Global Studies

http://open.bu.edu

BU Open Access Articles

2022-12

Measurement of unmet need for contraception: a counterfactual approach

This work was made openly accessible by BU Faculty. Please share how this access benefits you. Your story matters.

Version	Published version
Citation (published version):	M. Karra. 2022. "Measurement of Unmet Need for Contraception: A Counterfactual Approach." Studies in Family Planning, Volume 53, Issue 4, pp.657-680. https://doi.org/10.1111/sifp.12216

https://hdl.handle.net/2144/46044 Boston University



Measurement of Unmet Need for Contraception: A Counterfactual Approach

Mahesh Karra 🝺

Unmet need plays a critical role in reproductive health research, evaluation, and advocacy. Although conceptually straightforward, its estimation suffers from a number of methodological limitations, most notably its reliance on biased measures of women's stated fertility preferences. We propose a counterfactual-based approach to measuring unmet need at the population level. Using data from 56 countries, we calculate unmet need in a population as the difference between: (1) the observed contraceptive prevalence in the population; and (2) the calculated contraceptive prevalence in a subsample of women who are identified to be from "ideal" family planning environments. Women from "ideal" environments are selected on characteristics that signal their contraceptive autonomy and decision-making over family planning. We find significant differences between our approach and existing methods to calculating unmet need, and we observe variation across countries when comparing indicators. We argue that our indicator of unmet need is preferable to existing population-level indicators due to its independence from biases that are generated from the use of reported preference measures, the simplicity with which it can be derived, and its relevance for cross-country comparisons as well as context-specific analyses.

INTRODUCTION

An estimated 40 percent of pregnancies, or 99 million pregnancies, each year are unintended (Bearak et al. 2018; Singh 1998; Darroch, Sedgh, and Ball 2011), either because they are unwanted or mistimed at the time of conception (Brown and Eisenberg 1995). The use of contraception may help women and couples to meet their desired fertility and to avert unintended

Mahesh Karra, is an Assistant Professor of Global Development Policy at the Frederick S. Pardee School of Global Studies, Boston University, 152 Bay State Road, Room G04C, Boston, MA, 02215, USA. E-mail: mvkarra@bu.edu.

^{© 2022} The Authors. *Studies in Family Planning* published by Wiley Periodicals LLC on behalf of Population Council.

This is an open access article under the terms of the Creative Commons Attribution-NonCommercial License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited and is not used for commercial purposes.

pregnancies and unwanted births (Bongaarts, Mauldin, and Phillips 1990; Lloyd and Ross 1992). However, up to 215 million women, or 26 percent of sexually active women of reproductive age (WRA), are not using a contraceptive method even when they want to avoid becoming pregnant—these women account for an estimated 82 percent of all unintended pregnancies (Darroch, Sedgh, and Ball 2011; Darroch et al. 2017).

A high-quality family planning program is not only determined by the achievement of good reproductive health outcomes but also prioritizes helping women and couples maximize a complex and evolving set of preferences around future fertility, health, and well-being. For this reason, the demand for (and use of) contraception differs from most other interventions in health; while one can assume that individuals have a demand for health interventions that reduce their risk of morbidity and mortality, the same cannot be said for the demand for contraception since women and couples may seek to become pregnant over their lifetimes (Cleland, Harbison, and Shah 2014). As a result, it has become incumbent on family planning and reproductive health programs to demonstrate that a demand for contraception and family planning exists, and to measure the extent to which this demand for contraception is met or remains unmet.

The commitment to effectively quantify and meet demand for family planning has been enshrined in several global agendas, most recently (and notably) as a key target (target 3.7) in the 2030 Sustainable Development Goals (SDGs; United Nations 2018). A key measure of progress to achieving target 3.7 of the SDGs is indicator 3.7.1, which is calculated as the proportion of WRA (15–49 years) who have their need, or demand, for family planning satisfied by using modern methods of contraception. Conversely, unmet need, which aims to estimate the proportion of women who want to delay or stop childbearing but are **not** using contraception, plays an equally fundamental role in family planning research, evaluation, and advocacy and has received significant attention from scholars from a range of fields, from human rights and reproductive justice to economics and demography (Cleland, Harbison, and Shah 2014).

Unmet Need: Current Definition and Measurement Challenges

Although the underlying concept of unmet need, the nonuse of contraception among women stating a desire to avoid pregnancy, appears to be straightforward, its operationalization is problematic and complex and has undergone multiple revisions in recent decades (Cleland, Harbison, and Shah 2014; Bradley and Casterline 2014). In its latest iteration, unmet need is calculated as the proportion of fecund and sexually active WRA who want to either limit or space their next birth for at least two years but are not using any contraceptive method (Bradley et al. 2012). Although this revision is a significant simplification from previous versions, its estimation still requires up to 15 items from survey responses are needed to capture a range of indicators related to: (1) a woman's potential exposure to the risk of pregnancy; (2) her sexual activity; (3) her physiological capacity to become pregnant (fecundity); and (4) the reliability of a woman's retrospective reporting of her preferences to space and limit births (Bradley et al. 2012; Bradley and Casterline 2014; Cleland, Harbison, and Shah 2014).

The current measure of unmet need is calculated as follows:

Unmet Need = $\frac{\text{WRA who want to limit/space births for 2 + years AND are not using contraception}}{\text{Fecund and sexually active WRA (ages 15 - 49)}}$.





In this measure, the denominator aims to capture the population of women who would be at risk of pregnancy and includes women who: (1) are either married or are in a sexual union; (2) report being sexually active; and (3) are fecund, and are therefore at risk of becoming pregnant. Among this population, women are classified into: (a) current contraceptive users, composed of women who either have a "met need for limiting births" or a "met need for spacing births," or (b) nonusers of contraception, composed of nonpregnant, currently pregnant, or postpartum amenorrheic women who are classified to either have an "unmet need for limiting births," an "unmet need for spacing births," or "no unmet need." The categorization of women into met need, unmet need, or no unmet need, and hence their relative contribution to the numerator, is a function of women's reported preferences to space or limit future births (in the case of women who are not pregnant or postpartum amenorrheic) or of women's retrospective preferences to space or limit their current (if pregnant) or most recent (if postpartum amenorrheic) birth. Figure 1 presents a flow diagram of the classification algorithm (Bradley et al. 2012).

