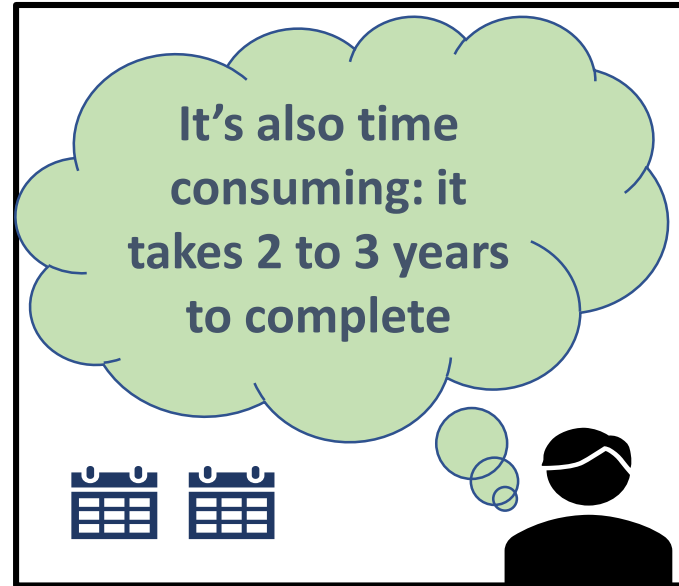




Project Lead:  
**Nobuo Yoshida,**  
*Lead Economist,*  
**Poverty & Equity Global Practice**

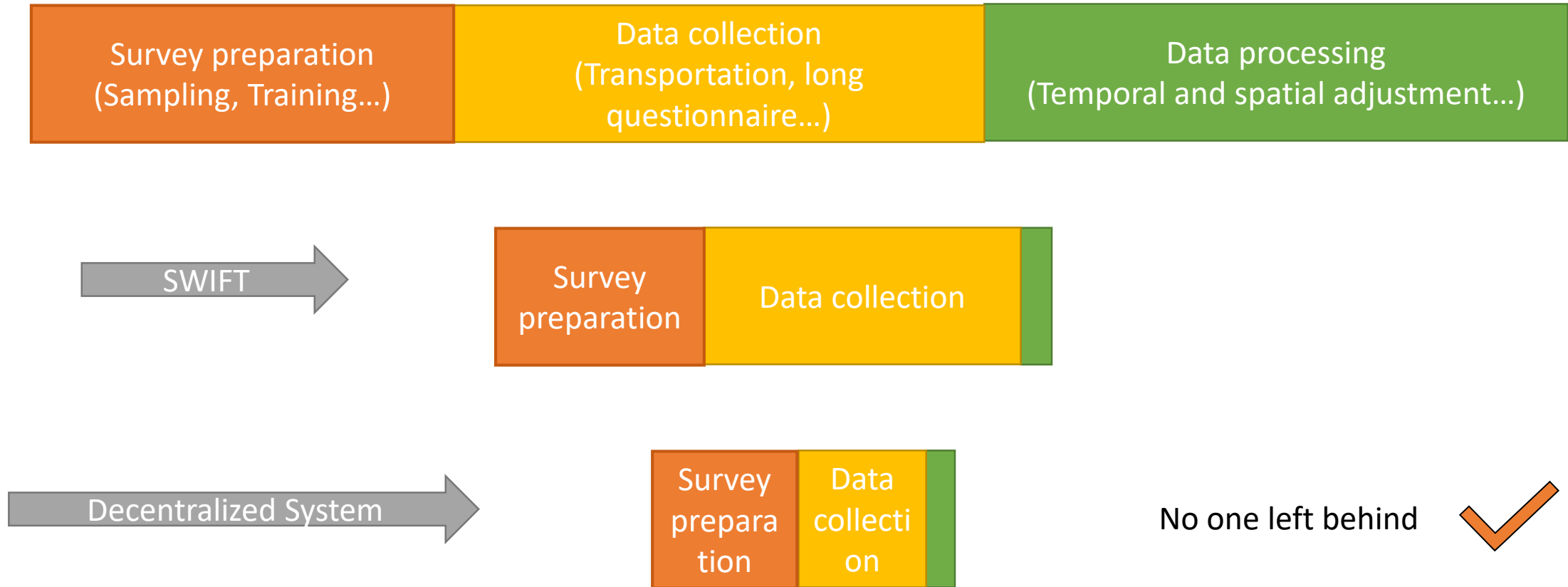


# Constraints to “no-one left behind” commitment



More than 60 countries cannot monitor poverty and shared prosperity

# How can we help?



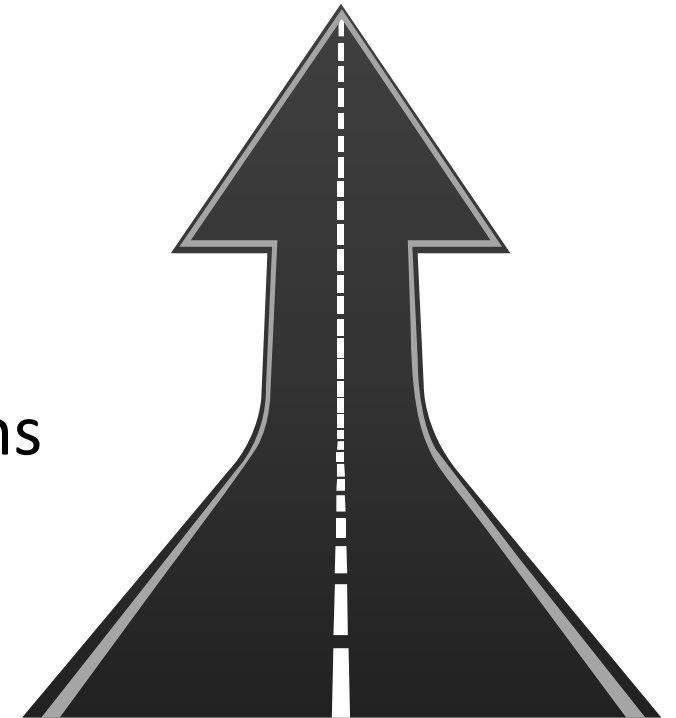
# Sequence of the presentation

- SWIFT Imputation

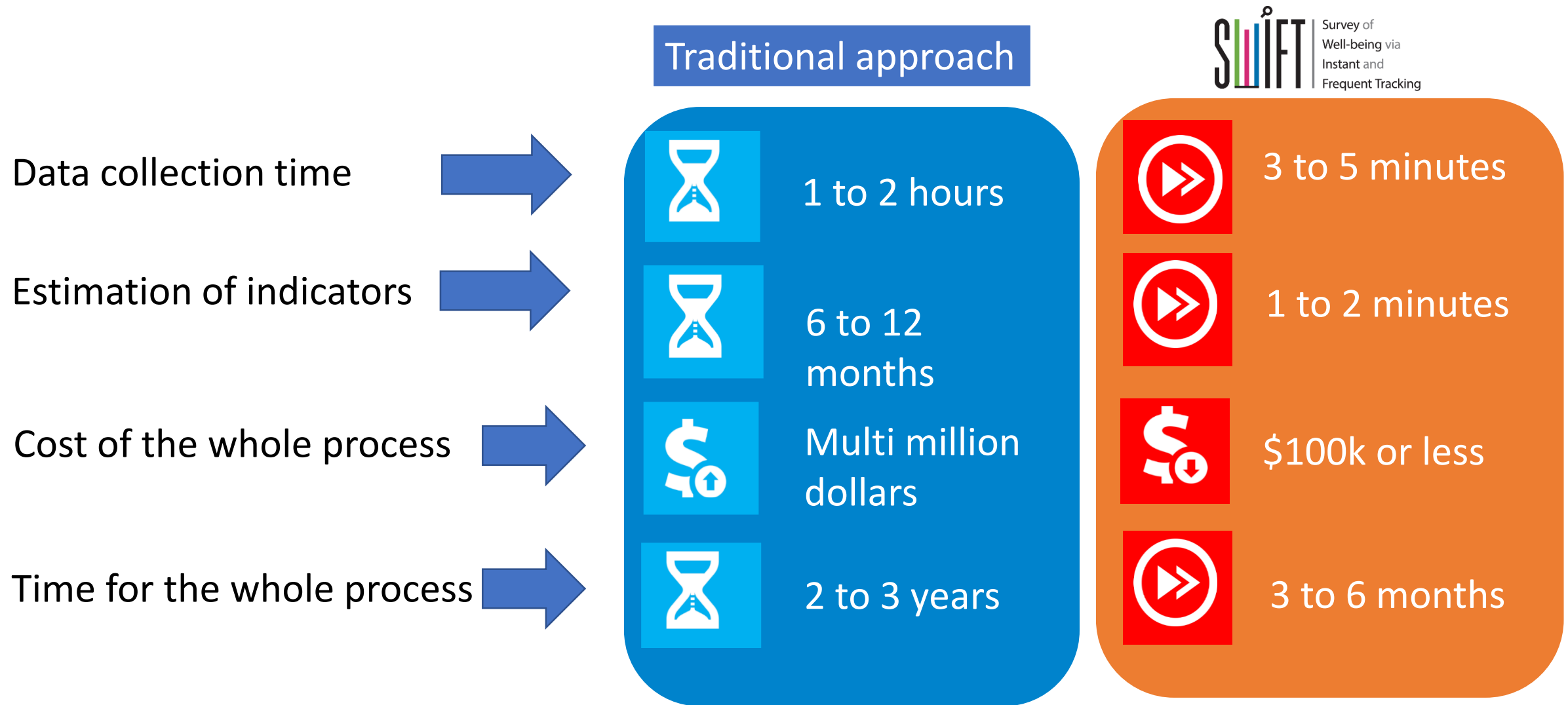
- 1 Standard SWIFT
- 2 Consumption Dummy Based SWIFT Imputation
- 3 SWIFT 2.0: ultimate solution to stability concerns

