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1 INTRODUCTION

<u>Performance Monitoring for Action (PMA)</u> uses innovative mobile technology to support low-cost, rapid-turnaround surveys that monitor key health and development indicators.

PMA surveys collect longitudinal data throughout a country at the household and health facility levels by female data collectors, known as resident enumerators, using mobile phones. The survey collects information from the same women and households over time for regular tracking of progress and for understanding the drivers of contraceptive use dynamics. The data are rapidly validated, aggregated, and prepared into tables and graphs, making results quickly available to stakeholders. PMA surveys can be integrated into national monitoring and evaluation systems using a low-cost, rapid-turnaround survey platform that can be adapted and used for various health data needs.

The PMA project is implemented by local partner universities and research organizations who train and deploy the cadres of female resident enumerators.

The purpose of this manual is to provide guidance on the analysis of harmonized panel data for women age 15-49 surveyed by PMA and published in partnership with <u>IPUMS PMA</u>. IPUMS provides census and survey products from around the world in an integrated format, making it easy to compare data from multiple countries. IPUMS PMA data are

PMA has also published a guide to cross-sectional analysis in both English and French.

available free of charge, subject to terms and conditions: please <u>register here</u> to request access to the data featured in this guide.

The contents of this guide are also available on the IPUMS PMA <u>data analysis blog</u>, which features coding examples in \underline{R} posted every two weeks.

1.1 BACKGROUND

Dating back to 2013, the original PMA survey design included high-frequency, **cross-sectional** samples of women and service delivery points collected from eleven countries participating in <u>Family Planning 2020</u> (FP2020) - a global partnership that supports the rights of women and girls to decide for themselves whether, when, and how many children they want to have. These surveys were designed to monitor annual progress towards <u>FP2020 goals</u> via population-level estimates for several <u>core indicators</u>.

Beginning in 2019, PMA surveys were redesigned under a renewed partnership called <u>Family Planning 2030</u> (FP2030). These new surveys have been refocused on reproductive and sexual health indicators, and they feature a **longitudinal panel** of women of childbearing age. This design will allow researchers to measure contraceptive dynamics and changes in women's fertility intentions over a **three year period** via annual in-person interviews.¹

Questions on the redesigned survey cover topics like:

- awareness, perception, knowledge, and use of contraceptive methods
- perceived quality and side effects of contraceptive methods among current users
- birth history and fertility intentions
- aspects of health service provision
- domains of empowerment

¹In addition to these three in-person surveys, PMA also conducted telephone interviews with panel members focused on emerging issues related to the COVID-19 pandemic in 2020. These telephone surveys are already available for several countries - see our series on PMA COVID-19 surveys for details.

1.2 SAMPLING

PMA panel data includes a mixture of **nationally representative** and **sub-nationally representative** samples. The panel study consists of three data collection phases, each spaced one year apart.

As of this writing, IPUMS PMA has released data from the first *two* phases for four countries where Phase 1 data collection began in 2019; IPUMS PMA has released data from only the *first* phase for three countries where Phase 1 data collection began in August or September 2020. Phase 3 data collection and processing is currently underway.

Now Available from IPUMS PMA

Sample	Phase 1 Data Collection*	Phase 1	Phase 2	Phase 3
Burkina Faso	Dec 2019 - Mar 2020	X	X	
Cote d'Ivoire	Sep 2020 - Dec 2020	Χ		
DRC - Kinshasa	Dec 2019 - Feb 2020	Χ	Χ	
DRC - Kongo Central	Dec 2019 - Feb 2020	Χ	Χ	
India - Rajasthan	Aug 2020 - Oct 2020	Χ		
Kenya	Nov 2019 - Dec 2019	Χ	Χ	
Nigeria - Kano	Dec 2019 - Jan 2020	Χ	Χ	
Nigeria - Lagos	Dec 2019 - Jan 2020	Χ	Χ	
Uganda	Sep 2020 - Oct 2020	Χ		

^{*}Each data collection phase is spaced one year apart

PMA uses a multi-stage clustered sample design, with stratification at the urban-rural level or by sub-region. Sample clusters - called <u>enumeration areas</u> (EAs) - are provided by the national statistics agency in each country. These EAs are sampled using a *probability proportional to size* (PPS) method relative to the population distribution in each stratum.

