

# Impact data analysis Uganda

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AN INITIATIVE OF  
THE NETHERLANDS  
RED CROSS

# Impact based forecasting:



## IMPACT BASED FORECASTING



### 1: Understanding risk

- DEVELOP RISK MODELS
- OVERVIEW OF VULNERABLE AREAS
- COMMUNITY RISK ASSESSMENT



POPULATION DATA



COMMUNITY RISK ASSESSMENT DASHBOARD

### 2: Identify impact

- HISTORICAL EVENTS DATA
- ANALYSIS & INSIGHTS
- MACHINE LEARNING
- IMPACT ON POPULATION
- IDENTIFY TRIGGER LEVELS



DATA ANALYSES



MACHINE LEARNING

### 3: Forecast triggered action

- IDENTIFY VULNERABLE PEOPLE
- TRIGGER RELEASE FUNDS
- TAKE ACTION (E.G. DIRECT CASH)



EARLY WARNING EARLY ACTION



SAVE TIME



SAVE LIVES



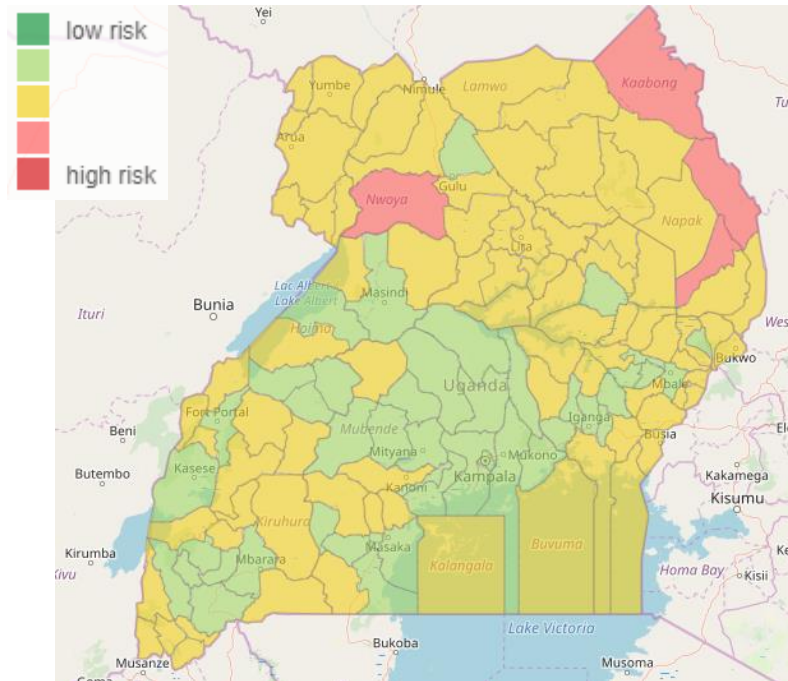
SAVE MONEY



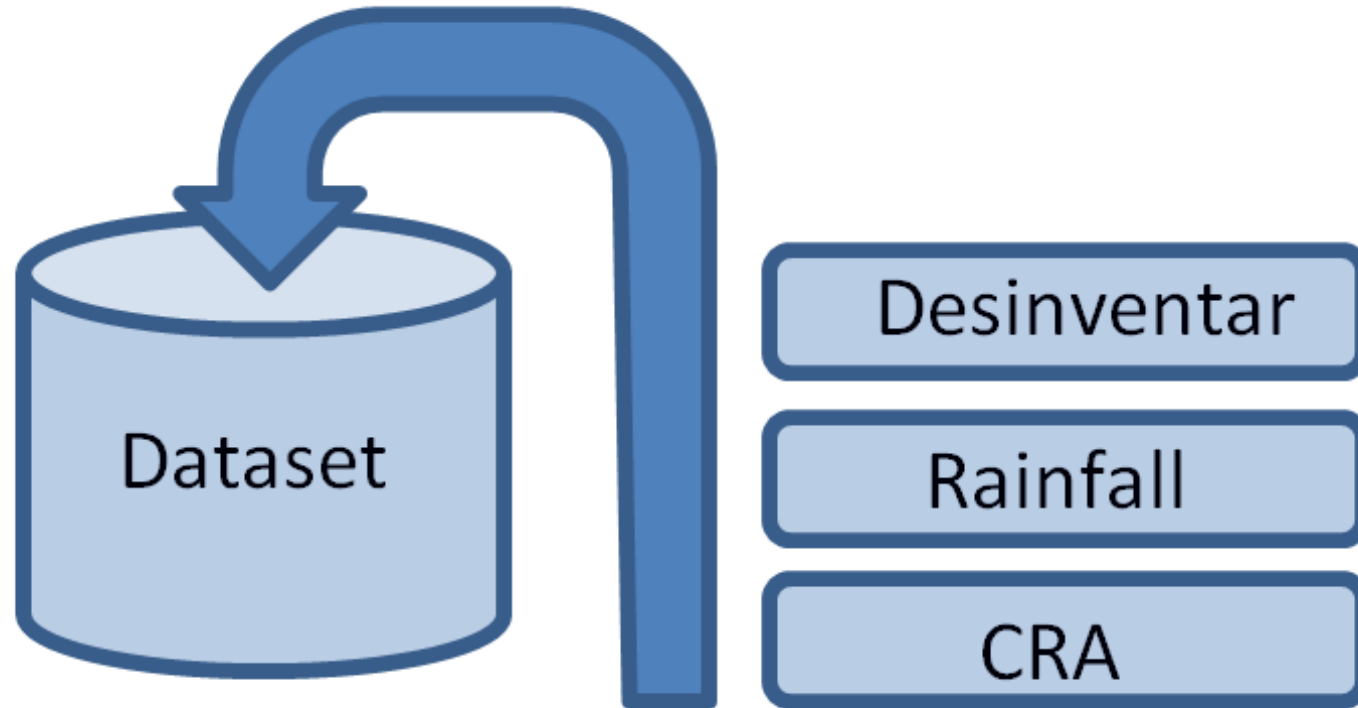
IBF IS THE FIRST THREE STEPS OF A LARGER PROCESS CALLED FBF  
FORECAST BASED FINANCING READ MORE HERE [Red Cross Red Crescent Climate Centre](#)