Over the years, a number of methodological concerns related to the operationalization of unmet need in survey data have been highlighted by scholars and practitioners alike. One key issue is the reliability of women's reported or assumed sexual activity, which serves as an indicator of exposure to the risk of pregnancy. For example, currently married women are assumed to be sexually active and exposed to the risk of pregnancy even if they report not using contraception because their partners are away or because they have no or infrequent sex. Including these women in the calculation of unmet need may therefore result in an overestimation of the measure (Bradley and Casterline 2014; Bradley et al. 2012). On the other hand, excluding unmarried but (potentially) sexually active women who demand contraception from the calculation would underestimate unmet need; Bradley and Casterline (2014) find a 3.4-percentage-point (equivalent to a 16 percent) average decline in unmet need when these women are excluded from the unmet need calculation.

Another important issue is the inclusion of pregnant and postpartum amenorrheic women, many of whom might soon demand contraception following their transition out of their temporary state of insusceptibility to pregnancy. The length of postpartum amenorrhea, during which time a woman is free from the risk of pregnancy, continues to be a source of debate, and sensitivity analyses show that reducing the length of time during which postpartum amenorrhea is a reliable signal of inability to conceive from 24 months to 6 months has significant impacts on the range of estimates of unmet need, ranging from a 0.7-percentage-point to a 6.4-percentage-point (20 percent) increase in unmet need (Bradley and Casterline 2014).

A third methodological concern involves the identification and exclusion of infecund women, whose contraceptive use or nonuse are independent from their risk of pregnancy. To this end, the measurement of fecundity from behavioral responses, and in the absence of biological indicators, is challenging and relies on questionable assumptions. Specifically, the revised algorithm assumes that women are infecund if they satisfy at least one of three criteria: (1) they first married five or more years ago, have not had a birth in the past five years, and have never used contraception; (2) they report having menstruated in the last 6 months and are not postpartum amenorrheic; or (3) they report that they are not able to become pregnant, are menopausal, or have had a hysterectomy (Bradley et al. 2012). Evidence from other studies have shown that the potential misclassification of women who may, in fact, be able to conceive has a substantial impact on the measurement of unmet need, whereby unmet need would increase by an average of 3.7 percentage points if infecund women were to be eligible to have an unmet need (Bradley and Casterline 2014).

Perhaps the most problematic feature of the current measure, however, is its reliance on women's reported fertility preferences, and particularly the measurement of women's wantedness of births through direct retrospective recall. This recall is ascertained by asking women "At the time you became pregnant with [name of the most recent birth], did you want to become pregnant then, did you want to wait until later, or did not want (more) children at all?" This approach clearly suffers from the ex-post rationalization bias that is present in women's reluctance to declare a past pregnancy or birth as unwanted, and particularly when the past birth of interest refers to a child who is alive at the time of the interview (Bongaarts 1990; Casterline and El-Zeini 2007; Casterline 2009). Many studies have demonstrated the significant bias of this approach to eliciting a woman's preferences and have proposed alternative measures for identifying fertility preferences, including:

- (1) Eliciting a woman's stated ideal number of children that she would want over her lifetime if she could go back to the time when she did not have children.
- (2) Eliciting fertility preferences prospectively using prospectively oriented questions (Although "Would you like to have (a/another) child, or would you prefer not to have

any (more) children?") in either cross-sectional or, preferably, longitudinal surveys where respondents are repeatedly interviewed. Although both alternatives have certain advantages, each approach falls well short of its goal to effectively and unbiasedly measure women's fertility preferences. In particular, the direct elicitation of a woman's ideal number of children, for the same reason as the retrospective recall approach, is limited in that women are likely to ex-post rationalize their past births and are therefore unlikely to report an ideal number of children that is less than their current number of living children. Moreover, empirical evidence on the measurement of this variables has shown that a considerable proportion of survey respondents are either unsure about their ideal number of children or do not provide a numeric response to the question (Casterline and El-Zeini 2007). Although arguments have been made to use prospective measures of stated preferences and intentions, these measures are typically elicited through a single question: "Do you want another child?" As Müller et al. (2022) note, the presentation of this question to respondents, without a temporal referent or presentation of additional alternative counterfactual futures in which childbearing may be more or less desirable, may lead to incorrect inference when directly linking this response to subsequent behavior. This may particularly be the case for women whose observed fertility behavior appears to be "inconsistent" based on the data but who may have, in fact, succeeding in fulfilling their latent fertility preferences, which are outside the frame of the researchers' scope of inquiry (Müller et al. 2022). On the other hand, the elicitation of preferences for future births using prospective

questions and longitudinal data methods can be appealing for its forward-looking approach (thereby eliminating any biases induced by retrospective inquiry) and in its potential to infer women's preferences for births occurring between survey rounds. Unfortunately, conducting longitudinal data collection is costly, requires tracking and follow-ups with respondents, and suffers from a new set of empirical concerns that limits inference, including attrition and nonresponse between waves, time-in-sample bias, compounded mismeasurement and selection bias, and so on (Hsiao 1985). For these reasons, most large-scale surveys of fertility and health (e.g., Demographic and Health Surveys [DHS], Multiple Indicator Cluster Surveys [MICS], World Fertility Surveys [WFS], etc.) are cross-sectional. Conceptually, the use of longitudinal measures also assumes that fertility preferences are stable over time and particularly between survey rounds; the stability of fertility preferences over time has been questioned, with studies demonstrating that both women's contraceptive and fertility preferences are likely to be malleable and unstable over relatively short intervals (Trinitapoli and Yeatman 2018; Karra and Zhang 2021; Müller et al. 2022).