- Systematic Revolution

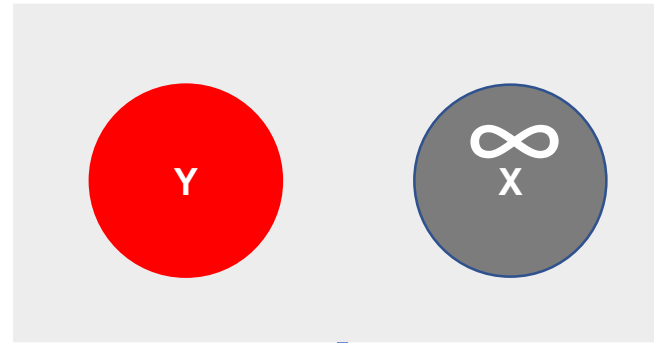
- Decentralized Survey System



# Difference between SWIFT and traditional approach



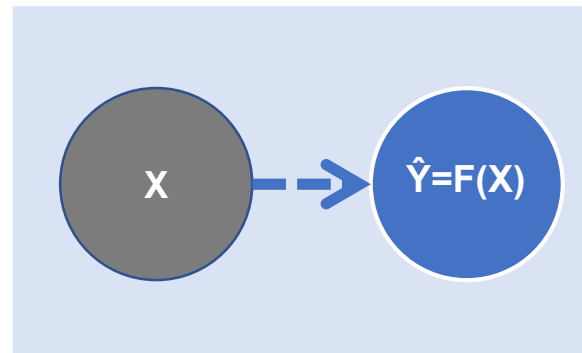
**Baseline Survey:**  
**Most recent HBS**



Use *Machine Learning* techniques to find a **formula** that connect consumption with limited number of non-consumption variables

$$Y=F(X)$$

**Endline Survey**



**Y: Consumption**

**X: Identified non-consumption variables (e.g. housing condition, asset ownerships)**

**$\hat{Y} = F(X)$ : Projected Consumption data**

SWIFT imputation utilizes the existing HBS, extracts 10-20 questions that can be merged into any follow up surveys.

Data collection time and cost reduced significantly.

Imputation to get final poverty estimates is done in real time.

# Modeling

- Create a model by running regressions

$$\ln y_h = \alpha + \beta_1 * x_{1h} + \beta_2 * x_{2h} + \dots + \beta_k * x_{kh} + \varepsilon_h$$

Left hand side variable:  
log of household  
expenditure per capita/per  
adult equivalence

Right hand side variables:  
Stepwise selection from  
candidates pool

Typical Categories	Examples
<b>Household characteristics</b>	Household size, female ratio, dependency ratio, Education level, age, gender, employment, marital status...
<b>Housing condition</b>	Wall material, water source, toilet facility, electricity...
<b>Livestock and agriculture</b>	Cattle, chicken, donkey, agriculture land...
<b>Asset ownership</b>	Bicycle, radio, mobile phone, television...
<b>Location and seasonality</b>	Region, season

# ! Issues

- Over-fitting
  - A model performs well in current dataset but might not project well using new data
- Multi-collinearity
  - Stepwise regression is vulnerable to multi-collinearity
- Stability of coefficients over time
  - Models developed in current round might not hold for other round of survey
- Misspecification of error structure
  - Error distributions can be very complex

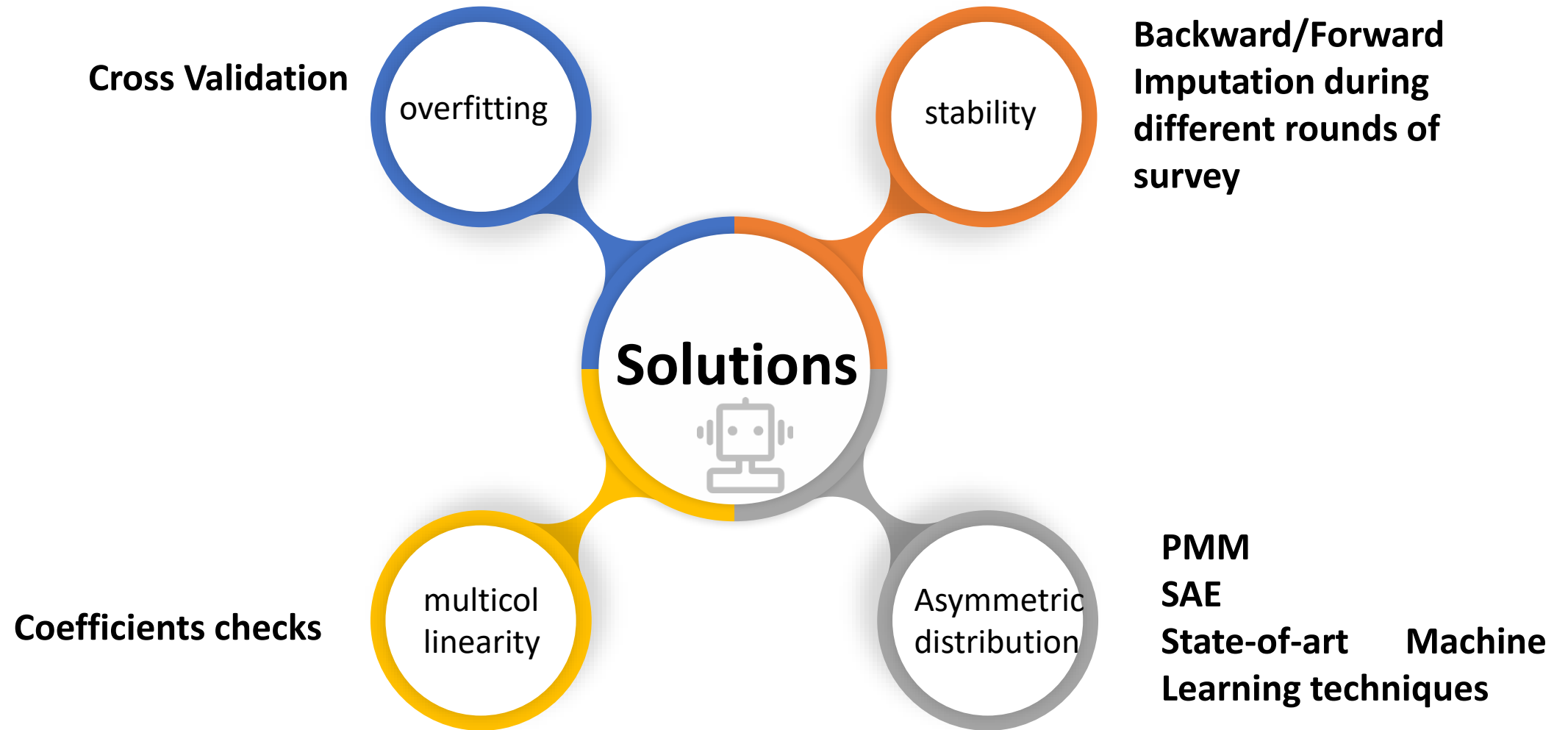




# Solutions

---

1. Cross Validation to minimize risk of overfitting
2. Multicollinearity checks
3. Stability test using Backward/Forward Imputation
4. Addressing distributional issues with PMM, SAE or State-of-art Machine Learning techniques