## Research question:

**“How accurate can we predict the impact of future floods in Uganda at district-level based on historical data (i.e. historical impact and historical rainfall) and Community Risk Assessment data?”**



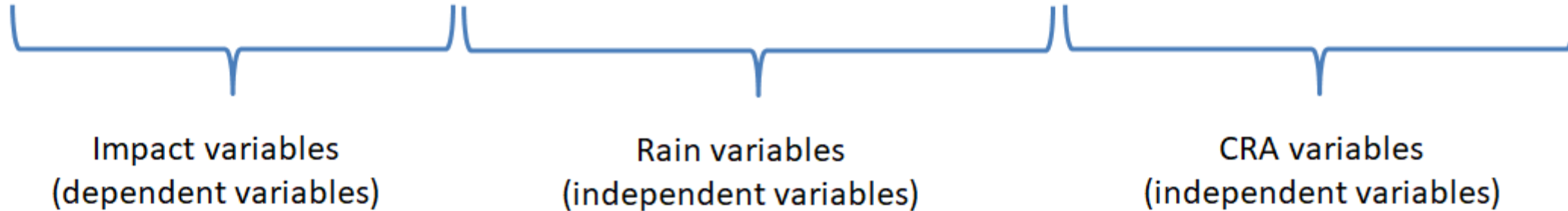
## Datasets:



## Data preparation:

- Merge three datasets

district	date	DI_deaths	DI_injured	DI_....	RAIN_at_day	RAIN_1day_before	RAIN_...	CRA_employed	CRA_literacy	CRA_...
ABIM	2011-04-11	0	0	....	16.53387	38.02542	....	0.9187	0.5891	....
ABIM	2012-07-23	0	0	....	8.214587	10.58498	....	0.9187	0.5891	....
....	....	....	....	....	....	....	....	....	....	....
ZOMBO	2012-06-09	600	3	....	0.000000	6.254879	....	0.9009	0.5305	....
ZOMBO	2017-04-13	0	0	....	2.145846	32.65487	....	0.9009	0.5305	....



## Data preparation:

- Aggregate floods in the same district on the same day or within several days

district	date	DI_deaths	DI_injured	DI_houses_destroyed	....
ABIM	2007-07-29	0	0	0	....
ABIM	2007-07-30	0	0	0	....
ABIM	2007-07-30	300	0	1000	....
ABIM	2007-08-02	600	3	350	....
ABIM	2007-08-02	0	0	0	....
....	....	....	....	....	....