A review of the literature on the links between prospective fertility preferences and behavior found the following (Cleland, Machiyama, and Casterline 2020):

- (1) Fertility preferences seem to be correlated to subsequent fertility behavior but mainly among women who state a preference for wanting to have another child, but significantly less so among women who state a preference for limiting childbearing.
- (2) Evidence of the relationship between baseline fertility preferences for spacing and postponement and subsequent fertility behavior is weak and mixed.

(3) There is significant uncertainty in women's reported stated fertility preferences and considerable variation in preference stability and strength across context and over time.

Taken together, the findings from the review further raise questions as to whether static stated preferences, as they are currently measured, may be meaningful for inference. More recently, a study by Müller et al. (2022) highlights the complexity and uncertainty with which fertility preferences are shaped, changed, and reconciled over time. Using a multi-year panel sample from Kenya, the authors find: (1) substantial variation in fertility preferences over multiple points in time; (2) the significant extent to which women's current expectations of future fertility outcomes are frequently misperceived; (3) a lack of precision with which women are able to recall their own past fertility intentions and preferences; and (4) the asymmetric nature with which women anchor at their current preferences. Taken together, the authors conclude: "While the illusion of stable and effective preferences was held by many of our research subjects, we see no reason why we as researchers should share this illusion" (p. 186).

More generally, unmet need's reliance on women's stated (reported) preferences as a proxy for their true (revealed) fertility preferences may itself be problematic. One of the main criticisms of using stated preferences is that their measurement typically relies on surveys in which respondents face hypothetical choice problems to elicit individual valuations over alternatives. As a result, respondents may not make the same choices in a hypothetical situation as they would in real life (Ami, Aprahamian, and Luchini 2017). In the case of fertility, this "hypothetical bias" implies that respondents may be willing to state a preference for more or fewer children when asked in a survey than they would if the opportunity to realize this preference were to truly present itself. This bias is generated both from a lack of incentive to tell the truth in a survey and from the difficulty that the respondent faces from projecting herself into a hypothetical situation that may not be directly familiar to her—this is particularly true for eliciting stated preferences when costs or constraints cannot directly be internalized. For example, women who have never been pregnant may be more likely to not internalize the costs of pregnancy and childbearing and may therefore be more likely to misreport (in this case, overstate) their ideal fertility. On the other hand, respondents who identify and internalize the costs of alternatives might narrow their choice set a priori even if these costs are misperceived. For example, women who have experienced difficulty conceiving in the past may anchor their fertility potential to this constraint and may therefore underreport their true desired fertility (Rosenzweig and Wolpin 1993).

The challenges to inferring latent preferences from stated preference data, whether retrospectively or prospectively collected, have been highlighted in studies in behavioral and cognitive science, to the extent that some studies have argued that fertility preferences may not even exist except when respondents are prompted, at which time people instead report on preferences that are constructed from information that is available and salient to them when they are asked (Bachrach and Morgan 2013; Bhrolcháin and Beaujouan 2019). These findings are concordant with a growing literature in behavioral economics and cognitive science on the presence of projection bias, where individuals incorrectly extrapolate the extent to which their future preferences and behavior will resemble their current tastes based on how they believe their tastes will change, even over relatively short periods of time (Loewenstein, O'Donoghue, and Rabin 2003). This bias can be observed in a range of consumer behavior, for example, purchasing more groceries than one eventually would consume when one enters the store hungry, but is understudied in the literature on fertility preferences and behavior. Taken together, while a respondent's stated fertility preferences are likely to be correlated with her true latent preferences, the gap between these preference measures is likely to be significant, to the extent that the utility of the stated preference measure is unclear.

Operationalizing Unmet Need with Counterfactuals: A Step Back and a Way Forward

In light of the conceptual, empirical, and operational challenges to estimating unmet need, combined with the numerous revisions and debates around its validity and usefulness as a measure over the years, it would be incumbent upon the family planning field to take a step back and remind ourselves of its potential utility and aim as a measure. The primary objective of unmet need is to estimate the proportion of women at an aggregate (population)-level who are not using contraception but who have a preference for limiting or spacing births.

Equivalently, unmet need can be understood through the following counterfactual thought experiment. Let us define the *current contraceptive prevalence* CP_{ct} as the proportion of women in a country c at time t (where time, in this study, is indicated by the survey year) who use contraception under the current state of the world.¹ Now, let us define the *ideal* contraceptive prevalence iCP_{ct} as the portion of women in country c at time (survey year) t who use contraception in the state of the world where family planning and fertility preferences in this population can be fully realized without constraint. In this hypothetical state of the world, women would face no barriers, costs, or constraints of any kind to identifying and realizing both their preferences for contraception as well as for limiting and spacing pregnancies over their lifetimes. Features of this state of the world include, but are not limited to: (1) women's ability to completely control their family planning and reproductive health decisions, including full, free, and informed choice over their contraceptive use, nonuse, and type of use (i.e., complete choice over methods and method type; Newman and Feldman-Jacobs 2015; Senderowicz 2020); (2) women's capability to realize any changes to preferences that they make over fertility and childbearing; and (3) a lack of social, structural, emotional, or physical barriers that women face to forming, identifying, and executing their contraceptive and fertility decision-making, with complete support from their partners, families, and communities on all such decisions.

Unmet need for contraception for a country c at survey year t can be simply calculated as the difference between the ideal contraceptive prevalence and the current contraceptive prevalence, that is,

Unmet Need_{ct} = $iCP_{ct} - CP_{ct}$.