At Phase 1, 35 household dwellings were selected at random within each EA. Resident enumerators visited each dwelling and invited one household member to complete a <u>Household Questionnaire</u>³ that includes a census of all household members and visitors who stayed there during the night

Questionnaires are administered inperson by resident enumerators visiting selected households in each EA.

These are typically women over age 21 living in (or near) each EA and who hold at least a high school diploma.

²<u>Displaced GPS coordinates</u> for the centroid of each EA are available for most samples <u>by request</u> from PMA. IPUMS PMA provides shapefiles for PMA countries <u>here</u>.

 $^{^{3}}$ Questionnaires administered in each country may vary from this Core Household Questionnaire - <u>click here</u> for details.

before the interview. Female household members and visitors aged 15-49 were then invited to complete a subsequent Phase 1 Female Questionnaire. 4

One year later, resident enumerators visited the same dwellings and administered a Phase 2 Household Questionnaire. A panel member in Phase 2 is any woman still age 15-49 who could be reached for a second Female Questionnaire, either because:

- she still lived there, or
- she had moved elsewhere within the study area, ⁵ but at least one member of the Phase 1 household remained and could help resident enumerators locate her new dwelling. 6

Additionally, resident enumerators administered the Phase 2 Female Questionnaire to *new* women in sampled households who:

- reached age 15 after Phase 1
- joined the household after Phase 1
- declined the Female Questionnaire at Phase 1, but agreed to complete it at Phase 2

When you select the new **Longitudinal** sample option at checkout, you'll be able to include responses from every available phase of the study. These samples are available in either "long" format (responses from each phase will be organized in separate rows) or "wide" format (responses from each phase will be organized in columns).

SAMEDWELLING indicates whether a Phase 2 female

respondent resided in her Phase 1 dwelling or a new

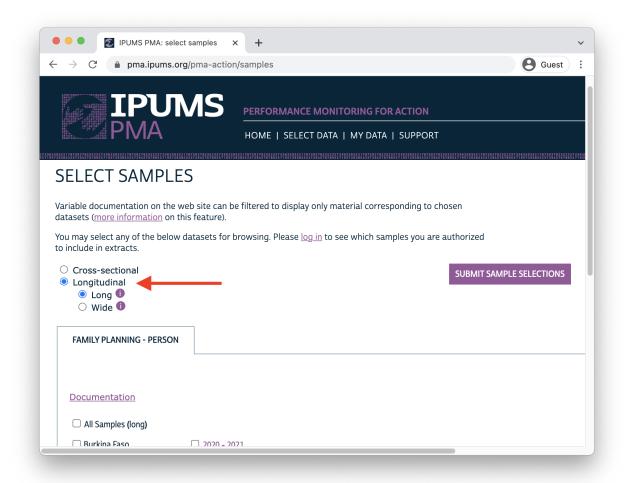
PANELWOMAN

indicates whether a Phase 2 household member completed the Phase 1 Female Questionnaire.

 $^{^4}$ Questionnaires administered in each country may vary from this Core Female Questionnaire - click here for details.

 $^{^{5}}$ The "study area" is area within which resident enumerators should attempt to find panel women that have moved out of their Phase 1 dwelling. This may extend beyond the woman's original EA as determined by in-country administrators - see PMA Phase 2 and Phase 3 Survey Protocol for details.