# Cross-Validation -- > Stepwise Regression

- Cross-Validation is used to see the **out-of-sample performance** rather than within-sample performance
- The risk of **overfitting** problem rises as more variables are included
- Using the **cross-validation** approach, we try to find the optimal number of variables
- The **optimal p-value** for the stepwise regression set the threshold for including more variables, which find the balance between in-sample accuracy and out-of-sample projection

# Simulation stage – Multiple Imputations

- ▶ Simulate HH expenditure for each household in SWIFT Sur

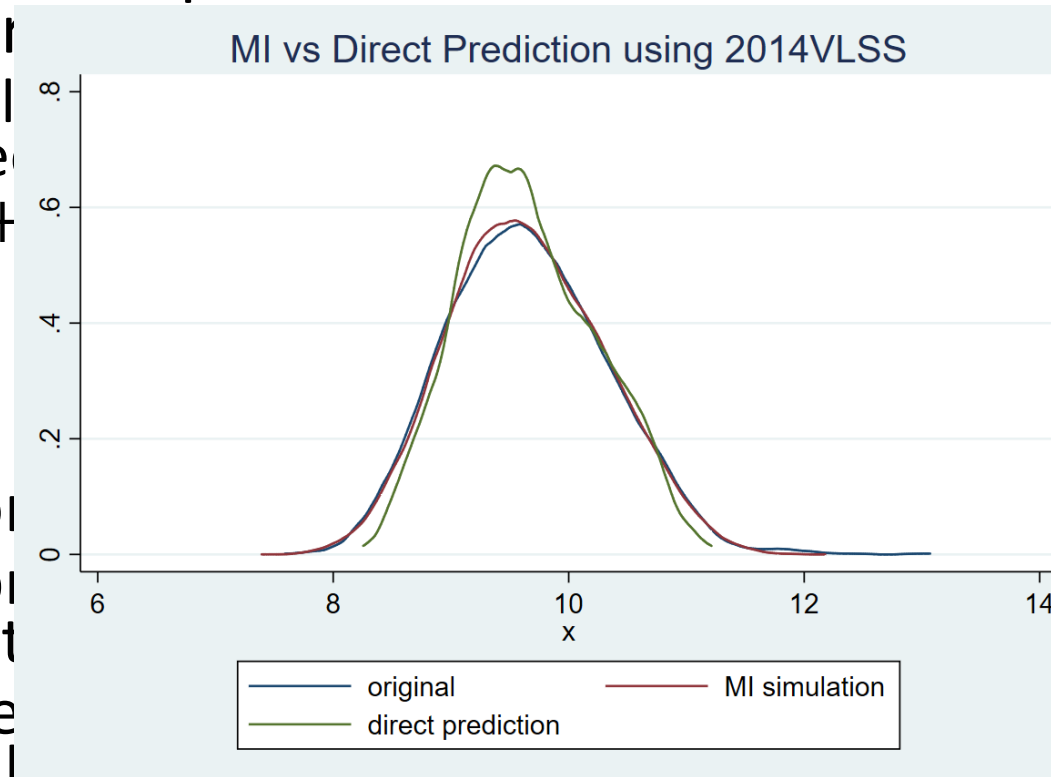
- ▶ Randomly estimate

$$\ln y_h = \alpha +$$

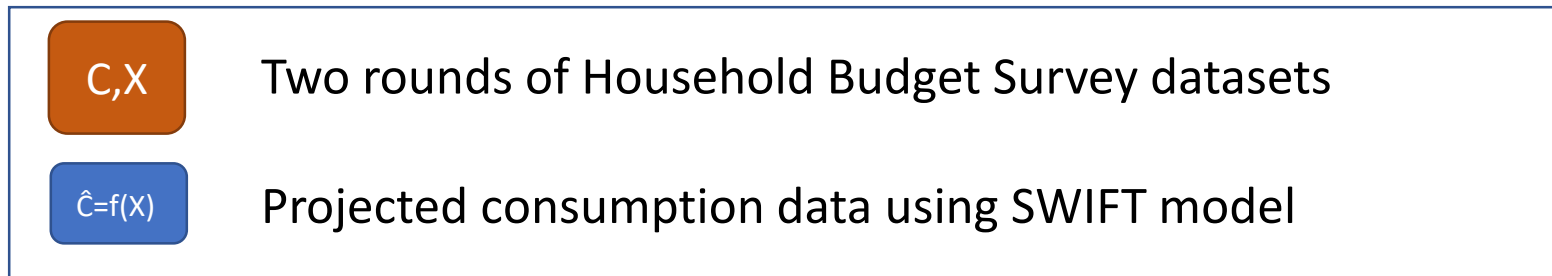
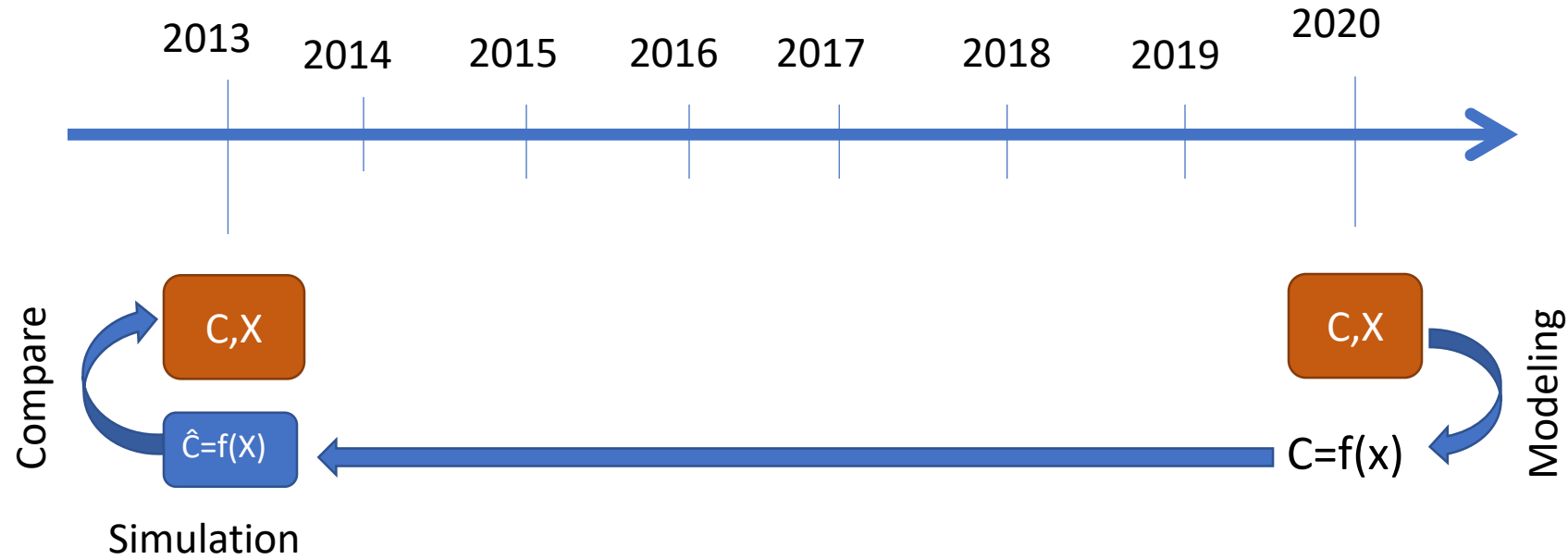
- ▶ Simulation

- ▶ Direct prediction around the

- ▶ Multiple randomly draws errors from the estimated distribution



# Stability Test



# Steps for SWIFT modeling & simulation

1. Cross-validation to decide the optimal p-value
2. Run the stepwise regression using the optimal p-value
3. Check the coefficients of the final model
4. Simulate household expenditures using Multiple Imputation
5. Backward imputation analysis to check model stability

# Standard SWIFT can fail -- A real case in Afghanistan

	Original	Standard SWIFT Projection using 2011 model
2011	38.3%	
2016	54.5%	44.8%



44.8% is far from original 54.5%.  
Why standard SWIFT fails to capture the change?

Household  
assets are  
accumulating

Standard SWIFT is  
Asset Ownership  
based. In Afghanistan,  
asset ownerships are  
not reliable welfare  
indicators.

Household report  
worsened economic  
condition, less  
satisfied with police  
work, and much  
insecure

# Consumption Based SWIFT Imputation

## Standard SWIFT

- Asset Ownership Approach
  - Asset ownership irreversibility
    - Can people trade in their useless assets for money?
    - Do they have fair second hand market?

