate(iso3 = "AFG",  # default: all iso3 codes
2    year = 2010:2030,  # Single year or range of years
3    var_short = NULL,  # default: all variables (NULL)
4    residence = "national") # default: national
```

3. Great potential for unforeseen use cases enabled by making the data readily accessible (research, teaching, joining with other data, etc.)

#### JMP raw data collection - Number of data points for 'emptying' since 2015



JMP raw data collection: Number of data points for 'emptying' since 2015

country	n
Philippines	62
Nigeria	48
Bangladesh	40
Japan	40
South Korea	32
Ethiopia	20
Niger	20
Belarus	16
China	16
Congo - Kinshasa	16

#### R DATA PACKAGE - WHAT'S NEXT?

- Submission of proposal for further development to ORD (Open Research Data)
  Programme of ETH Domain (15k in-kind + 15k ETH Board), due: 12th December
- Submission of a proposal to Colorado WASH Symposium (focus on Linda's work),
   due: 25th November

### MASTER THESIS PROJECT

#### MASTERS THESIS OUTLINE

#### **OBJECTIVES**

- Generate sanitation estimates for rural and urban from raw packaged data using documented methods
  - only the primary indicators and derived secondary service levels:
    - basic sanitation service
    - limited sanitation service
    - unimproved sanitation service
    - no sanitation service
- Fit different statistical models to assess goodness of fit

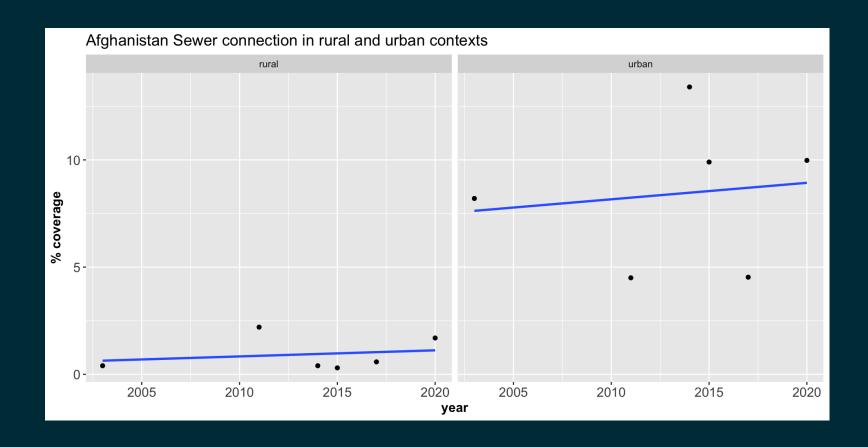
#### **STATUS**

 Managed to derive estimates for the primary indicators and derived secondary service levels

# PLOT LINEAR FIT FOR DIFFERENT COUNTRIES

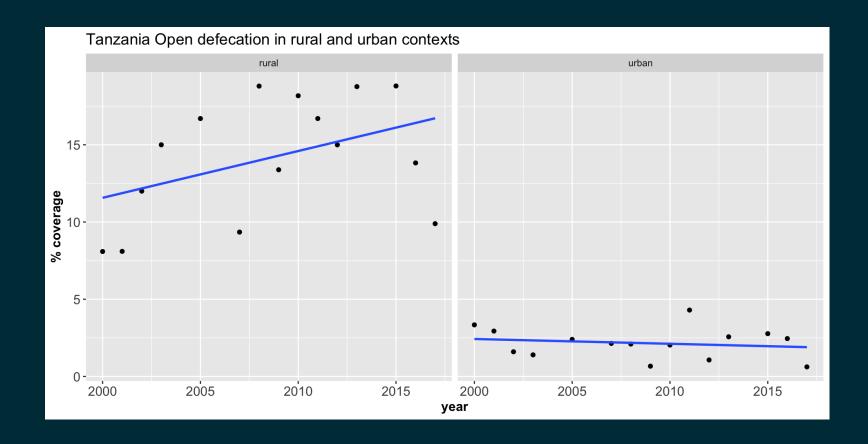
### OLS REGRESSION: R SQUARED < 0.20

### AFGHANISTAN SEWER CONNECTION



- Afghanistanr.sq rural =0.0426018
- Afghanistanr.sq urban =0.0170736

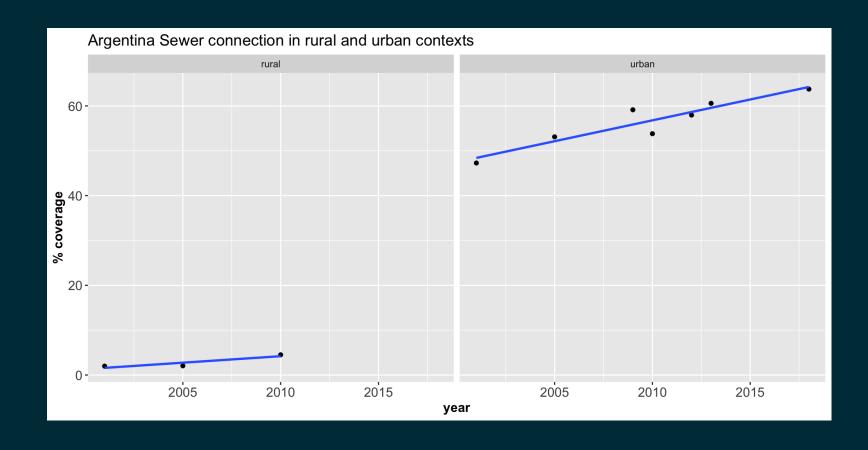