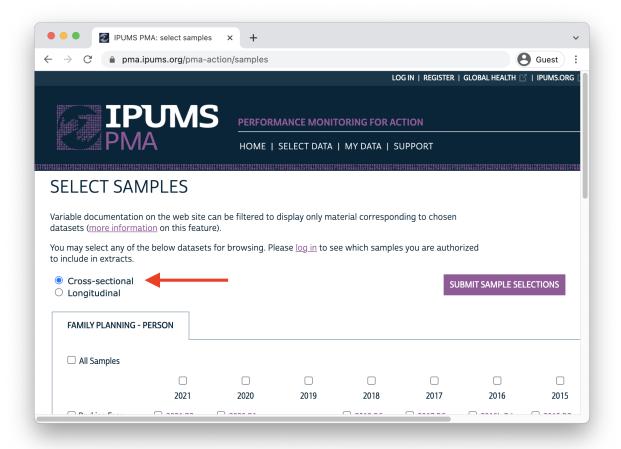
⁶In cases where no Phase 1 household members remained in the dwelling at Phase 2, women from the household are considered lost to follow-up (LTFU). A panel member is also considered LTFU if a Phase 2 Household Questionnaire was not completed, if she declined to participate, or if she was deceased or otherwise unavailable.



In addition to following up with women in the panel over time, PMA also adjusted sampling so that a cross-sectional sample could be produced concurrently with each data collection phase. These samples mainly overlap with the data you'll obtain for a particular phase in the longitudinal sample, except that replacement households were drawn from each EA where more than 10% of households from the previous phase were no longer there. Conversely, panel members who were located in a new dwelling at Phase 2 will not be represented in the cross-sectional sample drawn from that EA. These adjustments ensure that population-level indicators may be derived from cross-sectional samples in a given year, even if panel members move or are lost to follow-up.

CROSS_SECTION indicates whether a household member in a longitudinal sample is also included in the cross-sectional sample for a given year (every person in a cross-sectional sample is included in the longitudinal sample).

You'll find PMA cross-sectional samples dating back to 2013 if you select the **Cross-sectional** sample option at checkout.



1.3 SURVEY DESIGN ELEMENTS

Throughout this guide, we'll demonstrate how to incorporate PMA sampling weights and information about its stratified cluster sampling procedure into your analysis. To do so, we'll rely on tools from the srvyr package. To do so, we'll a supplied the strategies of the strat

Whether you intend to work with a new **Longitudinal** or **Cross-sectional** data extract, you'll find the same set of sampling weights available for all PMA Family Planning surveys dating back to 2013:

- <u>HQWEIGHT</u> can be used to generate cross-sectional population estimates from questions on the Household Questionnaire. 9
- <u>FQWEIGHT</u> can be used to to generate cross-sectional population estimates from questions on the Female Questionnaire. 10
- <u>EAWEIGHT</u> can be used to compare the selection probability of a particular household with that of its EA.

A fourth Family Planning survey weight, <u>POPWT</u>, is currently available only for **Cross-sectional** data extracts and Phase 1 panel data.

Additionally, PMA created a new weight, <u>PANELWEIGHT</u>, which should be used in longitudinal analyses spanning multiple phases, as it adjusts for loss to follow-up. PANELWEIGHT is available only for **Longitudinal** data extracts.

For example, suppose we wanted to estimate the proportion of reproductive age women in Burkina Faso who were using contraception at the time of data collection for both Phase 1 and Phase 2. In a cross-sectional or "long" longitudinal extract, you'll find this information in the variable <u>CP</u>. In a "wide" longitudinal extract, you'll find it in <u>CP_1</u> for Phase 1, and in <u>CP_2</u> for Phase 2. We'll be working with a "wide" extract loaded into R as an object called dat.

Variable names in a "wide" extract have a numeric suffix corresponding with a data collection phase. CP_1 is the Phase 1 version of CP, while CP_2 comes from Phase 2.

dat %>% count(CP_1, CP_2)

⁷The srvyr package is a <u>tidy</u> implementation of the popular <u>survey</u> package for R, authored by Dr. Thomas Lumley. For thorough discussion of the types of weights available in both R and Stata, we recommend <u>this blog post</u> by Dr. Lumley.

⁸POPWT can be used to estimate population-level *counts* - <u>click here</u> or view <u>this video</u> for details.