In reflecting on this calculation, we recognize that while CP_{ct} is relatively more straightforward to infer with reported survey data, the identification of the ideal contraceptive

¹ Contraceptive prevalence, which is a proportion, is often erroneously labeled as a "rate" (i.e., a "contraceptive prevalence rate" or CPR), when it does not have any reference to a period or interval of time in its denominator, as is required by the standard epidemiological definition of a "rate." To be consistent with established conventions in epidemiology, I therefore identify the proportion of WRA using contraception *at a point in time* (at the time of interview) as the "contraceptive prevalence."

prevalence *iCP_{ct}* is, by construction, a hypothetical measure. To estimate this prevalence, previous estimators of unmet need have relied on first estimating women's latent family planning and fertility preferences, measured with stated preferences, and then inferring the extent to which contraceptive use concords with these preferences. We propose an inverse approach: first, we infer the ideal environment under which all preferences can be realized, and we then estimate the contraceptive prevalence in this environment. This approach hinges on the premise that the contraceptive prevalence under this ideal environment would reflect women's revealed preferences and, by extension, their demand for contraceptive advantage over traditional estimators in that it captures women's level of contraceptive autonomy and decision-making without the need for any direct elicitation or estimation of preferences. Similar approaches have been utilized in the child development literature, where studies have constructed "ideal" reference populations and have conducted comparative analyses that identify gaps in child growth and stunting relative to the reference group (Karra, Subramanian, and Fink 2017).

As an attempt to identify this counterfactual environment, we could imagine that contraceptive prevalence under an "ideal" environment would be the prevalence among the subpopulation of women who are situated in "ideal" conditions in which they have full, free, informed choice over their contraceptive use and are capable of acting on their preferences to the greatest possible extent—this approach broadly speaks to the Sen capability approach to welfare gain and on subsequent developments in women's empowerment in reproductive decision-making (Nussbaum and Sen 1993; Anand, Hunter, and Smith 2005). To identify this "ideal" subpopulation, we narrow down the sample of women based on characteristics that are more likely to signal their level of contraceptive and reproductive autonomy. These observable characteristics can be selected based on the set of determinants been theorized to be correlated with women's contraceptive autonomy, access, and reproductive decision-making. Obvious characteristics for selection include women's socioeconomic status (those from the topmost income or wealth echelons), educational attainment (those who are the most educated), knowledge of family planning (those who are the most informed about contraceptive methods), general autonomy (particularly those who have autonomy to make decisions and seek their own health care), familial and social support (those who have their partner's and community's approval to use/not use contraception), and access (those who are able to receive the full range of contraceptive methods without constraint). These are but a few of the characteristics that would approximate an "ideal" enabling environment for women; however, a key advantage in this approach is that women who live in these selective environments can be identified using routine survey data (e.g., DHS, MICS).

Testing a New Unmet Need Measure: Empirical Evidence

We estimate this new approach to estimating unmet need using data on 2,073,523 women from 80 DHS surveys that cover 56 countries from 2010 to 2019. We then identify the sub-sample of women who meet the following five criteria:

 They belong to the highest wealth quintile, a proxy for their socioeconomic status. Women who belong to this group are less likely to face access or cost constraints and are more likely, in general, able to follow through on their contraceptive preferences.

- (2) They are either currently married or have been sexually active for the past month. These two variables serve as part of the selection criteria that is used to define the population of women who are at risk of pregnancy.
- (3) They have attained at least a tertiary level of schooling, which selects on those women who are less likely to have information or access barriers.
- (4) They know at least one contraceptive method, which also serves as a proxy for being informed about family planning and reproductive health services.
- (5) They do not report distance to a facility as being a significant problem in their access to health care. This measure of perceived access is likely to be correlated with true access and may be more likely to impact a woman's care-seeking behavior.

When filtering the full sample of women by these five criteria, we are left with a sample of 55,318 women from 52 countries across 73 DHS surveys, which constitute 2.71 percent of the full sample of women. Table 1 presents the distribution of women who are selected from ideal environments within each DHS survey.

Table 2 presents descriptive statistics of the full sample and the selected sample of women from ideal environments. We find significant differences between women from the full sample and women who were selected to be from ideal environments. In particular, women from ideal environments are more likely to reside in urban settings (80.7 percent) compared to women in the full sample (37.6 percent); more likely to be older, on average (33.1 years) compared to women in the full sample (29.6 years); have fewer children, on average (1.7 children) than women in the full sample (2.3 children); are married to husbands who are significantly more likely to have a tertiary level of education (74.7 percent) compared to women in the full sample (11.4 percent); and are more likely to earn as much or more than their husbands/partners (37.1 percent) compared to women in the full sample (26.6 percent).

Figures 2 and 3 present estimates for *iCP* by country and estimates for changes in *iCP* over time by country, respectively. Estimates for *iCP* vary significantly by country, ranging from a low of 20 percent in Chad to more than 81 percent in Honduras. Among countries that have multiple DHS survey rounds represented (though we note that no country had more than two surveys), we calculate the changes to *iCP* over time by country. *iCP* varies over time within country; for some countries, we find declines in *iCP* over time by as much as 14 percentage points (p.p.) in Gabon, while *iCP* was found to increase by as much as 21 percent in other countries such as Tanzania.

Table 3 presents descriptive statistics comparing the newly calculated measure of unmet need and the currently used measures of unmet need. Unmet need using the new counterfactual-based measure is, on average, 5–6 percentage points (30 percent) higher than the standard measures of unmet need that are currently used by the DHS. Moreover, we find that the variation in unmet need, as indicated by the standard deviation, is also higher with our new counterfactual measure as compared to traditional measures of unmet need. This implies that the distribution of unmet need measures using the counterfactual approach is wider, yielding more extreme estimates of unmet need on both the lower and higher end;