### TANZANIA OPEN DEFECATION



- Tanzania r.sq rural = 0.1824534
- Tanzania r.squrban =0.0296412

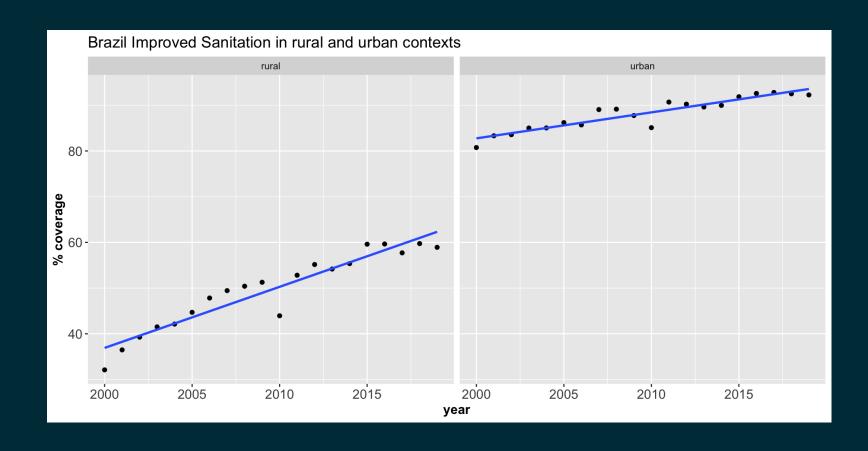
### OLS REGRESSION: R SQUARED > 0.80

### ARGENTINA SEWER CONNECTION OLS REGRESSION



- Argentina r.sq rural = 0.82
- Argentina r.sq urban = 0.87

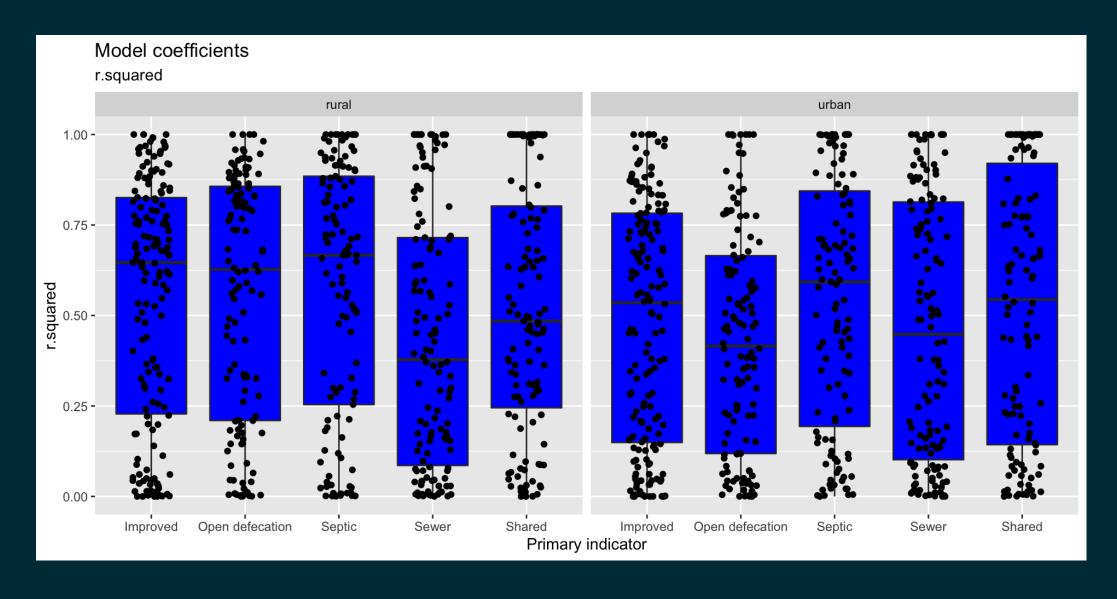
### BRAZIL IMPROVED SANITATION OLS REGRESSION



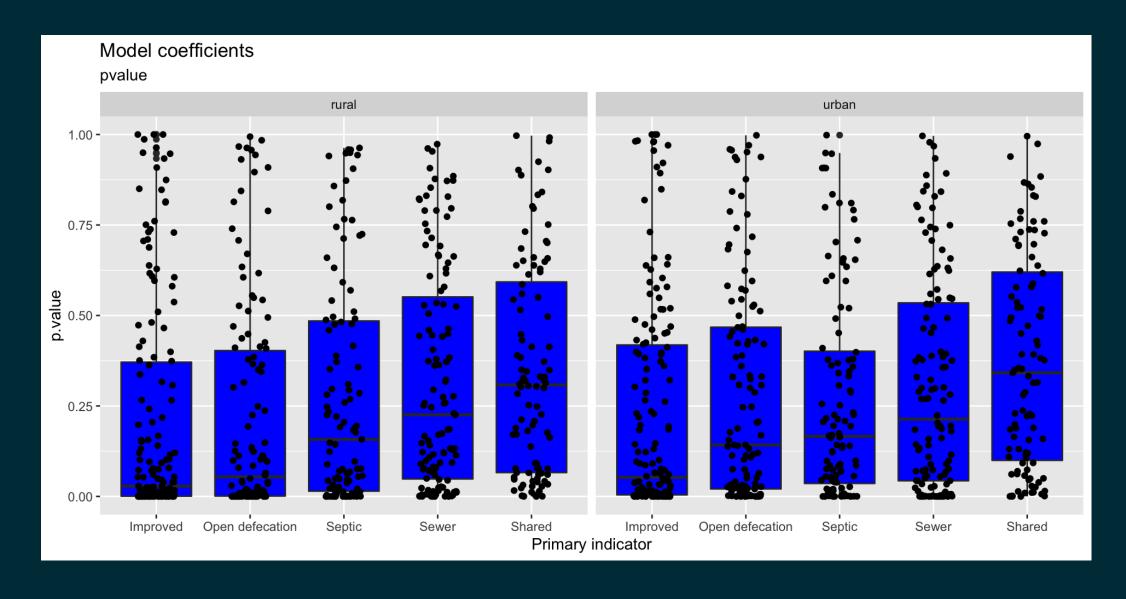
- Brazil r.sqrural = 0.90
- Brazil r.sq urban = 0.87

## PLOTTING MODEL COEFFICIENTS

### **MODEL FIT: R SQUARED**



### MODEL FIT: P VALUE

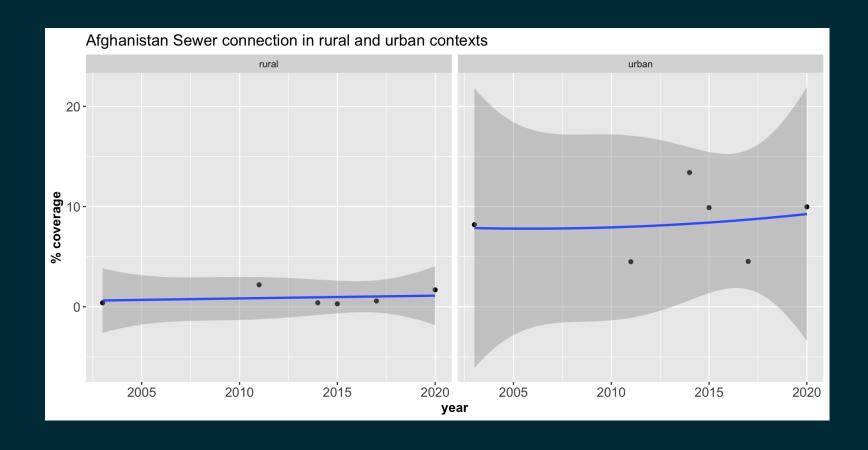


### ALTERNATIVE MODELS

• Fitting different models to the data to compare goodness of fit

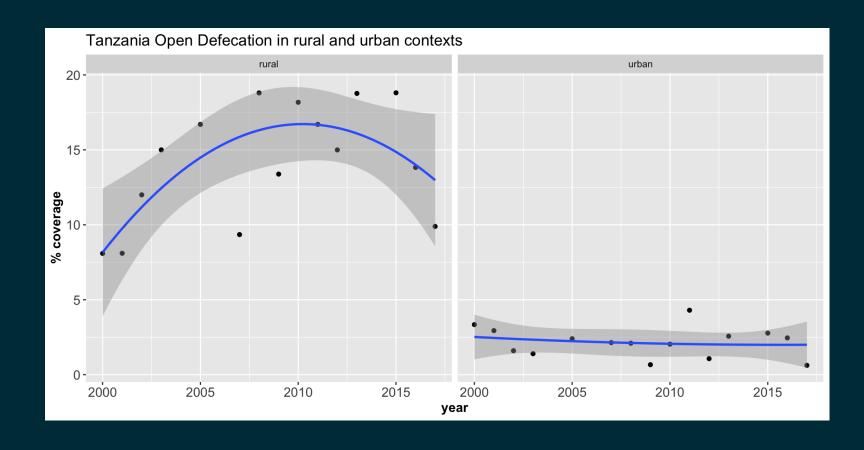
# PLOT USING A 4TH ORDER POLYNOMIAL

### AFGHANISTAN SEWER CONNECTION



- Afghanistanr.sq rural =0.043
- Afghanistanr.sq urban =0.022

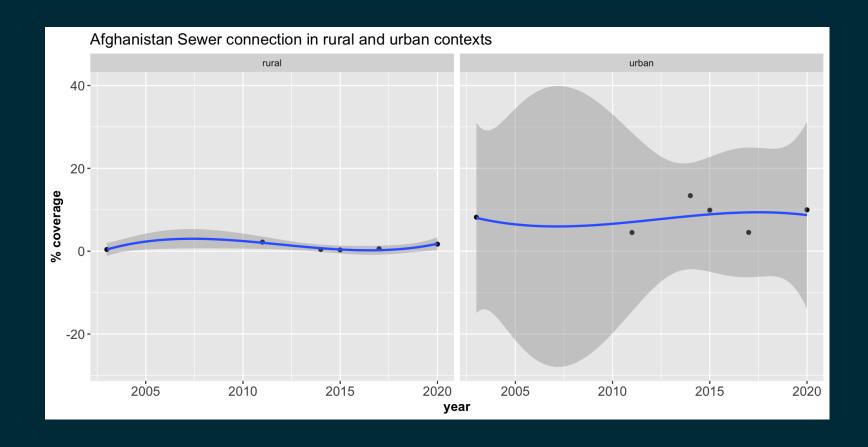