⁹HOWEIGHT reflects the <u>calculated selection probability</u> for a household in an EA, normalized at the population-level. Users intending to estimate population-level indicators for *households* should restrict their sample to one person per household via <u>LINENO</u> - see <u>household weighting guide</u> for details.

¹⁰ FOWEIGHT adjusts HOWEIGHT for female non-response within the EA, normalized at the population-level - see <u>female</u> <u>weighting guide</u> for details.

The srvyr package provides two functions we'll need to obtain our population estimate. The first, <u>as_survey_design</u>, allows us to specify PANELWEIGHT as a sampling weight. The second, <u>survey_mean</u>, uses that weight in an estimating function; in this case, we'll get the estimated proportion where CP_1 and CP_2 both have the value 1 [Yes] after removing missing / NIU responses with CP_1 < 90 & CP_2 < 90.

In subsequent chapters, we'll use vartype = "ci" to include a 95% confidence interval set by level = 0.95 any time we calculate a population estimate. For discrete variables, we'll also include proportion = TRUE and prop_method = "logit". In practice, there are large number of ways to calculate a confidence interval for a proportion. The <u>srvyr</u> package includes several options for prop_method, but we'll use these settings because:

- 1. they ensure that each proportion's confidence interval only includes values between 0% and 100%.
- 2. they will include the real-world population proportion close to 95% of the time,
- 3. the logit method yields a relatively narrow interval compared with other options, and
- 4. these intervals will match the default intervals reported by Stata and SPSS survey proportion functions.

coef shows the estimated population proportion

_low and _upp show the lower and upper bounds of a 95% confidence interval

¹¹See Dean & Pagano [-@Dean-Pagano] for discussion.

¹²See svyciprop for a complete list of methods.

```
library(srvyr)

dat %>%
    as_survey_design(weight = PANELWEIGHT) %>%
    filter(CP_1 < 90 & CP_2 < 90) %>%
    summarise(
    survey_mean(
        CP_1 * CP_2,
        vartype = "ci",
        level = 0.95,
        proportion = TRUE,
        prop_method = "logit"
    )
)
```

You can also provide information about sample clusters via <u>as_survey_design</u>. In general, we expect households selected from the same EA to share certain characteristics, such that some degree of variation seen in a variable of interest may be non-random at the EA-level. To compensate, you may wish the expand the standard errors produced by survey_mean by providing EA identifiers in <u>EAID</u>.

Here, we include $id = EAID_1$. Compared with our original estimate, notice that the 95% confidence interval for our contraceptive use estimate is wider when we provide information about the clustered sample design - these are "cluster-robust" standard errors.

¹³As we'll see in an upcoming post, women are considered "lost to follow-up" if they moved outside the study area after Phase 1. Therefore, EAID_1 and EAID_2 are identical for all panel members: you can use either one to identify sample clusters.

```
dat %>%
  as_survey_design(weight = PANELWEIGHT, id = EAID_1) %>%
  filter(CP_1 < 90 & CP_2 < 90) %>%
  summarise(
    survey_mean(
        CP_1 * CP_2,
        vartype = "ci",
        level = 0.95,
        proportion = TRUE,
        prop_method = "logit"
    )
)
```

```
# A tibble: 1 × 3
    coef `_low` `_upp`
    <dbl>    <dbl>    <dbl>
1 0.188     0.163     0.215
```

Finally, we'll also use <u>as _survey_design</u> to specify sample strata. For most samples, including Burkina Faso, this information is included in the variable <u>STRATA</u>. We'll include it here with strata = STRATA_1. $\frac{14}{}$

```
dat %>%
  as_survey_design(weight = PANELWEIGHT, id = EAID_1, strata = STRATA_1) %>%
filter(CP_1 < 90 & CP_2 < 90) %>%
summarise(
  survey_mean(
    CP_1 * CP_2,
    vartype = "ci",
    level = 0.95,
    proportion = TRUE,
    prop_method = "logit"
  )
)
```

```
# A tibble: 1 × 3
    coef `_low` `_upp`
    <dbl>    <dbl>    <dbl>
1 0.188     0.164     0.214
```

 $^{^{14}}$ As with <u>EAID</u>, you may use either STRATA_1 or STRATA_2 if your analysis is restricted to panel members.