1	Full N	Full Pct.	Ideal N	Ideal Pct.
Afghanistan	29,461	1.45	214	0.39
Albania	15,000	0.74	429	0.78
Armenia	12,038	0.59	818	1.48
Angola	14,379	0.71	186	0.34
Bangladesh	55,739	2.74	0	0.00
Burkina Faso	17,087	0.84	69	0.12
Benin	32,527	1.60	285	0.52
Burundi	26,658	1.31	167	0.30
DRC	18,827	0.92	191	0.35
Congo, Republic	10,819	0.53	84	0.15
Cote d'Ivoire	10,060	0.49	99	0.18
Cameroon	30,103	1.48	506	0.91
Colombia	92,239	4.53	0	0.00
Dominican Republic	9,372	0.46	481	0.87
Egypt	21,762	1.07	1,548	2.80
Ethiopia	32,198	1.58	744	1.34
Gabon	8,422	0.41	78	0.14
Ghana	9,396	0.46	179	0.32
Gambia	10,233	0.50	127	0.23
Guinea	20,016	0.98	147	0.27
Guyana	25,914	1.27	623	1.13
Honduras	22,757	1.12	449	0.81
Haiti	29,800	1.46	515	0.93
India	699,686	34.34	25,539	46.17
Indonesia	95,234	4.67	4,315	7.80
Jordan	26,041	1.28	1,240	2.24
Kenya	31,079	1.53	490	0.89
Cambodia	36,332	1.78	453	0.82
Comoros	5,329	0.26	115	0.21
Kyrgyz Republic	8,208	0.40	508	0.92
Liberia	9,239	0.45	108	0.20
Lesotho	6,621	0.32	253	0.46
Mali	20,943	1.03	175	0.32
Maldives	7,699	0.38	99	0.18
Malawi	47,582	2.34	473	0.86
Mozambique	13,745	0.67	177	0.32
Nigeria	80,769	3.96	3,350	6.06
Niger	11,160	0.55	54	0.10
Namibia	9,176	0.45	267	0.48
Nepal	25,536	1.25	824	1.49
Philippines	41,229	2.02	2,531	4.58
Pakistan	25,922	1.27	2,035	3.68
Rwanda	27,168	1.33	346	0.63
Sierra Leone	32,232	1.58	543	0.98
Senegal	15,688	0.77	41	0.07
Chad	17,719	0.87	25	0.05
Togo	9,480	0.47	96	0.17
Tajikistan	9,656	0.47	552	1.00
Timor Leste	25,744	1.26	393	0.71
Turkev	9,746	0.48	0	0.00
Tanzania	23,405	1.15	91	0.16
Uganda	27.180	1.33	793	1.43
Yemen	25.434	1.25	0	0.00
South Africa	8.514	0.42	116	0.21
Zambia	30 094	1 4 8	797	1 44
Zimbabwe	19.126	0.94	580	1.05
Total	2,037,523	0.7 1	55,318	1.00
	_,,00		,	

 TABLE 1
 Total sample of women, sample of women from ideal environments

Figure 4 corroborates this implication by plotting the distributions of predicted unmet need under the various methodologies.

When disaggregating the comparative analysis of unmet need at the DHS survey (country-year) level (Table 4), we observe a lot of variation across the surveys; in some cases, we see that our approach estimates a significantly higher (up to 30 percentage points higher)

	Mean	SD	Ν
Full sample			
Outcome and selection characteristics			
Contraceptive use $(1 = Yes)$	0.337		681,542
Highest wealth quintile $(1 = Yes)$	0.199		406,073
Currently married $(1 = Yes)$	0.686		1,396,560
Sexually active $(1 = Yes)$	0.664		752,338
Tertiary education $(1 = Yes)$	0.106		212,751
Knows $1+$ FP method $(1 = Yes)$	0.958		1,884,897
Distance to facility not problem $(1 = Yes)$	0.652		1,206,110
Other characteristics			
Place of residence $(1 = \text{Urban})$	0.376		766,643
Age (years)	29.608	9.830	
Children ever born	2.329	2.336	
Husband has tertiary education $(1 = Yes)$	0.114		115,262
Respondent earns more than husband $(1 = \text{Yes})$	0.266		101,328
Observations	2,037,523		
Ideal environment sample			
Contraceptive use $(1 = Yes)$	0.521	32,786	
Place of residence $(1 = Urban)$	0.807		44,655
Age (years)	33.168	7.495	
Children ever born	1.705	1.278	
Husband has tertiary education $(1 = Yes)$	0.747		24,200
Respondent earns more than husband $(1 = Yes)$	0.371		7,141
Observations	55,318		

TABLE 2 Descriptive statistics, woman characteristics

NOTE: The unit of observation is the woman.



FIGURE 2 Estimated *iCP_{ct}* by country

NOTE: For countries with multiple survey rounds, country-level *iCP* are calculated by averaging survey-round *iCP* estimates.

unmet need than what is currently estimated with the DHS methodology, while in other cases, our approach yields significantly lower estimates (up to 20 percentage points lower) of unmet need compared to the DHS. Figure 5 plots the differences between the counterfactual unmet need measure and the currently used DHS measures. In the survey-based calculations of unmet need, we also note that several calculations of the measure were conducted using small samples of women from identified ideal environments (fewer than 100 women)--the

667



FIGURE 3 Changes in *iCP_{ct}* over time, by country

TABLE 3 Descriptive statistics, unmet need

	Mean	SD	Min	Max
	Mean	50	101111	Mux
Unmet Need, New Definition	0.216	0.088	0.021	0.510
Unmet Need, Definition 1	0.161	0.060	0.053	0.279
Unmet Need, Definition 2	0.152	0.059	0.010	0.271
Difference 1 (New - Def. 1)	0.051	0.102	-0.193	0.338
Difference 2 (New – Def. 2)	0.060	0.101	-0.184	0.348
Ν	80			

NOTE: The unit of observation is the DHS survey round. The variable "Unmet Need, New Definition" is defined as the difference in CPR between the subsample of WRA who are from "ideal" environments (highest wealth quintile, highest educational attainment, currently married, and knows of at least one FP method) and CPR for all WRA. The variable Unmet Need, Definition 1 is calculated using the categorical unmet need variable (v624) in the DHS survey round, which classifies women to fall into one of the following categories: (1) no unmet need; (2) an unmet need for spacing; (3) an unmet need for limiting; (4) having a spacing failure or limiting failure; or (5) infecund. The variable Unmet Need, Definition 2 is calculated using a second categorical unmet need variable (v626) in the DHS survey round. The variable Difference 1 is the calculated difference between the new Unmet Need variable and the Unmet Need, Definition 1 variable. The variable Difference 2 is the calculated difference between the new Unmet Need variable and the Unmet Need, Definition 2 variable.

lack of sample in some surveys poses an empirical concern over the extent to which we have enough statistical precision to estimate ideal contraceptive prevalence.

