

# Can Soft Skills Training Improve Enterprise and Employment Growth in Tanzania? A Randomized Evaluation of the Strengthening Rural Youth Development through Enterprise (STRYDE) 2.0 Program

Pre-analysis plan<sup>1</sup>

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<sup>1</sup> This document outlines the pre-analysis plan for a randomized evaluation of the STRYDE 2.0 program – a comprehensive youth employment program implemented by TechnoServe and funded by MasterCard Foundation in urban and rural Tanzania. This pre-analysis plan has been prepared in advance of any study-related fieldwork commencing in Tanzania (including, in particular, pre-testing of the survey instruments). We will update this document as necessary once all associated preparatory activities are complete but before data analyses have commenced.

## 1. Description of the sample to be used in this study

The research team partnered with TechnoServe (TNS), a non-governmental organization, to evaluate their Strengthening Rural Youth Development through Enterprise (STRYDE) 2.0 program using a cluster randomized controlled trial. The STRYDE 2.0 program in Tanzania aims at targeting 10,800 youth aged between 18 and 30 from two regions located in the southwest of Tanzania - Mbeya and Songwe region - to be trained in six cohorts between 2015 and 2019.

This study follows cohort number four of STRYDE 2.0 participants. The cohort was mobilized between April and June 2017, trained between July and October 2017, awarded in a Business Plan Competition (BPC) of December 2017, and supported during the aftercare till spring 2018. Our goal is to understand the impact of the training, as well as the additional impact of receiving capital to buy materials and tools to run a small business.

In cohort four, mobilization was attempted in 161 clusters: 9 wards in the Mbeya Urban area and 152 villages in the rural areas of the Mbeya and Songwe regions. Mobilization was conducted in villages grouped in pairs based on location: each pair was close enough to allow trainers (hereinafter referred to as Business Counselors, BC) to access both, but far enough to minimize spillovers. Mobilization was considered successfully completed if at least 20 participants per cluster completed the baseline. Once the mobilization was completed, clusters were randomly assigned to treatment or control within each BC strata. The final baseline sample includes 135 clusters and 4,537 observations divided between 72 treatment and 63 control clusters.

For the endline, we plan to follow-up 24 individuals per cluster for a total of 3,200 observations (about 80% of the baseline sample). These individuals will be randomly selected from the full sample. The experimental sample for the evaluation of the BPC was reduced and includes 88 clusters<sup>2</sup> for a total of 1,524 observations, plus 264 standard winners. The BPC sample is partially overlapping with the STRYDE baseline as it involves the groups of the BCs that were not part of the impact evaluation as well as the groups, and students, of the other BCs who were part of the evaluation but allowed to re-mobilize participants as needed<sup>3</sup>.

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<sup>2</sup> These include 75 clusters/ training groups from the experimental sample (i.e. 72 clusters comprising 69 rural villages and 3 urban wards with two training groups each) and 13 clusters from the non-experimental one.

<sup>3</sup> The 40 TNS BCs involved in the impact evaluation committed to work as hard as possible on keeping IPA-registered youth engaged. However, they were also allowed to re-mobilize participants in whatever strategy they saw fit in order to reach their target of two training groups of minimum 32 students. In addition, 4 BCs worked outside of the experimental sample either because they were assigned to areas in which the mobilization failed or because, for personal reasons, their work location was not flexible and, hence, were spared from being part of the randomized evaluation.

We plan to include a random sample of 1,100 of these BPC youth in the endline survey. Given the overall with main sample, we expect the total sample at endline will include 3,800 people.

For tests of randomization balance, we check balance of the following variables: household demographics and composition, involvement in income generating activities and income, locus of control, as well as questions about motivation and obstacles for attending the training.

## **2. Key data sources**

Data for the study come from both individual and household surveys and administrative attendance data provided by TNS.

In June 2017, participants who signed up for the program also hand-filled a brief questionnaire which collected baseline information of respondents. This information includes key socio-economic characteristics of respondents, such as basic demographics, living situation, involvement in different income generating activities, income and its sources, and spending patterns and savings. It also includes a series of questions aimed at measuring locus of control, as well as specific questions about motivation for wanting to enroll into the business training program, expected income post-training, and fears about possible challenges to participate in the training.