Replaced

## Revised SWIFT

- Consumption Dummy Approach
  - Food and non-food dummies are more sensitive to welfare change





## Standard SWIFT

82% asset increased

Replaced

## Revised SWIFT

73% consumption decreased



Asset Ownership Dummy	2011/12	2013/14	2016/17
Refrigerator	12.0%	16.2%	16.3%
Stove	64.7%	75.8%	78.9%
Sewing Machine	63.7%	63.6%	64.6%
Iron	56.7%	58.2%	63.7%
Radio	57.1%	43.3%	30.9%
TV	34.9%	39.4%	42.3%
Motorcycle	23.8%	28.0%	32.1%
Car	12.5%	13.6%	14.6%
Tractor	1.9%	3.4%	2.6%
Carpets	17.8%	31.4%	17.2%
Blankets	99.6%	96.7%	99.6%

No fair second hand market

Consumption Item Dummy	2011/12	2016/17	Consumption Item Dummy	2011/12	2016/17
Cigarettes	17.5%	15.1%	Women shoe	96.8%	93.6%
Tobacco	35.0%	37.1%	Child shoe	88.3%	87.8%
Matches	98.4%	96.9%	Wedding	7.8%	8.0%
Cleaning	95.6%	96.7%	Celebration	44.1%	32.4%
Soap	82.9%	81.1%	Hospital	3.5%	3.3%
Shampoo	94.3%	93.6%	high quality rice	43.6%	37.6%
Toothpaste	60.0%	54.9%	low quality rice	75.4%	68.3%
Grooming	61.8%	60.9%	Purchased Nan	10.9%	12.4%
Bath out	16.3%	10.7%	Beans	53.2%	50.7%
Laundry	24.8%	12.9%	Pasta	26.0%	25.0%
Transport	52.5%	57.6%	Beef	28.0%	12.2%
Fuel	25.4%	33.1%	Chicken	35.8%	33.8%
Tax	5.0%	6.4%	Milk	45.2%	31.5%
Construction	27.6%	17.4%	Yogurt	49.0%	44.2%
Education	11.0%	11.2%	Dough	24.1%	20.6%
Uniform	42.2%	39.3%	Egg	38.9%	27.2%
Test book	20.4%	22.0%	Potato	91.9%	84.9%
Pen	49.8%	57.9%	Onion	92.2%	93.3%
Repair	16.7%	24.3%	Tomato	39.7%	46.5%
Airfare	1.8%	1.5%	Fruit	61.6%	52.3%
Men cloth	96.5%	95.1%	White sugar	86.0%	82.9%
Women cloth	96.9%	95.5%	Chocolate	58.6%	52.9%
Child cloth	90.4%	89.9%	Black tea	45.1%	34.0%
Men shoe	96.9%	93.5%	Green tea	84.0%	87.0%

Tight budget. Less consumption

# Consumption Based SWIFT Imputation

## Simulation Results

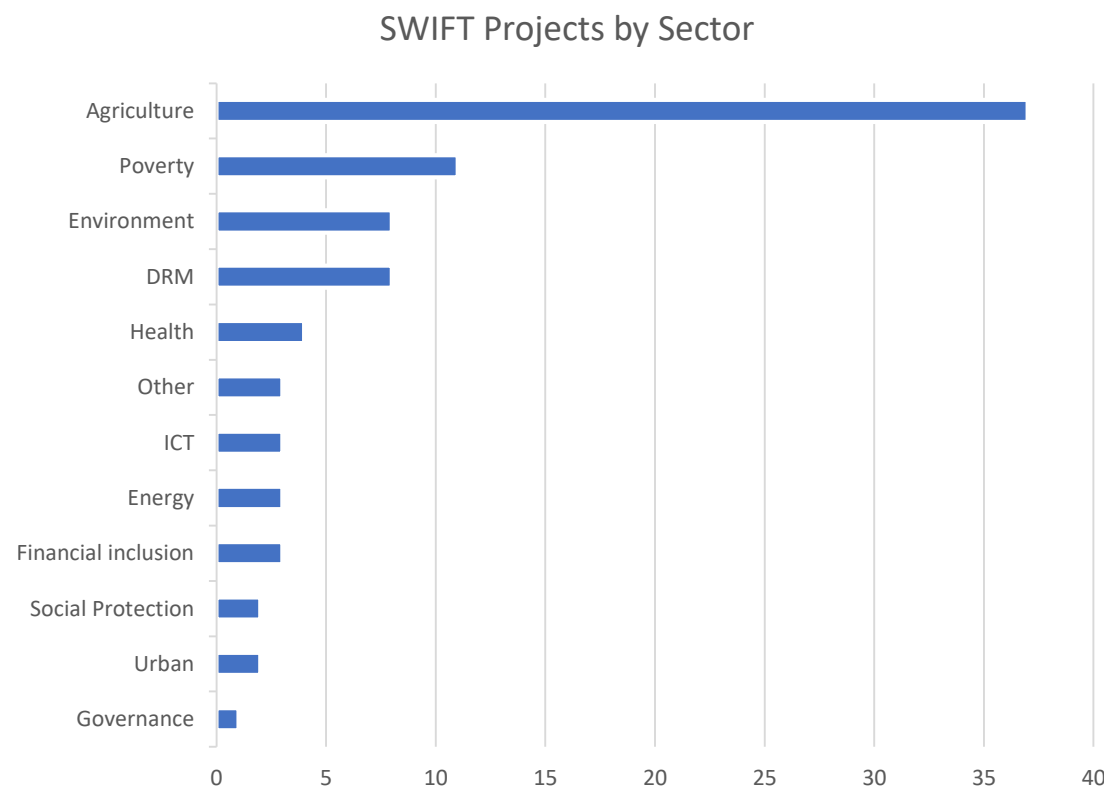
	Original	Standard SWIFT Imputation	Consumption Based SWIFT
2011	38.3%		
2016	54.5%	44.8%	53.5%



Significant improvement

# Current status of SWIFT projects

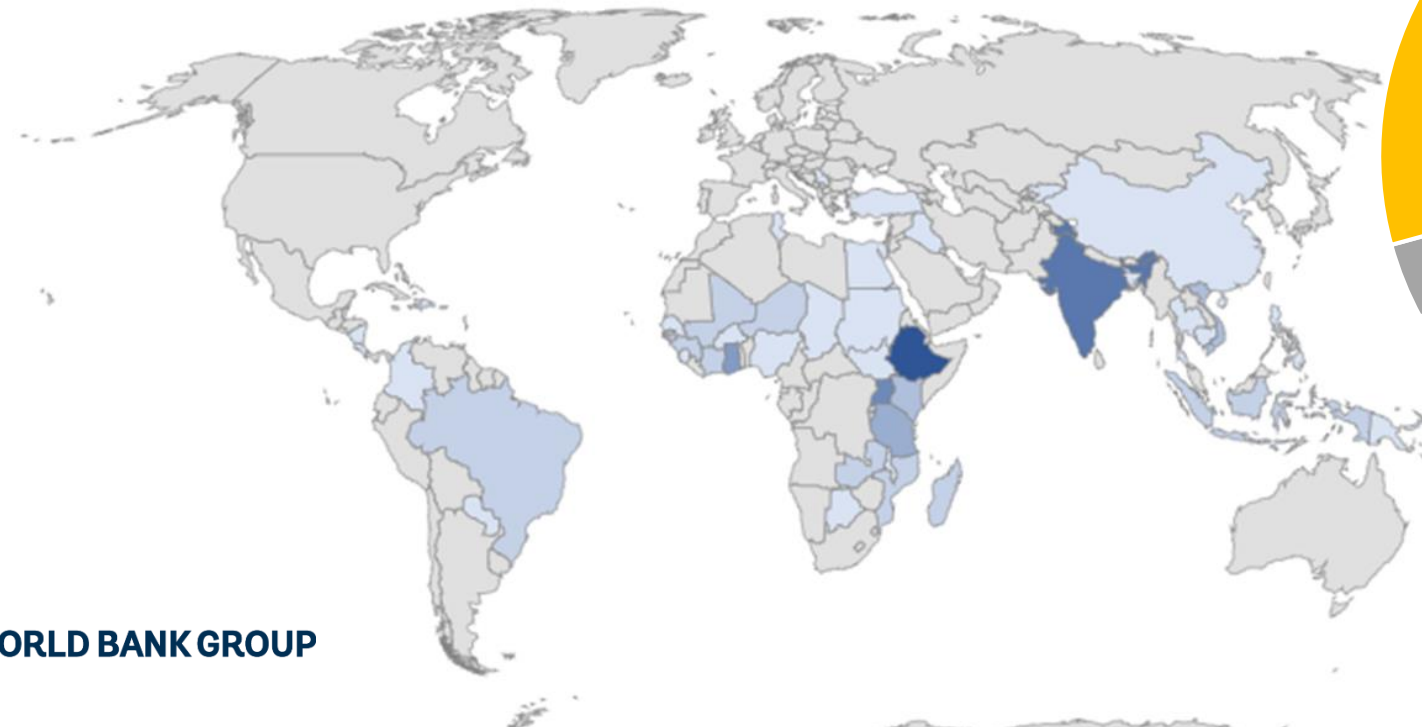
- SWIFT project was launched in June 2015
- Over 100 SWIFT surveys in 52 countries are under implementation or completed
- Varieties of field including DRM, ICT, health, agriculture, energy, etc.