Before aggregation

district	date	DI_deaths	DI_injured	DI_houses_destroyed	....
ABIM	2007-08-02	450	3	675	....
....	....	....	....	....	....

After aggregation

## Data preparation:

- Create one total binary impact variable (impact yes/no)
- Based on only the 9 binary impact variables

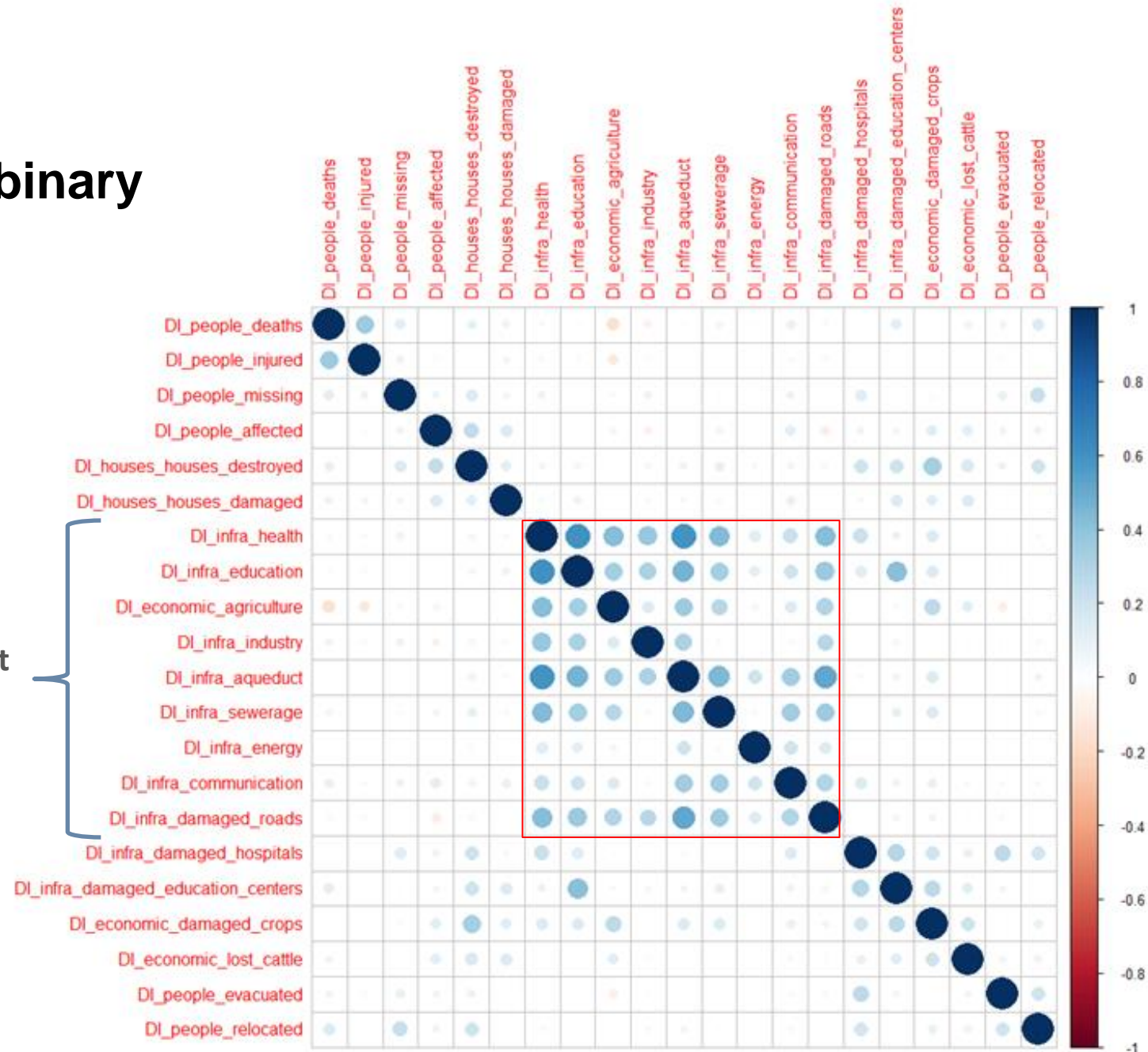


# Data preparation:

## 1. Higher correlation between binary impact variables

9 binary impact variables

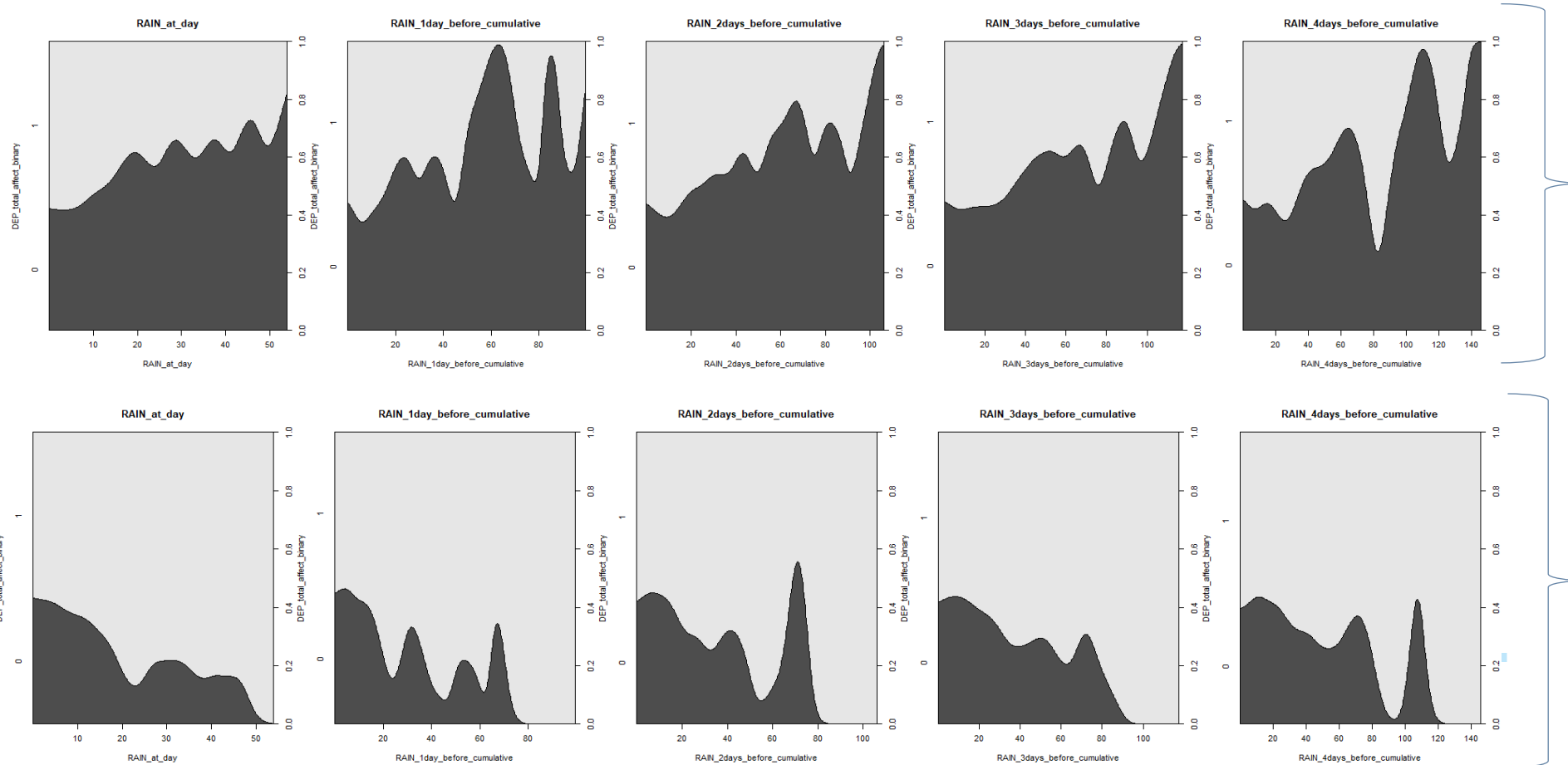
Correlationmatrix impact-variables





# Data preparation:

1. Higher correlation between binary impact variables
2. Positive relationship binary impact variables vs. rainfall



Total binary impact created based on 12 continuous impact variables

Total binary impact created based on 9 binary impact variables

## Data preparation:

- 1. Higher correlation between binary impact variables**
- 2. Positive relationship binary impact variables vs. rainfall**
- 3. More info available for binary impact variables**