At the end of August 2017, baseline respondents were re-interviewed with a 2-minute phone-based survey during which they were asked to confirm their identity and contact details, as well as to identify a family member which could have been contacted for an additional household survey<sup>4</sup>.

The household survey took approximately 20 minutes and was conducted via phone<sup>5</sup>. The

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<sup>4</sup> Married people were asked to provide contact details of their spouse, while unmarried individuals were asked to provide contact details of their household head –if different from self.

<sup>5</sup> Calls started at the end of August 2017 and stopped in October, when most of the participants were about to graduate from training and the information that would have been collected could no longer be regarded as baseline data. In such a short period, the data collection team managed to attempt calling 87% of the overall sample: 3,948 respondents, 2,150 from the treatment group and 1,798 from the control. Within each cluster, respondents were randomly sorted to be reached during the follow-up. Each individual from the sorted list was called three times, with calls being made at least one day apart, before being replaced for not responding by the next person on the list. Interviews were considered completed for that cluster once 25 (50 in urban areas) respondents picked-up the call, consented to confirm their identity, and provided the details of a family member to contact for the household survey. Out of 3,948 individuals, the team reached 2,811 respondents -71% of the sample. Out of the 2,811 respondents that picked-up the phone, 2,548 (i.e. 91%) consented to participate in the survey and confirm their identity and 2,192 (78%) provided contact details of a family member. In the short time remaining, the team managed to attempt calling 1,842 family members -84% of the sample of respondents that provided a contact. Out of a sample of 1,842 observations, 1,507 (82%) respondents were reached and 1,425 (77%) consented to participate in the additional household survey.

questionnaire focuses on measuring how individuals interact with their families in order to better understand how families affect the investment decisions of participants. The household survey elicits data regarding key household characteristics, training decisions by youth and their families, expectations about earnings after training, and aspiration levels.

Furthermore, we use administrative data by TNS to determine attendance rates by matching our baseline data and TNS data.

The endline survey will be implemented in the middle of 2019, two years after the end of the STRYDE 2.0 program. It will be collected using face-to-face interviews and will take approximately 1 hour to administer.

### **3. Hypotheses to be tested through the causal chain**

We present our hypotheses to test in this section. We refer to the treatment arm of only receiving the STRYDE 2.0 training as treatment A and to the treatment arm of receiving funds from the BPC as treatment B.

- a.  $H_0/H_a$ : No impact (positive impact) of receiving Treatment A mainly on participants'
  - i Labor force participation and employment,
  - ii Socio-economic status,
  - iii Entrepreneurial and career skills, and
  - iv Psychological and social characteristics.
- b.  $H_0/H_a$ : No impact (positive impact) of receiving Treatment B mainly on participants'
  - i Labor force participation and employment,
  - ii Socio-economic status,
  - iii Entrepreneurial and career skills, and
  - iv Psychological and social characteristics.
- c. Heterogeneity analysis according to the following dimensions
  - i Gender of the participant, and
  - ii The role of the family unit in the ability of youth to take-up the training program, attend classes, graduate, and benefit from it.

#### **4. Measurement of key variables**

The primary outcomes are the following:

- Labor force participation and employment. The relevant measures will include:
  - Economic activities (formal or informal employment status, self-employed in new enterprise, self-employed in pre-existing enterprise, employed or actively looking for a job)
  - Quantity and quality of work (weekly employment hours -unconditional and conditional on working-, low or high skill labor)
  - Duration of work (permanent, seasonal, or short-term)
  - Vulnerability of work (good and bad month sales, months the business was not operational)
  - Business performance (employees, sales, expenses, profits)
- Measures on economic welfare, personal finance and working capital. The relevant measures will include:
  - Cash earnings (TSh/month)
  - Savings behavior (accumulation of savings -dummy, amount of savings held in bank account, in savings circle or club, at home in TSh/month)
  - Borrowing behavior (taken out a loan -dummy, amount, repayment, usage)
  - Household assets
  - Income hiding

The secondary outcomes are the following:

- Entrepreneurial and career skills. The relevant measures will include:
  - Self-confidence with regard to job applications (e.g. ability to find information about job opportunities in the community, CV writing and interviewing skills)
  - Self-confidence with regard to entrepreneurial activities (e.g. run an own business, work in a team to accomplish a task, identify income generating activities to start a new business, obtain credit and use this responsibly)
  - Ability to manage financial accounts and calculate profit and sales
  - Financial numeracy, financial attitudes, and financial awareness

- Psychological and social characteristics. The relevant measures will include:
  - Self-regulation (ability to define goals and to adhere to these goals)
  - Decision making power (including control over monetary resources)
  - Social capital measured by the number of memberships in social networks, leadership position in group, family support
  - Well-being (current status of life satisfaction and in participant's future life -optimism,)
  - Business and personal aspirations
  - Locus of control
  - Grit

## **5. Sample size and power calculations**

The baseline sample for the impact evaluation of the STRYDE 2.0 program includes 135 clusters, 72 treatment and 63 control clusters, with a mean of 33.61 individuals per cluster (and a variance of 7.474) and a total of 4,537 observations.

We consider that program effects of 20-25% will be necessary to justify the costs of the STRYDE 2.0 program. Given the expected take-up rate, the adjusted effect size will be around 11%.

For the endline, we plan to follow-up about 24 individuals per cluster for a total of 3,200 observations in the training sample. Power calculations, conducted in Optimal Design using baseline data, suggest that such a sample would be well powered (attaining a power level of 80%) to detect effect sizes of 11% in a standardized indicator of monthly income.

The experimental sample for the evaluation of the BPC includes 88 clusters for a total of 1,524 observations, plus 264 standard winners. We plan to include 3/4 of them in the endline survey - meaning that in treatment clusters we will track on average 15 respondents, 10 of which should be represented in the baseline sample. Power calculations suggest that a sample with 1,143 observations (3/4 of the experimental sample of 1,524 respondents), 440 treatment and 703 control individuals, would be well powered to detect effect sizes on income of about 11% for the BPC alone.

## 6. Treatment effect equation to be estimated

To test hypotheses outlined in section 3a, we will employ an ANCOVA specification (McKenzie, 2012; Bruhn and McKenzie, 2009). We will be estimating the following simple model using Ordinary Least Squares, which regresses the outcome indicator on the treatment status of individual  $i$  controlling for the baseline value of the indicator:

$$(1) \quad Y_{i\text{ Post}} = \alpha + \beta T_i + \eta Y_{i\text{ Pre}} + \delta X_{i\text{ Pre}} + \varepsilon_{i\text{ Post}}$$

where  $Y_{i\text{ Post}}$  represents the different outcomes of interest as outlined above (3a.i, 3a.ii, 3a.iii, 3a.iv) for individual  $i$ , measured after the intervention.  $T_i$  is a dummy variable equal to 1 if the individual was registered in a cluster randomly selected to receive the STRYDE 2.0 training and 0 if not.  $X_{i\text{ Pre}}$  represents control variables unbalanced at baseline. The estimation includes BC fixed effects since the randomization was implemented within BC strata. Finally,  $\varepsilon_{i\text{ Post}}$  represents the unobserved individual-specific residual. Standard errors will be adjusted for clustering at the ward / village level. The Intent-To-Treat effect of the program will be estimated by  $\beta$ .

We then assess the effects of the BPC cash intervention (see 3b) by modifying equation (1) as follows:

$$(2) \quad Y_{i\text{ Post}} = \alpha + \beta C_i + \eta Y_{i\text{ Pre}} + \delta X_{i\text{ Pre}} + \varepsilon_{i\text{ Post}}$$

where  $C_i$  is a binary variable equal to 1 if the individual from the treatment group was randomly assigned to receive the cash of the BPC honorable mention prizes and 0 if not, while other variables are the same as those defined in equation (1).