The variable <u>STRATA</u> is *not available* for samples collected from DRC - Kinshasa or DRC - Kongo Central. If your extract includes any DRC sample, you'll need to amend this variable to include one unique numeric code for each of those regions.

For example, let's look at a different "wide" extract, dat2, containing all of the samples included in this data release. Notice that STRATA_1 lists the sample strata for every <u>COUNTRY</u> except for DRC, where you see the value NA.

dat2 %>% count(COUNTRY, STRATA_1)

```
# A tibble: 27 \times 3
  COUNTRY
                                STRATA 1
                                                                      n
  <int+lbl>
                                 <int+lbl>
                                                                  <int>
 1 1 [Burkina Faso]
                               85401 [Urban, Burkina Faso]
                                                                   3058
                               85402 [Rural, Burkina Faso]
 2 1 [Burkina Faso]
                                                                   2154
 3 2 [Congo, Democratic Republic]
                                                                   3487
                                 40410 [Bungoma - urban, Kenya]
 4 7 [Kenya]
                                                                    153
 5 7 [Kenya]
                                40411 [Bungoma - rural, Kenya]
                                                                    489
                               40412 [Kakamega - urban, Kenya]
 6 7 [Kenya]
                                                                    133
                               40413 [Kakamega - rural, Kenya]
                                                                    438
7 7 [Kenya]
                               40414 [Kericho – urban, Kenya]
                                                                    249
8 7 [Kenya]
9 7 [Kenya]
                               40415 [Kericho - rural, Kenya]
                                                                    453
10 7 [Kenya]
                               40416 [Kiambu – urban, Kenya]
                                                                    214
11 7 [Kenya]
                               40417 [Kiambu - rural, Kenya]
                                                                    311
12 7 [Kenya]
                               40418 [Kilifi – urban, Kenya]
                                                                   170
13 7 [Kenya]
                               40419 [Kilifi - rural, Kenya]
                                                                    455
14 7 [Kenya]
                               40420 [Kitui - urban, Kenya]
                                                                    153
15 7 [Kenya]
                                40421 [Kitui – rural, Kenya]
                                                                    586
16 7 [Kenya]
                               40422 [Nairobi — urban, Kenya]
                                                                   494
17 7 [Kenya]
                                40423 [Nandi - urban, Kenya]
                                                                    260
18 7 [Kenya]
                               40424 [Nandi - rural, Kenya]
                                                                    711
                               40425 [Nyamira - urban, Kenya]
19 7 [Kenya]
                                                                    143
20 7 [Kenya]
                               40426 [Nyamira - rural, Kenya]
                                                                    382
21 7 [Kenya]
                               40427 [Siaya - urban, Kenya]
                                                                    130
22 7 [Kenya]
                               40428 [Siaya - rural, Kenya]
                                                                    437
                               40429 [West Pokot - urban, Kenya] 104
23 7 [Kenya]
                               40430 [West Pokot - rural, Kenya] 474
24 7 [Kenya]
25 9 [Nigeria]
                               56606 [Lagos, Nigeria]
                                                                   1089
26 9 [Nigeria]
                                56611 [Kano - Urban]
                                                                    437
                                 56612 [Kano - Rural]
27 9 [Nigeria]
                                                                    561
```

Now let's see what happens when we try to produce population-level estimates with STRATA_1:

```
dat2 %>%
   as_survey_design(weight = PANELWEIGHT, id = EAID_1, strata = STRATA_1) %>%
   filter(CP_1 < 90 & CP_2 < 90) %>%
   group_by(COUNTRY, GEOCD, GEONG) %>%
   summarise(
    survey_mean(
        CP_1 * CP_2,
        vartype = "ci",
        level = 0.95,
        proportion = TRUE,
        prop_method = "logit"
   )
)
```

```
Error in (function (object, ...) : missing values in `strata'
```