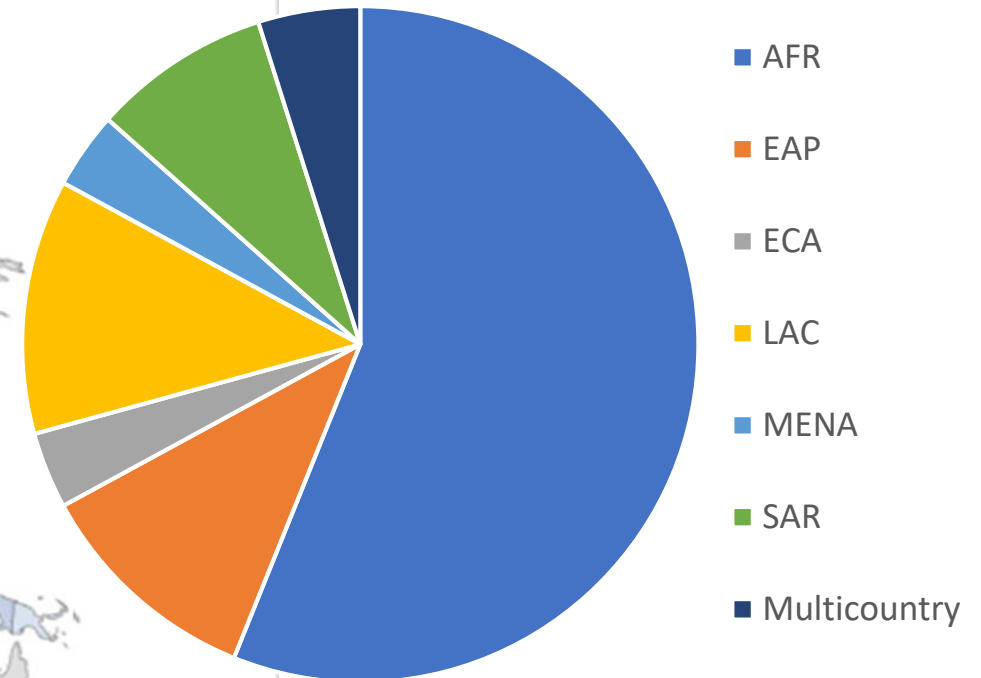
# Current status of SWIFT projects

SWIFT Projects by Country

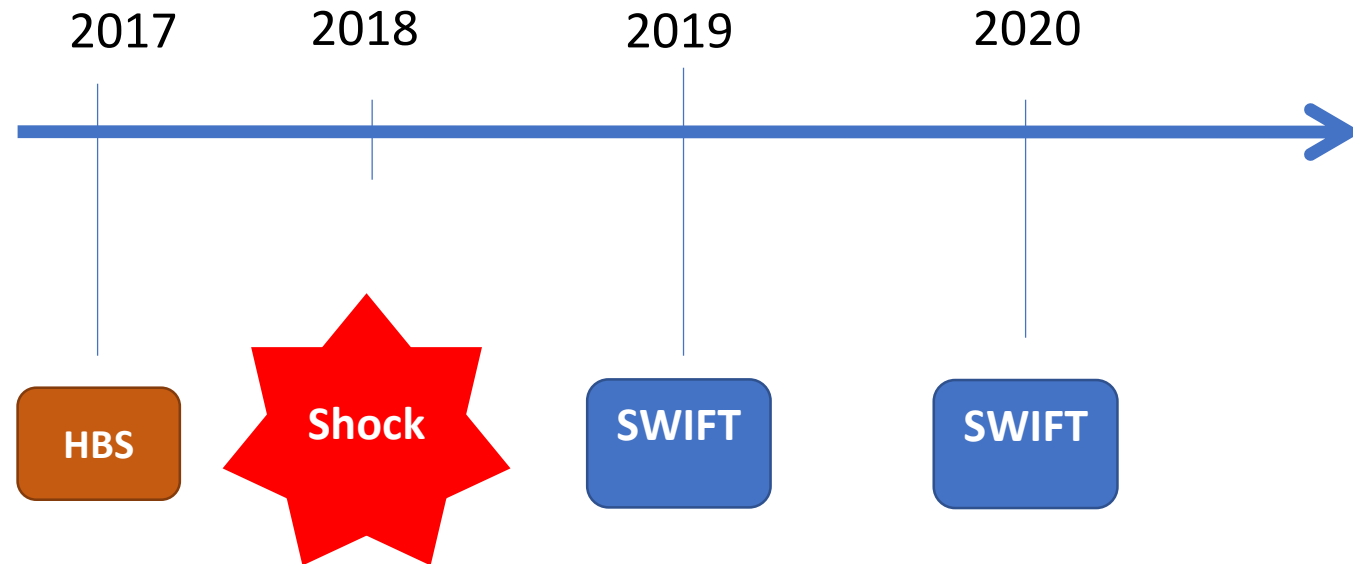
Number of Projects   
1 9



Projects by region

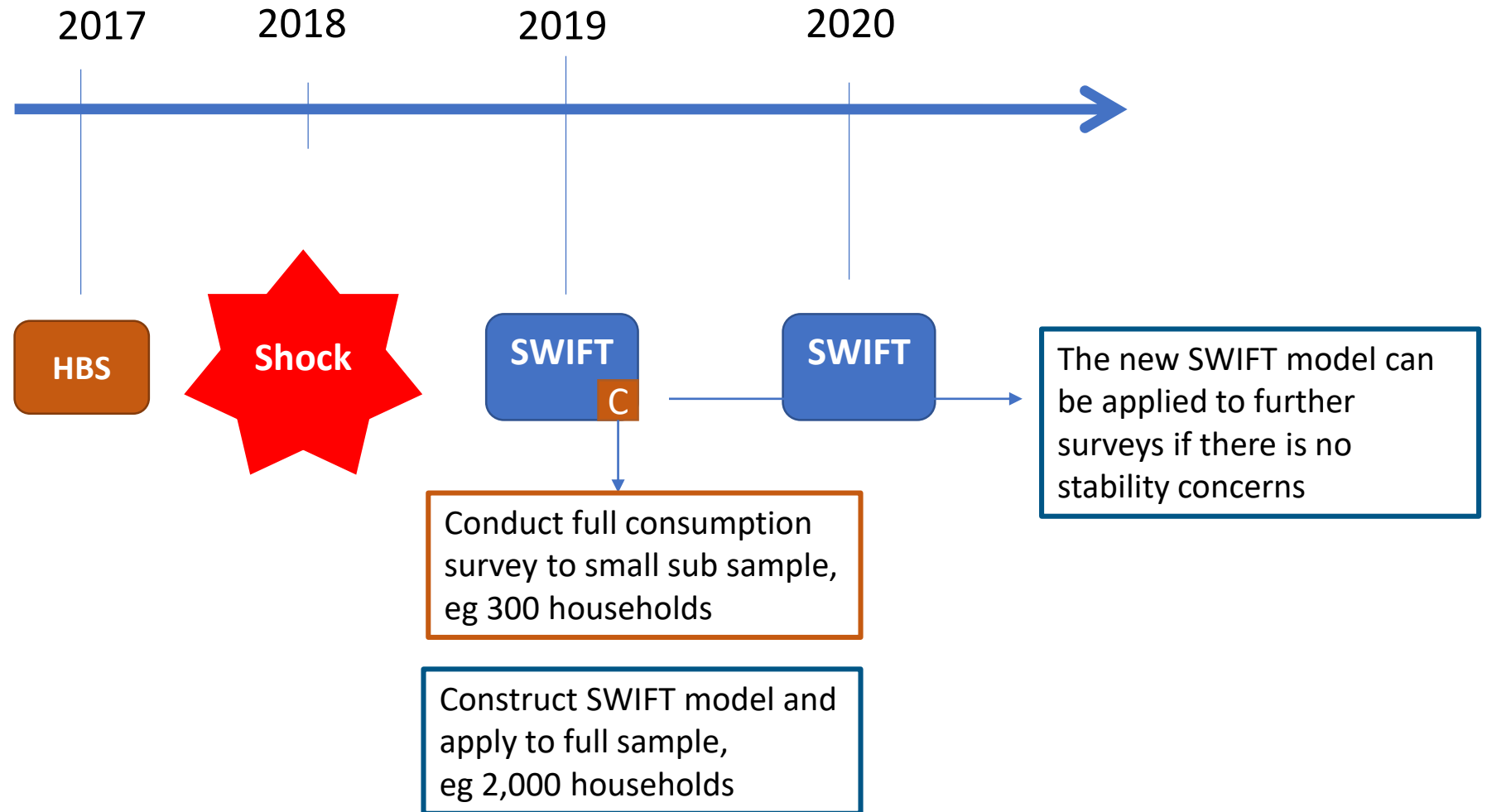


# A model might not be stable over time



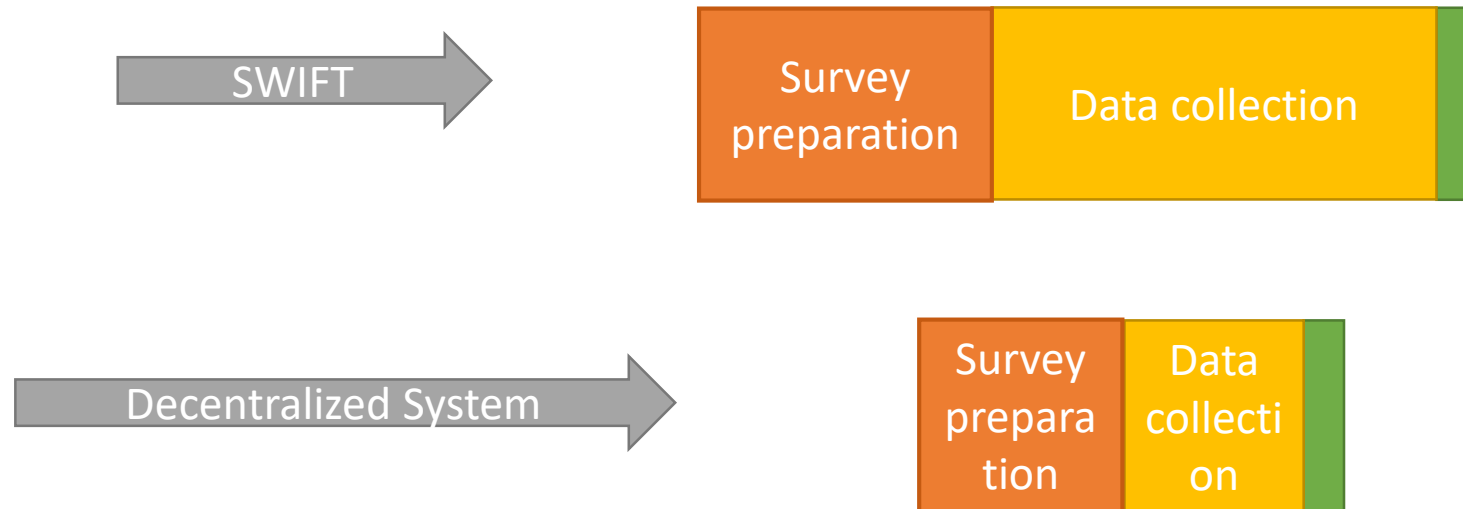
# SWIFT 2.0: Ultimate solution to Stability concerns

- Collect consumption data from a **small** sub-sample in each round of hybrid surveys
- Use them to recreate imputation models for each round



# But even with SWIFT, preparation of household survey is painful, costly, time-consuming...

- Survey preparation includes many activities:
  - Design sampling with help of National Statistics Office or using satellite images