## Data preparation:

### - Remove independent variables with more than 85% NA's

#### Rain variables:

- Rain\_at\_day
- Rain\_1\_day\_before\_cumulative
- Rain\_2\_days\_before\_cumulative
- Rain\_3\_days\_before\_cumulative
- Rain\_4\_days\_before\_cumulative
- Rain\_1\_day\_before
- Rain\_2\_days\_before
- Rain\_3\_days\_before
- Rain\_4\_days\_before
- Rain\_5\_day\_before

#### CRA variables:

- CRA\_violent\_incidents
- CRA\_drought\_exposure
- CRA\_earthquake\_exposure
- CRA\_flood\_exposure
- CRA\_disability
- CRA\_employed
- CRA\_literacy
- CRA\_mosquito\_nets
- CRA\_orphans
- CRA\_poverty
- CRA\_roof\_type
- CRA\_wall\_type
- CRA\_subsistence\_farming
- CRA\_drinking\_water
- CRA\_educational\_facilities
- CRA\_time\_to\_city
- CRA\_electricity
- CRA\_health\_facilities
- CRA\_sanitation
- CRA\_internet\_access
- CRA\_mobile\_access
- CRA\_land\_area
- ~~CRA\_displaced\_persons~~
- ~~CRA\_displaced\_local\_population~~
- CRA\_elevation
- CRA\_population\_density
- CRA\_population
- CRA\_general\_coping
- CRA\_general\_risk
- CRA\_general\_hazard
- CRA\_general\_vulnerability

## Data preparation:

- Remove independent variables with less than 10% unique values

### Rain variables:

- Rain\_at\_day
- Rain\_1\_day\_before\_cumulative
- Rain\_2\_days\_before\_cumulative
- Rain\_3\_days\_before\_cumulative
- Rain\_4\_days\_before\_cumulative
- Rain\_1\_day\_before
- Rain\_2\_days\_before
- Rain\_3\_days\_before
- Rain\_4\_days\_before
- Rain\_5\_day\_before

### CRA variables:

- ~~CRA\_violent\_incidents~~ → 16/578
- CRA\_drought\_exposure
- ~~CRA\_earthquake\_exposure~~ → 4/578
- ~~CRA\_flood\_exposure~~ → 23/578
- CRA\_disability
- CRA\_employed
- CRA\_literacy
- CRA\_mosquito\_nets
- CRA\_orphans
- CRA\_poverty
- CRA\_roof\_type
- CRA\_wall\_type
- CRA\_subsistence\_farming
- CRA\_drinking\_water
- CRA\_educational\_facilities
- CRA\_time\_to\_city
- CRA\_electricity
- CRA\_health\_facilities
- CRA\_sanitation
- ~~CRA\_internet\_access~~ → 57/578
- CRA\_mobile\_access
- CRA\_land\_area
- ~~CRA\_displaced\_persons~~
- ~~CRA\_displaced\_local\_population~~
- CRA\_elevation
- CRA\_population\_density
- CRA\_population
- CRA\_general\_coping
- CRA\_general\_risk
- CRA\_general\_hazard
- CRA\_general\_vulnerability

# Data preparation:

## - Remove incorrect variables

### Rain variables:

- Rain\_at\_day
- Rain\_1\_day\_before\_cumulative
- Rain\_2\_days\_before\_cumulative
- Rain\_3\_days\_before\_cumulative
- Rain\_4\_days\_before\_cumulative
- Rain\_1\_day\_before
- Rain\_2\_days\_before
- Rain\_3\_days\_before
- Rain\_4\_days\_before
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### CRA variables:

- CRA\_violent\_incidents
- CRA\_drought\_exposure
- ~~CRA\_earthquake\_exposure~~
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- CRA\_disability
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- ~~CRA\_population~~
- CRA\_general\_coping
- CRA\_general\_risk
- CRA\_general\_hazard
- CRA\_general\_vulnerability

# Data preparation:

## - Remove unimportant variables

### Rain variables:

- Rain\_at\_day
- Rain\_1\_day\_before\_cumulative
- Rain\_2\_days\_before\_cumulative
- Rain\_3\_days\_before\_cumulative
- Rain\_4\_days\_before\_cumulative
- ~~Rain\_1\_day\_before~~
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- ~~Rain\_4\_days\_before~~
- ~~Rain\_5\_day\_before~~

### CRA variables:

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- ~~CRA\_general\_coping~~
- CRA\_general\_risk
- ~~CRA\_general\_hazard~~
- CRA\_general\_vulnerability