### TANZANIA OPEN DEFECATION



- Tanzania r.sqrural = 0.487
- Tanzania r.sq urban = 0.033

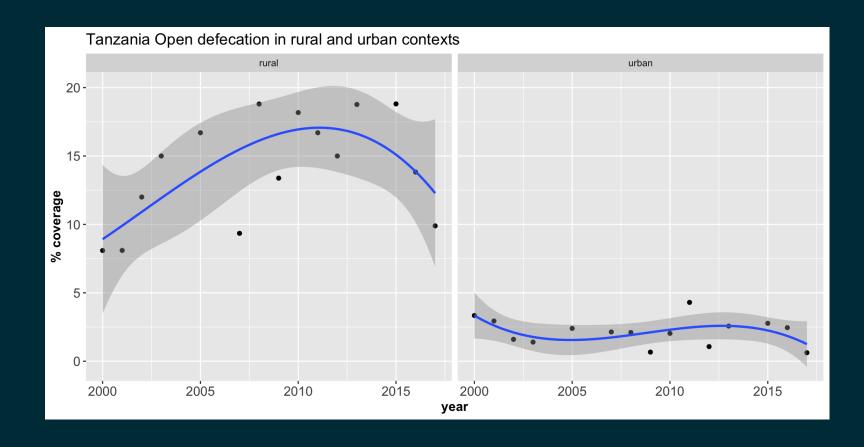
## PLOT USING SPLINES

### AFGHANISTAN SEWER CONNECTION



- Afghanistanr.sq rural =1.00
- Afghanistanr.sq urban =1.00

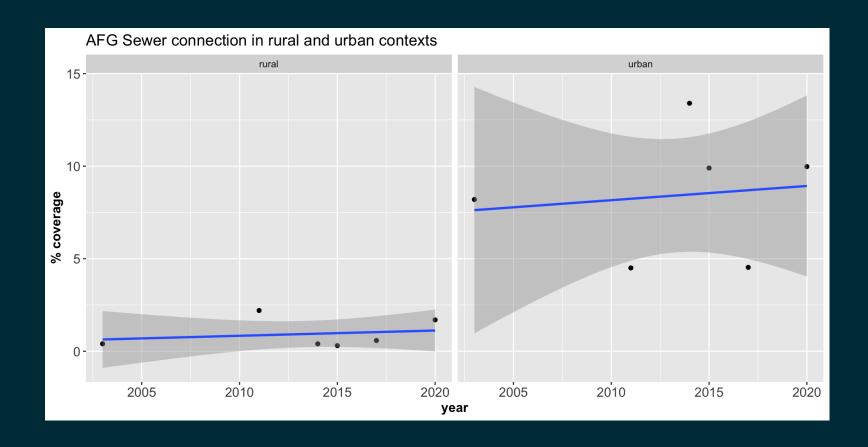
### TANZANIA OPEN DEFECATION



- Tanzania r.sq rural = 0.65
- Tanzania r.sq urban = 0.37

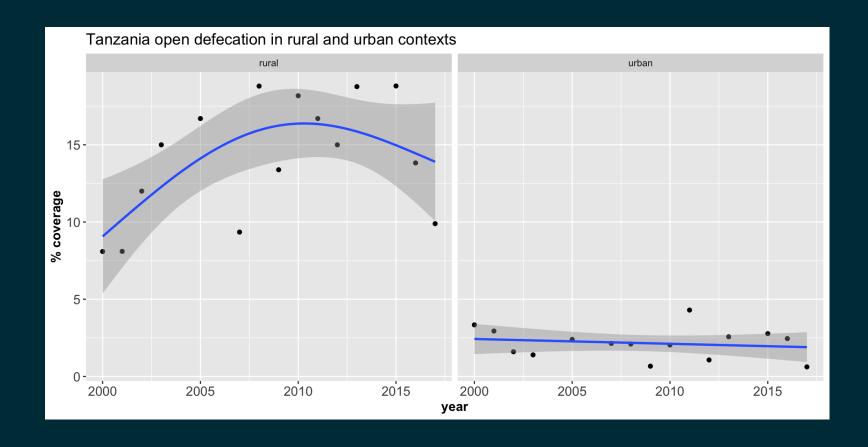
# PLOT USING GAM

### AFGHANISTAN SEWER CONNECTION



- AfghanistanBIC rural =18.5
- AfghanistanBIC urban =36.1

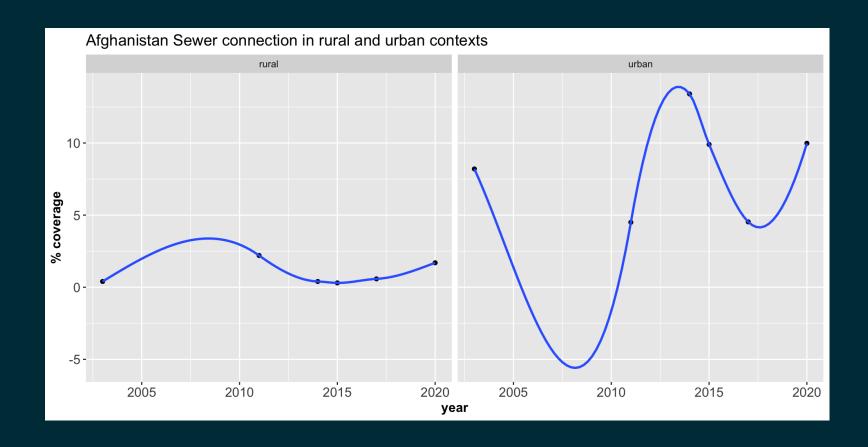
### TANZANIA OPEN DEFECATION



- Tanzania BIC rural = 87.6
- Tanzania BIC urban = 49.2

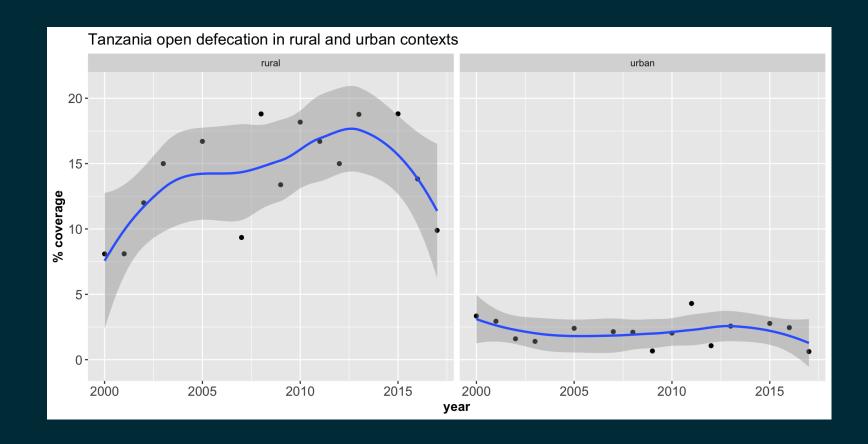
# PLOTTING USING LOESS

### AFGHANISTAN SEWER CONNECTION



- AfghanistanRSE rural =0.043
- AfghanistanRSE urban =0.022

### TANZANIA OPEN DEFECATION



- Tanzania RSE rural = 0.487
- Tanzania RSE urban = 0.033