This fails because <u>as _survey _design</u> encounters NA values in STRATA_1. Fortunately, we can replace those values with numeric codes from the variable <u>GEOCD</u>:

```
dat2 %>% count(GEOCD)
```

```
# A tibble: 3 × 2

GEOCD n

<int+lbl> <int>
1 1 [Kinshasa] 1973
2 2 [Kongo Central] 1514
3 NA 14238
```

If GEOCD is not NA, we'll use its numeric code in place of STRATA_1. Otherwise, we'd like to leave STRATA_1 unchanged. However, because both variables include *value labels*, we'll first need remove them with <u>as.numeric</u>. To avoid confusion with the original variable STRATA_1, we'll call our new variable STRATA_RECODE.

```
dat2 <- dat2 %>%
  mutate(
    STRATA_RECODE = if_else(
    is.na(GEOCD),
    as.numeric(STRATA_1),
    as.numeric(GEOCD)
    )
)
dat2 %>% count(GEOCD, STRATA_1, STRATA_RECODE)
```

```
# A tibble: 28 \times 4
   GEOCD
                      STRATA 1
                                                        STRATA RECODE
   <int+lbl>
                      <int+lbl>
                                                                <dbl> <int>
 1 1 [Kinshasa]
                         NA
                                                                    1 1973
 2 2 [Kongo Central]
                                                                    2 1514
                      40410 [Bungoma - urban, Kenya]
                                                                40410
                                                                        153
 4 NA
                      40411 [Bungoma - rural, Kenya]
                                                                40411
                                                                        489
                      40412 [Kakamega - urban, Kenya]
 5 NA
                                                                        133
                                                                40412
                     40413 [Kakamega - rural, Kenya]
 6 NA
                                                                40413
                                                                        438
 7 NA
                     40414 [Kericho - urban, Kenya]
                                                                        249
                                                                40414
8 NA
                     40415 [Kericho - rural, Kenya]
                                                                        453
                                                                40415
9 NA
                     40416 [Kiambu - urban, Kenya]
                                                                40416
                                                                        214
                    40417 [Kiambu - rural, Kenya]
10 NA
                                                                        311
                                                                40417
11 NA
                    40418 [Kilifi – urban, Kenya]
                                                                        170
                                                                40418
12 NA
                     40419 [Kilifi - rural, Kenya]
                                                                40419
                                                                        455
13 NA
                    40420 [Kitui – urban, Kenya]
                                                                40420
                                                                        153
14 NA
                     40421 [Kitui - rural, Kenya]
                                                                40421
                                                                        586
15 NA
                    40422 [Nairobi - urban, Kenya]
                                                                40422
                                                                       494
16 NA
                     40423 [Nandi – urban, Kenya]
                                                                        260
                                                                40423
17 NA
                    40424 [Nandi - rural, Kenya]
                                                                40424
                                                                        711
18 NA
                     40425 [Nyamira - urban, Kenya]
                                                                40425
                                                                        143
19 NA
                    40426 [Nyamira - rural, Kenya]
                                                                40426
                                                                        382
20 NA
                     40427 [Siaya - urban, Kenya]
                                                                40427
                                                                        130
21 NA
                    40428 [Siaya - rural, Kenya]
                                                                40428
                                                                        437
22 NA
                     40429 [West Pokot - urban, Kenya]
                                                                40429
                                                                        104
                     40430 [West Pokot - rural, Kenya]
23 NA
                                                                40430
                                                                        474
24 NA
                      56606 [Lagos, Nigeria]
                                                                56606 1089
25 NA
                      56611 [Kano - Urban]
                                                                56611
                                                                        437
                      56612 [Kano - Rural]
26 NA
                                                                56612
                                                                        561
27 NA
                      85401 [Urban, Burkina Faso]
                                                                85401 3058
28 NA
                      85402 [Rural, Burkina Faso]
                                                                85402 2154
```