To quantify the relative contribution of each selecting factor to ideal contraceptive use, we calculate the proportion of the variation in ideal contraceptive use that could be attributed to each of the five key selecting factors that were used for defining the ideal family planning environment. We estimate a logistic model of contraceptive use on these factors together and take the product of the estimated factor coefficients, each of which captured the association between that particular factor and contraceptive use, and the proportion of the pooled DHS sample who did not exhibit that factor. Table 5 presents the results from this analysis. We find that selecting on socioeconomic status (proxied by wealth) and current sexual activity explain the most variation in ideal contraceptive use in women who live in ideal environments relative to the general sample of women. For example, up to 40 percent of the variation in ideal contraceptive use can be attributed to higher socioeconomic status. Interestingly, we find that



FIGURE 4 Kernel density plots, unmet need across definitions





DH3 survey Country Left Left <thleft< th=""> Left Left</thleft<>	LFa Lfa Lfa 9.88% 29.13% 9.88% 33.32% 3.20% 37.46% 10.15% 15.69% 13.81%	Unmet Need, New Def.	(Def. 1)	Difference (New _ Dof 1)	(Def. 2)	Difference (New
AF7Afghanistan2015 36.92% AL7Albania2018 38.20% AM7Armenia2016 60.84% AM7Armenia2016 61.24% BU6Burundi2010 63.77% BU7Burundi2012 33.64% BU7Burundi2011 43.33% CG6Congo, Republic2012 61.90% CM6Conecon2011 43.09% CM6Conecon2011 43.09% CM6Conecon2011 43.09% CM6Conecon2011 43.09% CM6Conecon2013 54.12% CM7Dominican Republic2013 54.12% CM6Conecon2011 54.12% GM6Gabon2013 54.12% GM6Gubana2016 54.12% GM6Gubana2013 31.97% GM6 <th>6.92% 19.52% 9.88% 29.13% 33.32% 0.84% 37.46% 33.32% 0.84% 37.46% 10.15% 10.15% 15.15% 15.15% 15.15% 15.14% 13.81%</th> <th></th> <th>12</th> <th>- 1761.1)</th> <th></th> <th>– Del. 2)</th>	6.92% 19.52% 9.88% 29.13% 33.32% 0.84% 37.46% 33.32% 0.84% 37.46% 10.15% 10.15% 15.15% 15.15% 15.15% 15.14% 13.81%		12	- 1761.1)		– Del. 2)
AI7Albania2018 4988% AM6Armenia2016 60.84% AM7Armenia2016 60.84% AM7Armenia2016 60.84% AM7Armenia2016 60.84% BF6Burkina Faso2016 60.84% BJ7Burkina Faso2010 63.77% BJ6Burundi2010 63.77% BU6Burundi2010 63.77% BU7Burundi2010 63.77% BU7Burundi2011 43.33% CM6Congo, Republic2011 43.03% CM6Comercon2011 43.09% CM6Conercon2011 43.09% CM6Conercon2011 43.09% CM6Conercon2013 71.52% CM6Comercon2011 43.09% CM7DN6Egytt 2013 54.12% CM6Conercon2013 71.52% GM6Gabon2013 71.52% GM6Guna2013 21.23% GM6Guna 2013 31.97% HT7Honduras 2013 31.97% HT6Haiti 2013 31.07% HT7Haiti 2013 31.07% HT6Honduras 2012 54.12% GM6Guna 2013 31.07% HT6Honduras 2013 31.07% HT7Haiti 2013 31.07% HT6Honduras 2012 56.5	9.88% 29.13% 29.13% 28.20% 33.32% 37.46% 35.746% 10.184% 10.15% 15.15% 15.15% 15.15% 15.12% 15.12% 13.81%	17.39%	25.38%	-7.99%	24.82%	-7.43%
AM6Armenia 2010 58.20% AM7Armenia 2016 6.084% AM7Armenia 2016 6.084% BF6Burkina Faso 2010 63.77% B16Benin 2016 6.084% B17Burundi 2010 63.77% B16Burundi 2010 63.47% B17Burundi 2010 63.47% B17Burundi 2010 63.47% B17Burundi 2010 53.52% B17Burundi 2010 53.52% B17Burundi 2011 43.39% CG6Cote d'Ivoire 2011 53.56% CM6Cameroon 2011 23.36% CM6Cameroon 2011 53.36% CM6Cameroon 2013 55.36% CM6Cameroon 2013 55.36% CM7Dominican Republic 2013 55.36% CM6Edoon 2013 54.12% GM6Edoon 2013 54.12% GM6Gabon 2013 54.12% GM6Guma 2013 31.97% GM6Huduras 2013	8.20% 33.32% 0.84% 37.46% 0.84% 10.15% 13.77% 15.69% 18.24% 13.81%	20.76%	12.72%	8.03%	12.39%	8.36%