  - Design questionnaires; hire/train enumerators; monitoring of data collection
- As a result, the data collection with SWIFT costs \$50k - \$100k and requires 2 – 3 months of preparation



# Structure Image

Central HQ + WB support team

- Management of the human resources, payment, contracting
- Real-time data monitoring & Analysis
- Coordination with other donors

- Data sent to the cloud
- Regular report on any unusual events

- Instruction when something is wrong
- Regular sharing of the results of the data

Supervisor

Supervisor

Supervisor

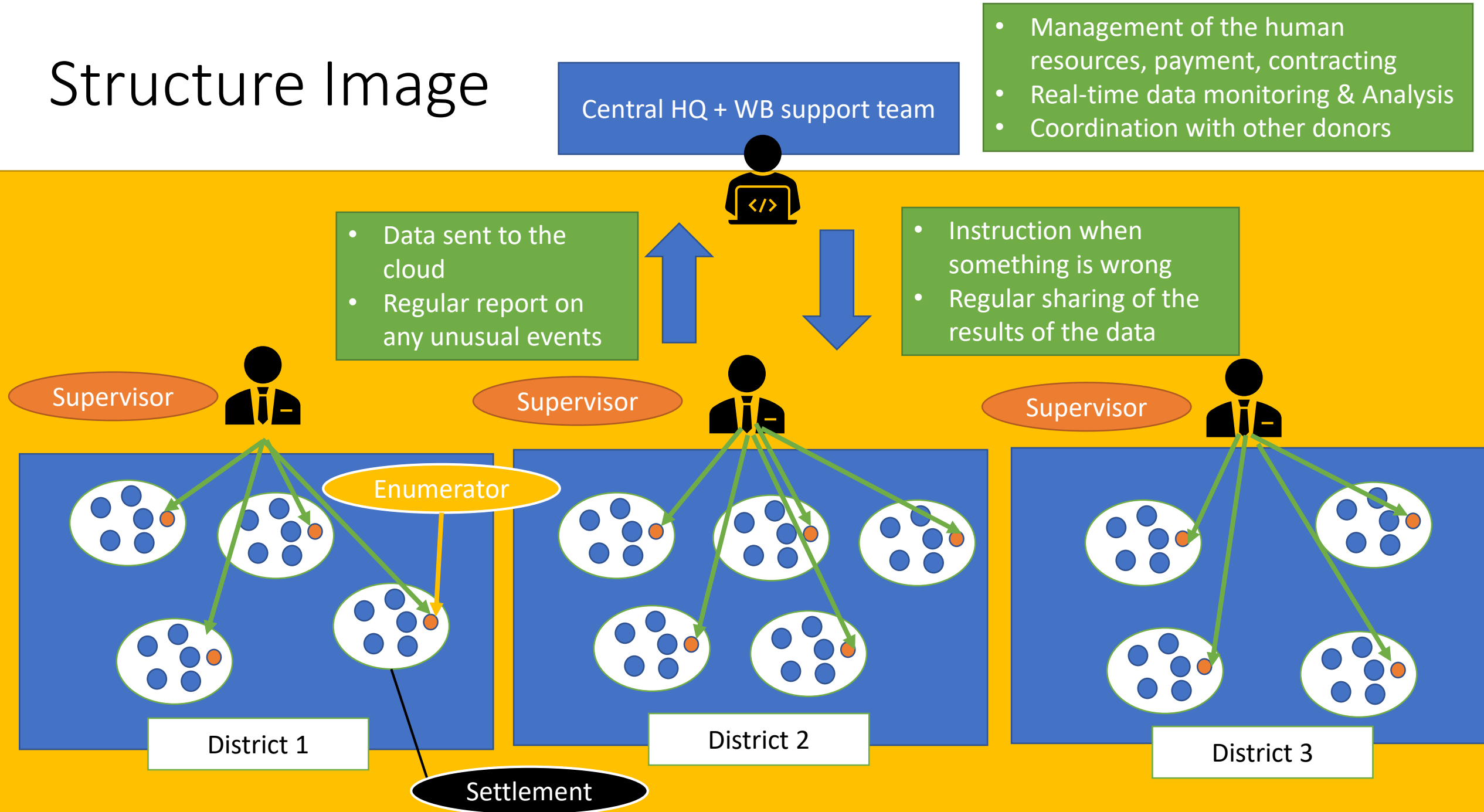
Enumerator

District 1

District 2

District 3

Settlement





Furthermore – Similar survey setup are redone by multiple agents  
eg Malawi's three rapid monitoring surveys