## Data preparation:

- 1 dependent variable (total binary impact variable)
- 17 independent variables → standardized

### 5 rain variables:

- Rain\_at\_day
- Rain\_1\_day\_before\_cumulative
- Rain\_2\_days\_before\_cumulative
- Rain\_3\_days\_before\_cumulative
- Rain\_4\_days\_before\_cumulative

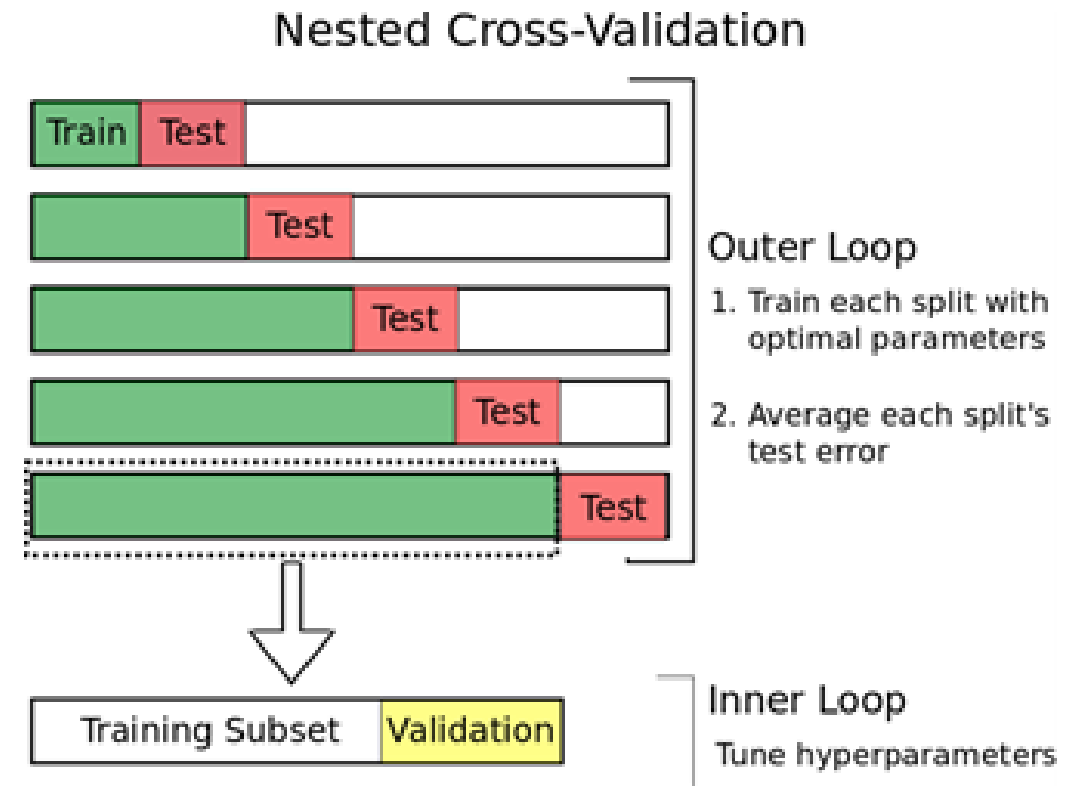
### 12 CRA variables:

- CRA\_disability
- CRA\_literacy
- CRA\_mosquito\_nets
- CRA\_orphans
- CRA\_roof\_type
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- CRA\_subsistence\_farming
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- CRA\_health\_facilities
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## Statistical models:

(Nested) 5-fold cross-validation to get estimates of several performance metrics for 4 different models:

- Stepwise logistic regression
- Lasso logistic regression
- Support vector machine (with radial basis kernel)
- Random forest



# Results:

	Stepwise logistic regression	Lasso logistic regression	Support vector machine	Random forest
AUC	0.666	0.675	0.641	0.644
Accuracy	0.675	0.671	0.678	0.652
F1 score	0.774	0.780	0.794	0.748



	Actual: no impact (0)	Actual: impact (1)
Predicted: no impact (0)	13	10
Predicted: impact (1)	28	65



	Actual: no impact (0)	Actual: impact (1)
Predicted: no impact (0)	10	7
Predicted: impact (1)	31	67



	Actual: no impact (0)	Actual: impact (1)
Predicted: no impact (0)	6	3
Predicted: impact (1)	35	72