AM7 Armenia 2016 60.84% AO7 BF6 Burkina Faso 2016 60.84% B76 Burkina Faso 2010 63.77% B7 Burundi 2010 63.77% B7 Burundi 2010 63.77% B7 Burundi 2010 63.77% B07 Burundi 2010 63.77% B07 Burundi 2010 63.77% B07 Burundi 2011 35.53% B07 DRC Congo, Republic 2011 53.36% CI6 Congo, Republic 2011 43.09% CM6 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM6 Cameroon 2011 43.09% CM6 Cameroon 2011 43.09% CM6 Cameroon 2011 54.12% CM7 Daminican Republic 2013 71.52% ET6 Ethiopia 2013 71.52% GA6 Gabon 2013 71.53%	0.84% 37.46% 2.47% 10.15% i3.77% 15.69% i8.24% 13.81%	24.88%	13.64%	11.24%	13.64%	11.24%
AO7 Angola 2016 42.47% BF6 Burkina Faso 2010 63.77% BJ6 Burnin 2012 38.24% BJ7 Burnin 2013 35.52% BU6 Burnini 2013 35.52% BU7 Burnini 2013 35.53% BU7 Burnini 2011 43.33% CG6 Congo, Republic 2011 43.33% CG6 Congo, Republic 2012 36.56% CM6 Cameroon 2011 43.09% CM6 Cameroon 2011 43.09% CM7 Cameroon 2011 55.38% CM6 Cameroon 2013 55.38% CM7 Cameroon 2013 55.38% CM6 Cameroon 2014 55.38% CM7 Cameroon 2013 71.52% ET6 Ethiopia 2014 55.38% CM6 Gabon 2013 21.37% G	2.47% 10.15% 53.77% 15.69% 8.24% 13.81%	23.38%	8.24%	15.15%	7.75%	15.63%
BF6 Burkina Faso 2010 63.77% BJ7 Benin 2012 38.24% BJ7 Benin 2013 35.52% BU6 Burundi 2010 35.52% BU7 Burundi 2010 35.53% BU7 Burundi 2010 35.54% BU7 Burundi 2011 43.33% CG6 Congo, Republic 2012 36.36% CI6 Cote d'Ivoire 2011 43.09% CM6 Cameroon 2013 71.52% CM7 Cameroon 2014 53.36% CM7 Cameroon 2013 71.52% ET6 Ethiopia 2014 53.31% ET7 Ethiopia 2012 54.12% GA6 Gabon 2013 71.52% GA6 Gabon 2013 71.52% GM6 Gabon 2013 31.97% HT6 Haiti 2013 21.2% GM6	3.77% 15.69% 8.24% 13.81%	32.32%	25.54%	6.78%	24.57%	7.75%
BJ6 Benin 2012 38.24% BU6 Burundi 2013 35.52% BU7 Burundi 2017 43.33% CD6 Burundi 2017 43.33% CJ6 Burundi 2017 43.33% CJ6 DRC 2014 52.36% CJ6 Congo, Republic 2011 43.09% CM6 Comercon 2011 43.09% CM7 Camercon 2011 43.09% CM7 Camercon 2013 71.52% ET6 Ethiopia 2013 71.52% ET7 Ethiopia 2014 54.12% GA6 Gabon 2014 54.12% GA6 Gabon 2013 71.52% GA6 Gabon 2014 55.12% GA6 Gabon 2013 71.52% GA6 Gabon 2013 71.52% GA6 Gabon 2013 31.97% GA6 Gabon <td>8.24% 13.81%</td> <td>48.08%</td> <td>19.72%</td> <td>28.36%</td> <td>19.09%</td> <td>28.99%</td>	8.24% 13.81%	48.08%	19.72%	28.36%	19.09%	28.99%
BJ7 Benin 2018 35.52% BU7 Burundi 2010 35.64% BU7 Burundi 2017 43.33% CD6 DRC DRC 2014 52.36% CJ6 Congo, Republic 2011 43.33% CM6 Cone d'Ivoire 2012 61.90% CM6 Coneroon 2011 43.09% CM7 Coneroon 2013 55.36% CM7 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM7 Cameroon 2013 71.52% E166 Ethiopia 2013 71.52% GA6 Gabon 2014 60.6% GA6 Gabon 2013 31.97% GT Guraa 2013 31.97% GT Guraa 2013 31.97% GT Gabon 2013 31.97% GT Guraa 2013 31.97% GT <td></td> <td>24.43%</td> <td>24.87%</td> <td>-0.45%</td> <td>24.31%</td> <td>0.12%</td>		24.43%	24.87%	-0.45%	24.31%	0.12%
BU6 Burundi 2010 33.64% BU7 Burundi 2017 33.64% BU7 Burundi 2017 43.33% CD6 DRC 2014 52.36% CI66 Congo, Republic 2011 51.30% CM6 Congo, Republic 2012 61.90% CM7 Comercon 2011 43.09% CM6 Camercon 2013 55.38% CM7 Camercon 2013 55.38% CM7 Camercon 2014 60.66% ET6 Ethiopia 2014 60.66% ET7 Ethiopia 2014 56.34% GA6 Gabon 2013 31.75% GA6 Gabon 2013 31.79% GM6 Gabon 2013 31.79% GM6 Gabon 2013 31.79% GA6 Gabon 2013 31.79% GM6 Gabon 2013 31.79% GM6 <	5.52% 14.55%	20.97%	25.52%	-4.55%	24.88%	-3.91%
BU7 Burundi 2017 43.33% CD6 DRC 2014 52.36% CG6 Congo, Republic 2012 61.90% CM6 Cone of Yorice 2012 61.90% CM7 Cameroon 2011 43.03% CM6 Cameroon 2013 55.38% CM7 Cameroon 2011 43.09% CM7 Cameroon 2013 55.38% DR6 Cameroon 2014 60.66% ET6 Ethiopia 2.014 60.66% ET6 Ethiopia 2.011 58.31% GA6 Gabon 2.011 58.31% GA6 Gabon 2.011 58.31% GM6 Gabon 2.013 31.97% HT7 Hinea 2.013 31.97% HT6 Honduras 2.013 31.97% HT6 Gabon 2.013 31.97% HT6 Gabon 2.013 31.97% HT6	3.64% 17.46%	29.33%	17.94%	11.39%	17.52%	11.81%