## USAID

- Health survey
- Survey preparation and data collection

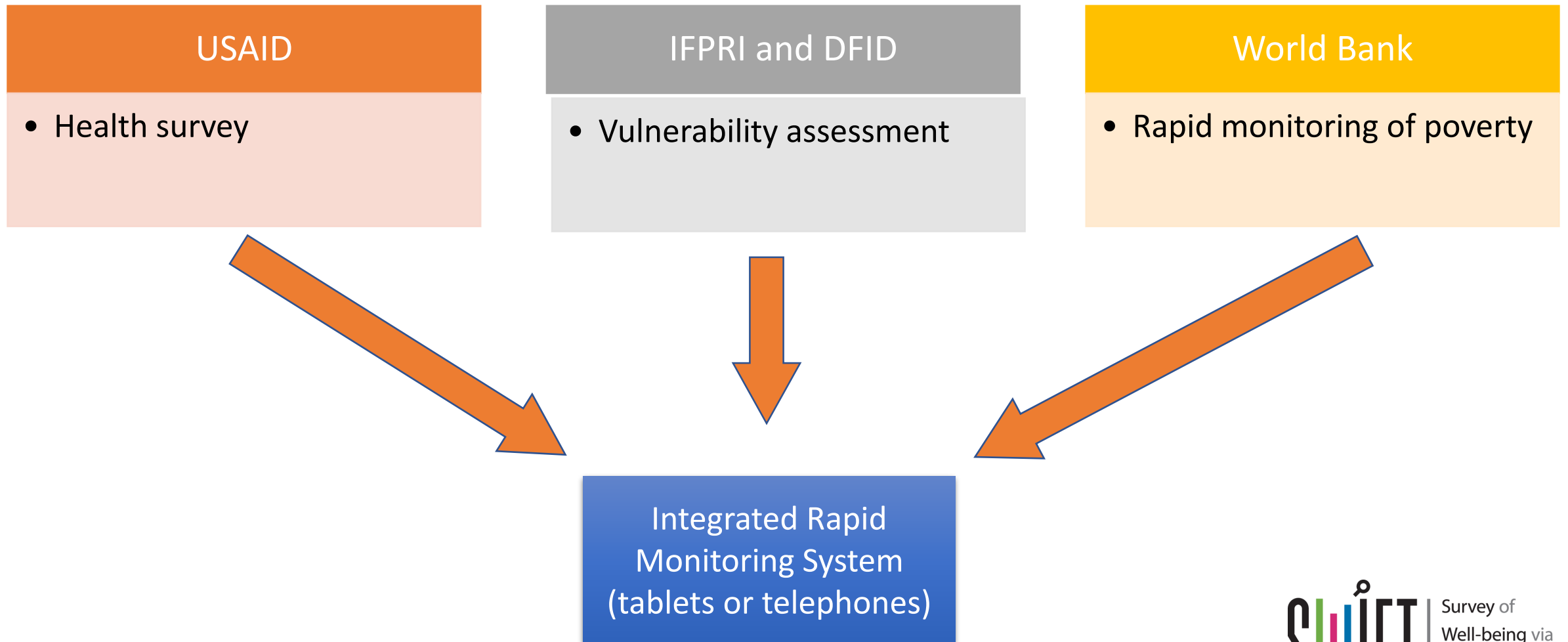
## IFPRI and DFID

- Vulnerability assessment
- Survey preparation and data collection

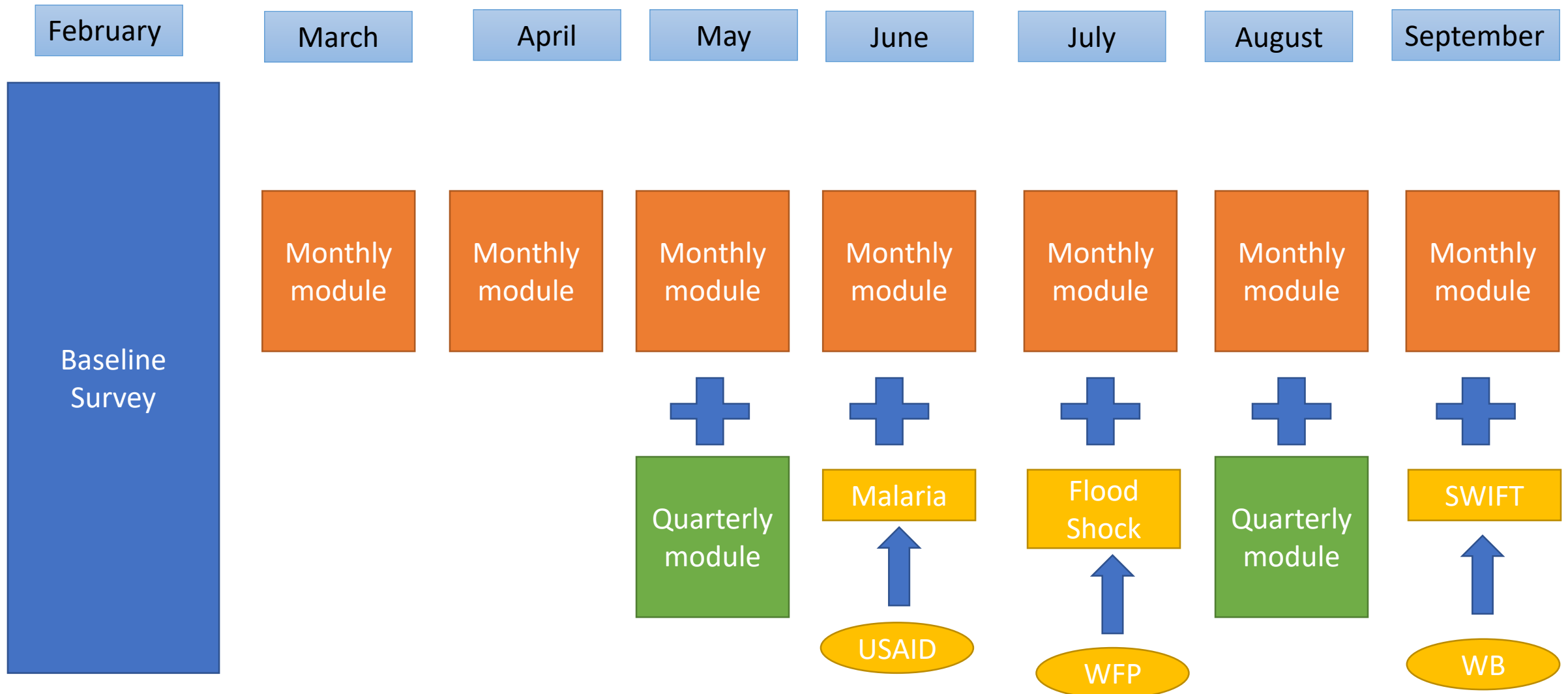
## World Bank

- Rapid monitoring of poverty
- Survey preparation and data collection

# We propose an Integrated Rapid Monitoring System



# Structure of the Questionnaire: Example



# Thank you!

Xiaomeng Chen

[xchen8@worldbank.org](mailto:xchen8@worldbank.org)

Nobuo Yoshida

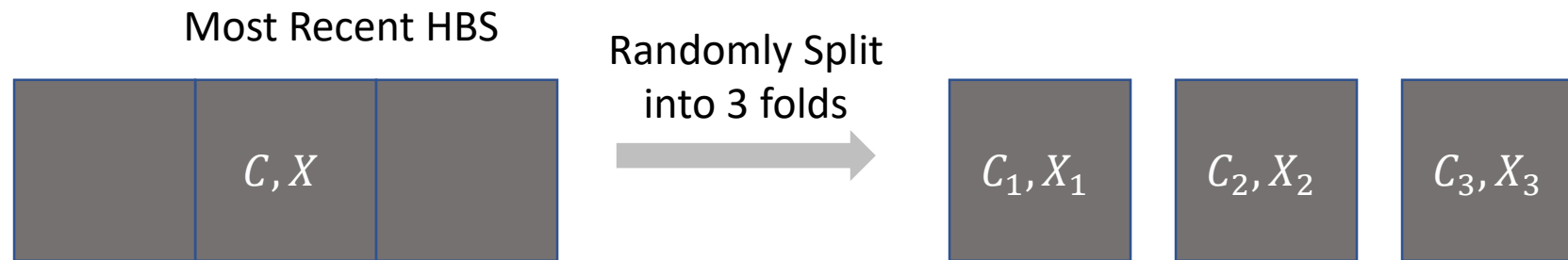
[nyoshida@worldbank.org](mailto:nyoshida@worldbank.org)