	Actual: no impact (0)	Actual: impact (1)
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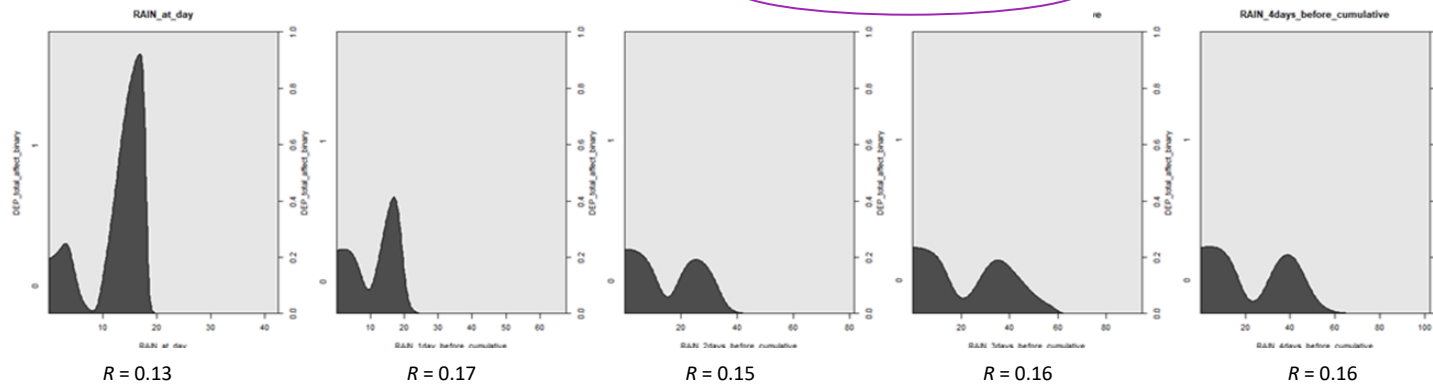
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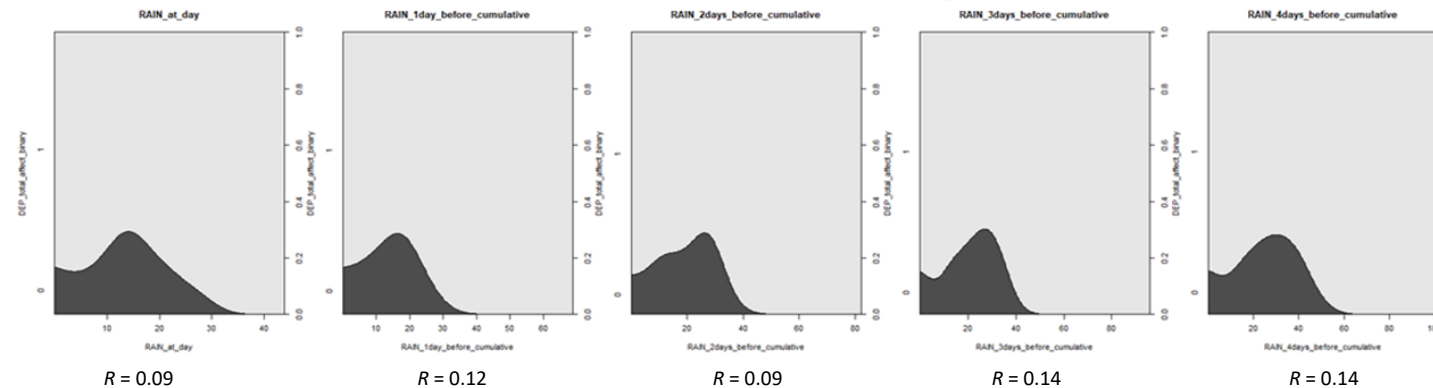
## Future improvements:

- Mean rainfall per catchment area of a district

Conditional density plots: total impact vs. each RAIN predictor  
(Created based on the mean rainfall per catchment area of a district)



Conditional density plots: total impact vs. each RAIN predictor  
(Created based on the mean rainfall per district)






## Future improvements:

- Mean rainfall per catchment area of a district
- Add GloFAS dataset
- Add more accurate impact data
- Predict different impact variables (i.e. related to people, houses etc.)
- Create total impact variable based on expertise knowledge
- Select most important variables based on expertise knowledge
- Resample the minority class (= no impact)
- Tune the parameters of the models (even further)
- Make R-script more reproducible



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# Questions...?