Now, we can use STRATA_RECODE with <u>as_survey_design</u> to obtain population estimates for each nationally representative or sub-nationally representative sample.

```
# A tibble: 6 \times 6
# Groups: COUNTRY, GEOCD [5]
 COUNTRY
                                                           coef `_low` `_upp`
                               GE0CD
                                                 GEONG
 <int+lbl>
                               <int+lbl>
                                                 <int+lbl> <dbl> <dbl> <dbl>
1 1 [Burkina Faso]
                                                          0.188 0.164 0.214
                               NA
                                                 NA
                                                          0.320 0.288 0.353
2 2 [Congo, Democratic Republic] 1 [Kinshasa]
                                                NA
3 2 [Congo, Democratic Republic] 2 [Kongo Central] NA
                                                           0.268 0.215 0.329
4 7 [Kenya]
                                                           0.366 0.350 0.382
                                                NA
                               NA
5 9 [Nigeria]
                               NA
                                                 2 [Lagos] 0.293 0.259 0.330
6 9 [Nigeria]
                                                 4 [Kano] 0.0537 0.0322 0.0880
                               NA
```

1.4 INCLUSION CRITERIA FOR ANALYSIS

The remainder of this guide will feature code you can use to reproduce key indicators included in the PMA Longitudinal Brief for each sample. In many cases, you'll find separate reports available in English and French, and for both national and sub-national summaries. For reference, here are the highest-level population summaries available in English for each sample where Phase 2 IPUMS PMA data is currently available:

- Burkina Faso
- DRC Kinshasa
- DRC Kongo Central
- Kenya
- Nigeria Kano
- Nigeria Lagos

Panel data in these reports is limited to the *de facto* population of women who completed the Female Questionnaire in both Phase 1 and Phase 2. This includes women who slept in the household during the night before the interview for the Household Questionnaire. The *de jure* population includes women who are usual household members, but who slept elsewhere that night. We'll remove *de jure* cases recorded in the variable <u>RESIDENT</u>.

For example, returning to our "wide" data extract for Burkina Faso, you can see the number of women who slept in the household before the Household Questionnaire for each phase reported in RESIDENT_1 and RESIDENT_2:

```
dat %>% count(RESIDENT_1)
```

NA cases in RESIDENT_2 represent women who were lost to follow-up in Phase 2. In Stata, these cases will be represented by blank values.

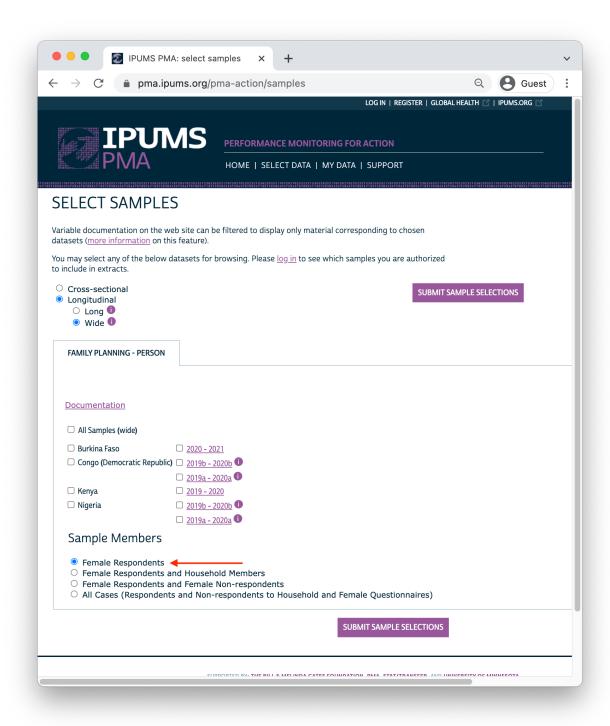
```
dat %>% count(RESIDENT 2)
```

The *de facto* population is represented in codes 11 and 22. We'll use filter to include only those cases.

```
dat_2 <- dat %>%
    filter(
        RESIDENT_1 == 11 | RESIDENT_1 == 22,
        RESIDENT_2 == 11 | RESIDENT_2 == 22
    )

dat_2 %>% count(RESIDENT_1, RESIDENT_2)
```

Additionally, these reports only include women who completed (or partially completed) both Female Questionnaires. This information is reported in RESULTFQ. In our "wide" extract, this information appears in RESULTFQ_1 and RESULTFQ_2: if you select the "Female Respondents" option at checkout, only women who completed (or partially completed) the Phase 1 Female Questionnaire will be included in your extract.