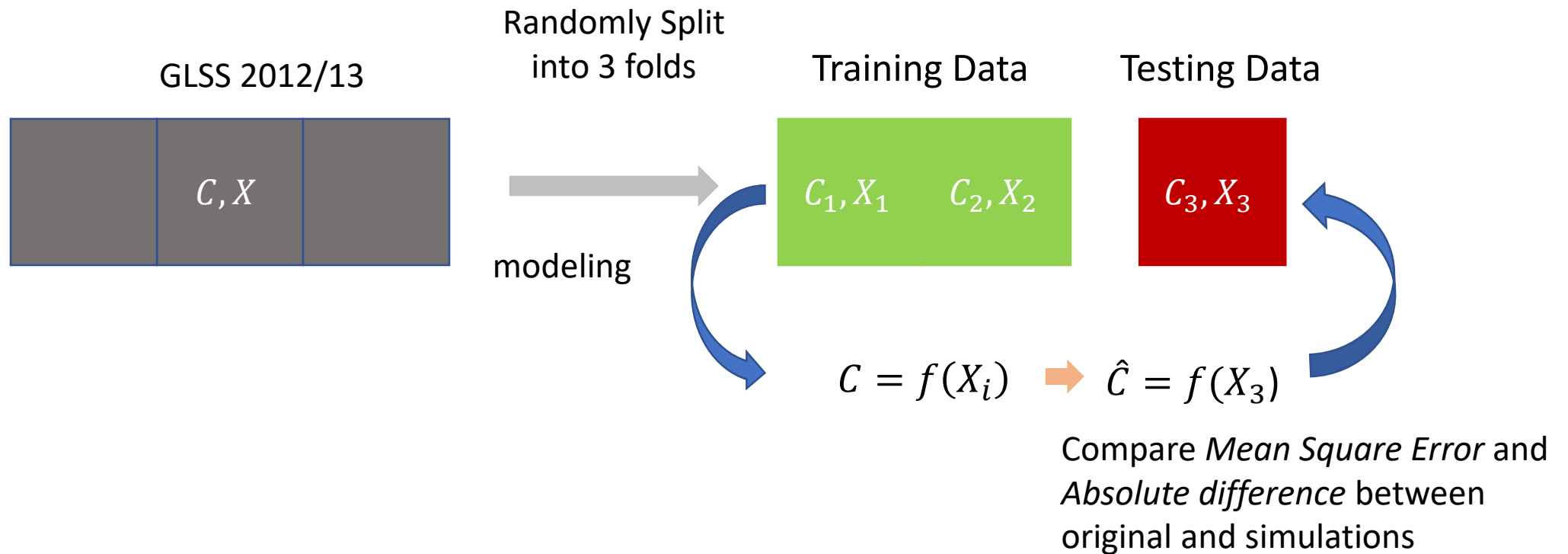
Survey of  
Well-being via  
Instant and  
Frequent Tracking

# Annex

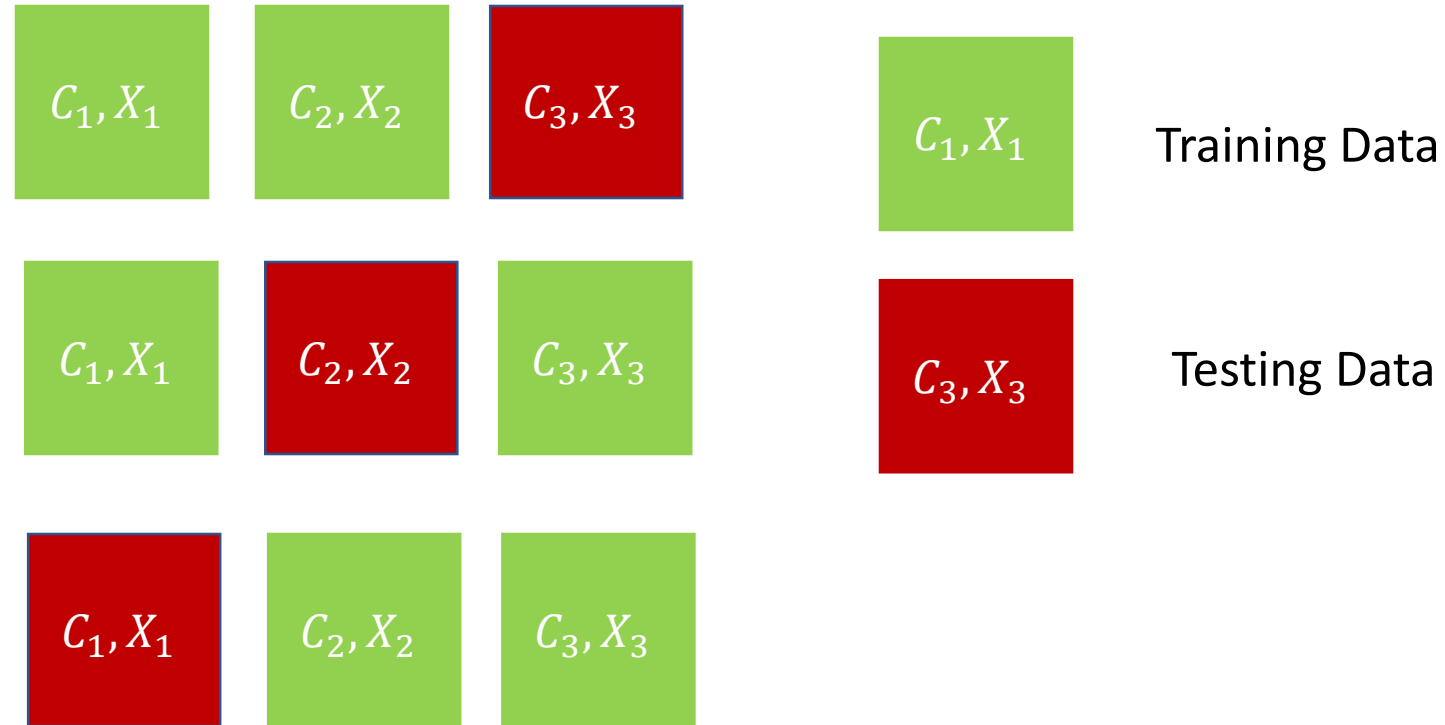
# Cross-Validation: Step 1



# Cross-Validation: Step 2



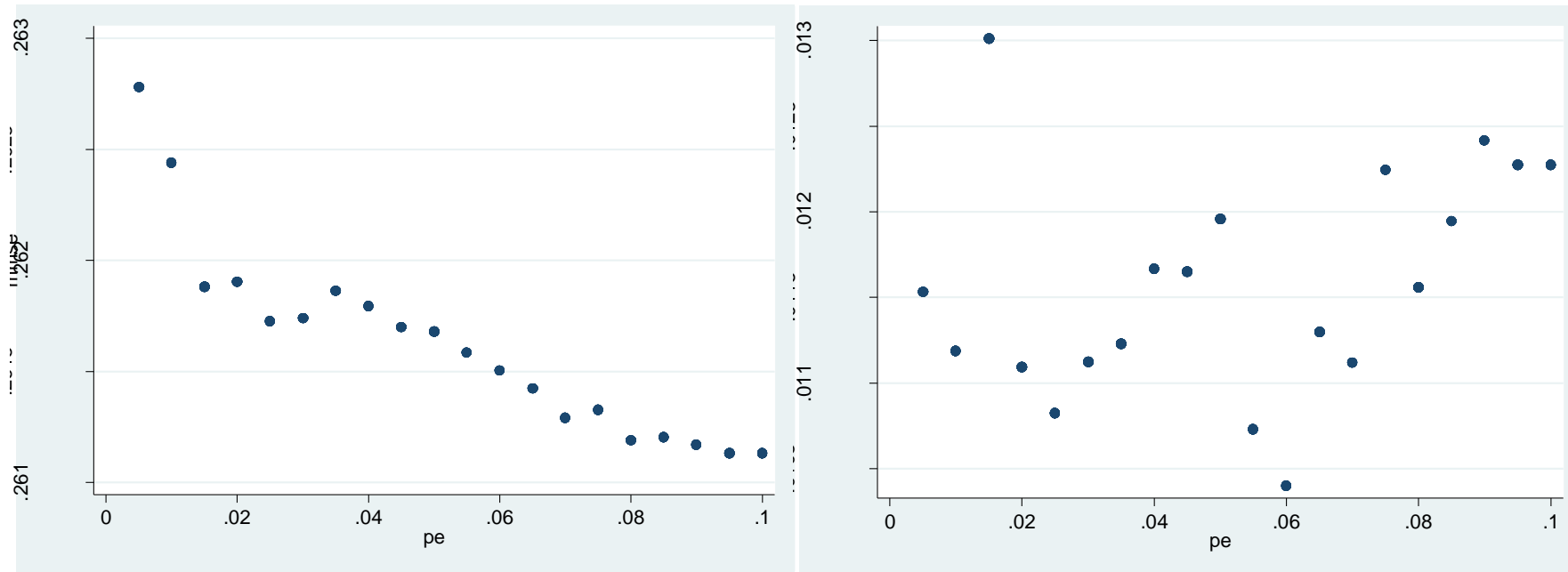
# Cross-Validation: Step 3





# Cross-Validation: Select the best p-value for stepwise regression

P-value = 0.06 is the optimal p-value



Mean Squared Errors

Average of absolute  
differences between actual  
and projected poverty  
rates