To test hypotheses outlined in 3c, we next conduct a heterogeneity analyses that allows estimation of the impact according to gender and family background. Heterogeneous treatment effects will be obtained by estimating (1) and (2) with an additional interaction effect that interacts treatment status with the variable of interest, as illustrated in (3) and (4), respectively:

$$(3) \quad Y_{i\text{ Post}} = \alpha + \beta_\tau \tau_i + \beta_{1 \times \tau} (T_i \times \tau_i) + \beta_1 T_i + \eta Y_{i\text{ Pre}} + \delta X_{i\text{ Pre}} + \varepsilon_{i\text{ Post}}$$

$$(4) \quad Y_{i\text{ Post}} = \alpha + \beta_\tau \tau_i + \beta_{1 \times \tau} (C_i \times \tau_i) + \beta_1 C_i + \eta Y_{i\text{ Pre}} + \delta X_{i\text{ Pre}} + \varepsilon_{i\text{ Post}}$$

where the variable  $\tau_i$  either indicates gender or the role of the family unit.

In addition, we will test the robustness of results from (1) by re-estimating the treatment effect of the STRYDE 2.0 training with a standard difference-in-differences (DiD) estimation. The approach controls for time-invariant unobserved differences between groups. We re-test the hypotheses outlined in 3a as following:

$$(5) \quad Y_{it} = \alpha + \beta_1 T_i + \beta_2 Post_t + \beta_3 (T_i \times Post_t) + \delta X_{it} + \varepsilon_{it}$$

where  $\beta_3$  is the DiD estimator for the impact of the intervention of a treated individual  $i$ .

## **7. Plan for how to deal with multiple outcomes and multiple hypothesis testing**

We have a relatively rich set of outcome measures to explore treatment effects along various interesting dimensions. To deal with multiple hypothesis testing, we will employ three different approaches.

First, we will group our outcome measures into domains where items within a domain are measuring an underlying common factor. The domains we are interested in including are: entrepreneurial and career skills, soft skills, labor market outcomes, and socio-economic status.

Second, on top of the standard model described in the section above, within each domain and across domain indexes, we will also calculate the Family-Wise Error Rate (FWER) adjusted p-values using the Westfall and Young step-down resampling method. The FWER represents the probability that at least one hypothesis out of a family of hypotheses is falsely rejected (type-1 error). Hence, the FWER results will be used to account for the multiple inference problem which increases the likelihood that some of the outcomes are statistically significant by chance even if there is no treatment effect.

Third, to account for multiple hypotheses testing by analyzing the treatment effect heterogeneity, we will use the method to minimize the false non-discovery rate (Benjamini and Hochberg, 1995; Fink et al., 2014; List et al., 2016).

## **8. Procedures to be used for addressing survey attrition, outliers and missing data**

### **8.1. Attrition**

We plan to use an extensive tracking exercise after the conclusion of the standard survey. We will perform a probability reweighting of the data and calculate the effective survey follow-up rate. To understand the drivers of attrition, we will estimate the following equation:

$$(6) \quad AT_{i\ Post} = \alpha + \beta T_i + \eta Y_{i\ Pre} + \delta X_{i\ Pre} + \varepsilon_{i\ Post}$$

where  $AT_i$  is an attrition indicator and the other variables are the same as those defined in equation (1). If the treatment status variable is significant at the 5% level, we will also perform a bounding exercise as suggested in Lee (2009). This semi-parametric approach relies on relatively weak assumptions about how a randomly assigned treatment influences outcomes of interest to obtain intervals on the estimated size of the treatment effect in the presence of non-random attrition.

## **8.2. Procedures to be used for addressing outliers**

At endline, monetary values will be top-censored at the 99th percentile to contain outliers.

## **8.3. Procedures to be used for addressing missing covariate values**

We will follow Lin and Green (2016) in treating missing covariates. If no more than 10 percent of the covariate's values are missing, we will recode the missing values to the overall mean (again testing sensitivity of estimates to these approaches by comparing results with those obtained from the sample with non-missing covariates). If more than 10 percent of the covariate's values are missing, we will include a missingness dummy as an additional covariate and recode missing values to 0.

## **8.4. Procedures to be used for addressing missing dependent variables**

To deal with missing values on our outcome measures, we will adopt the approach described in Kling et al. (2007) and impute missing values by setting them equal to the mean of the respective outcome variable for the relevant treatment group, and testing sensitivity of main coefficient estimates to this approach by comparing results with those obtained from the sample with non-missing outcome variables.

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