We'll further restrict our sample by selecting only cases where RESULTFQ_2 shows that the woman also completed the Phase 2 questionnaire. Notice that, in addition to each of the value 1 through 10, there are several **non-response codes** numbered 90 through 99. You'll see similar values repeated across all IPUMS PMA variables, except that they will be left-padded to

match the maximum width of a particular variable (e.g. 9999 is used for <u>INTFQYEAR</u>, which represents a 4-digit year for the Female Interview).

```
dat %>% count(RESULTFQ 2)
```

```
# A tibble: 11 \times 2
  RESULTFQ 2
 <int+lbl>
                                                 <int>
1 1 [Completed]
                                                  5491
2 2 [Not at home]
                                                    78
3 3 [Postponed]
                                                    22
4 4 [Refused]
                                                    66
5 5 [Partly completed]
                                                    12
6 7 [Respondent moved]
                                                    15
7 10 [Incapacitated]
                                                    19
8 95 [Not interviewed (female questionnaire)]
                                                   4
9 96 [Not interviewed (household questionnaire)]
                                                   192
10 99 [NIU (not in universe)]
                                                   399
11 NA
                                                   492
```

Possible **non-response codes** include:

- 95 Not interviewed (female questionnaire)
- 96 Not interviewed (household questionnaire)
- 97 Don't know
- 98 No response or missing
- 99 NIU (not in universe)

The value NA in an IPUMS extract indicates that a particular variable is not provided for a selected sample. In a "wide" **Longitudinal** extract, it may also signify that a particular person was not included in the data from a particular phase. Here, an NA appearing in RESULTFQ_2 indicates that a Female Respondent from Phase 1 was not found in Phase 2.

You can drop incomplete Phase 2 female responses as follows:

```
dat_3 <- dat %>% filter(RESULTFQ_2 == 1)
dat_3 %>% count(RESULTFQ_1, RESULTFQ_2)
```

```
# A tibble: 2 × 3
RESULTFQ_1 RESULTFQ_2 n
<int+lbl> <int+
1 1 [Completed] 1 [Completed] 5487
2 5 [Partly completed] 1 [Completed] 4</pre>
```

Generally, we will combine both filtering steps together in a single function like so:

```
dat <- dat %>%
  filter(
    RESIDENT_1 == 11 | RESIDENT_1 == 22,
    RESIDENT_2 == 11 | RESIDENT_2 == 22,
    RESULTFQ_2 == 1
)
```

In subsequent analyses, we'll use the remaining cases to show how PMA generates key indicators for **contraceptive use status** and **family planning intentions and outcomes**. The summary report for each country includes measures disaggregated by demographic variables like:

- MARSTAT marital status
- EDUCATT and EDUCATTGEN highest attended level of education 15
- <u>AGE</u> age¹⁶
- WEALTHQ and WEALTHT household wealth quintile or tertile 17
- <u>URBAN</u> and <u>SUBNATIONAL</u> geographic location 18

¹⁵ Levels in EDUCATT may vary by country; EDUCATTGEN recodes country-specific levels in four general categories.

¹⁶Ages are frequently reported in five-year groups: 15-19, 20-24, 25-29, 30-34, 35-39, 40-44, and 45-49.

¹⁷Households are divided into quintiles/tertiles relative to the distribution of an asset <u>SCORE</u> weighted for all sampled households. For subnationally-representative samples (DRC and Nigeria), separate wealth distributions are calculated for each sampled region

¹⁸SUBNATIONAL includes subnational regions for all sampled countries; country-specific variables are also available on the <u>household - geography</u> page.