# SUMMARY

### **COMPARISON OF MODELS**

#### 1. Higher order polynomial

- Pros
  - Easy implementation
  - Significant increase in r.sq for above 10 data points and for coverage values above 10%. The r.sq for open defecation in Tanzania increased from 18% to 49%
- Cons
  - r.sq does not seem to improve for coverage values below 10% and for few data points

#### 2. Splines

- Pros
  - Takes the shape of the data
  - r.sq significantly improves. 100% for Afghanistan sewer connection in rural and urban contexts

- Cons
  - Selection of knots
- 3. Generalized additive models
  - Pros
    - Automatic selection of knots
  - Cons
    - Tends to have a large standard error
    - Not easily explainable

#### 4. Loess

- Pros
  - Takes the shape of the data
  - Low standard error for few data points
- Cons
  - Tends to overfit for few data points
  - Not easily explainable
  - Not suitable for prediction

### **NEXT STEPS**

- Work through secondary indicators
- Create a decision tree that recommends different models for different countries

### DISCUSSION

### QUESTIONS

- 1. How are country files updated on the server?
  - How often? All together, or one by one as there is new data?
- 2. The current database for raw data shows 379 different sources. They are all abbreviated. Do you have a table where all these abbreviations are spelled out?
- 3. Ratio RS1

"Other ratios used for 'basic' indicators (RW1, RS1) are calculated using simple averages" - JMP Methodology, March 2018

# THANKS

#### THANKS!

Data source: washdata.org

Slides - source code: https://github.com/KaraniLinda/SDG-6-2-

Reproducibility/blob/main/slides/r-packages-for-sdg62-data-lkarani-lschoebitz.qmd

Slides - PDF download: https://github.com/KaraniLinda/SDG-6-2-

Reproducibility/raw/main/slides/r-packages-for-sdg62-data-lkarani-lschoebitz.pdf