CD6 DRC 2014 52.36% CG6 Congo, Republic 2012 61.90% CI6 Cone d'Ivoire 2012 61.90% CM6 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM6 Cameroon 2011 55.33% DR6 Ethiopia 2013 71.52% ET7 Ethiopia 2014 56.33% GA6 Gabon 2012 58.31% GA6 Gabon 2013 31.97% GN6 Gambia 2013 22.83% GN6 Gabina 2013 22.83% GN6 Gabina 2013 22.83% GN6 Gabina 2013 31.97% GN6 Gabina 2013 31.97% GN6 Gabina 2013 22.83% GN6 <	3.33% 14.01%	16.19%	16.50%	-0.32%	15.85%	0.34%
CG6 Congo, Republic 2012 61.90% CI6 Cone d'Ivoire 2012 61.90% CM6 Cameroon 2011 43.09% CM7 Cameroon 2013 55.38% CM6 Cameroon 2013 55.38% CM7 Cameroon 2011 55.38% CM6 Cameroon 2013 71.52% ET6 Ethiopia 2014 55.38% ET7 Ethiopia 2014 58.31% ET7 Ethiopia 2014 58.31% ET7 Ethiopia 2014 55.38% GM6 Gabon 2014 55.73% GM6 Gabon 2013 31.97% GM6 Guinea 2013 31.97% HN6 Haiti 2013 31.97% HN6 Haiti 2012 30.15% HT7 Haiti 2012 30.15% HT6 Haiti 2012 30.15% IA6	52.36% 16.86%	35.50%	21.63%	13.87%	20.64%	14.86%
CI6 Cote d'Ivoire 2012 36.36% CM6 Cameroon 2011 43.09% CM7 Cameroon 2011 43.09% CM7 Cameroon 2013 71.52% EG6 Egypt 2014 60.66% ET7 Ethiopia 2014 60.66% ET7 Ethiopia 2014 54.12% GA6 Gabon 2014 53.35% GA6 Gabon 2014 55.412% GA6 Gabon 2014 35.75% GM6 Gambia 2013 32.53% GN6 Gambia 2013 32.53% GU6 Guna 2013 31.97% HT6 Houluras 2013 31.97% HT7 Haiti 2013 31.97% HT6 Haiti 2013 31.97% HT7 Haiti 2013 31.97% HT6 Huduras 2013 31.97% IA6 IA6	51.90% 42.37%	19.53%	14.30%	5.24%	13.96%	5.58%
CM6 Cameroon 2011 43.09% CM7 Cameroon 2013 55.38% DR6 Dominican Republic 2013 71.52% ET6 Ethiopia 2013 71.52% ET7 Ethiopia 2014 60.66% ET7 Ethiopia 2014 60.66% ET7 Ethiopia 2014 58.31% ET7 Ethiopia 2014 58.31% ET7 Ethiopia 2014 58.31% GA6 Gabon 2014 58.31% GT6 Gabana 2013 21.23% GN7 Guna 2013 31.97% GN7 Guna 2013 31.97% HT6 Haiti 2013 31.07% HT7 Haiti 2013 31.07% HT7 Haiti 2012 38.68% HT7 Haiti 2012 30.15% IA6 Indonesia 2017 30.15% IA6	6.36% 18.79%	17.58%	22.27%	-4.69%	21.57%	-3.99%
CM7 Cameroon 2018 55.38% DR6 Dominican Republic 2013 55.38% ET6 Ethiopia 2014 60.66% ET7 Ethiopia 2011 58.15% ET7 Ethiopia 2011 58.13% ET7 Ethiopia 2011 58.13% ET7 Ethiopia 2011 58.13% ET7 Ethiopia 2011 58.12% GA6 Gabon 2012 59.10% GH6 Gambia 2013 35.75% GN7 Guinea 2013 31.97% HT6 Honduras 2013 31.97% HT7 Hait 2013 31.97% HT6 Honduras 2012 38.68% HT7 Hait 2012 31.97% IA6 Indomesia 2012 31.97% IA6 Indomesia 2012 31.97% IA6 Indomesia 2012 31.97% IA6	3.09% 20.26%	31.72%	18.26%	13.46%	17.35%	14.36%
DR6 Dominican Republic 2013 71.52% EG6 Egypt 60.66% 53.31% ET7 Ethiopia 2014 60.66% ET7 Ethiopia 2015 54.12% GA6 Gabon 2016 54.12% GA6 Gabon 2016 54.12% GA6 Gabon 2016 54.12% GA6 Gabon 2013 53.75% GM6 Gambia 2013 22.83% GN7 Guinea 2013 22.83% GV6 Gambia 2013 31.97% GU6 Guinea 2013 31.97% GU7 HN6 Honduras 2013 31.97% HT7 Haiti 2012 31.07% 41.07% HT7 Haiti 2012 30.15% 41.07% HT7 Haiti 2012 30.15% 41.07% IA6 India 2012 30.15% 41.07% IA6 Indo	5.38% 23.67%	22.83%	17.63%	5.20%	16.72%	6.11%
EG6 Egypt 2014 60.66% ET7 Ethiopia 2011 58.31% ET7 Ethiopia 2016 54.12% GA6 Gabon 2012 50.00% GA6 Gabon 2013 54.12% GM6 Gabna 2013 54.12% GM6 Gabna 2013 35.75% GM6 Gambia 2013 22.83% GN7 Guinea 2013 35.75% GU6 Guinea 2013 35.75% GU6 Guinea 2013 31.97% GU6 Houduras 2013 31.97% HN6 Houduras 2013 31.6% HT7 Haiti 2012 38.6% HT7 Haiti 2012 30.15% IA6 Indonesia 2012 56.54% ID7 Indonesia 2012 56.54% ID7 Jordan 2013 65.8% FG6 Indonesia	71.52% 55.99%	15.53%	8.11%	7.42%	7.82%	7.71%
ET6 Ethiopia 2011 58.31% ET7 Ethiopia 2016 54.12% GA6 Gabon 2012 50.00% GA6 Gabon 2013 57.75% GM6 Gambia 2013 32.83% GN6 Gunea 2013 32.83% GN6 Gunea 2013 32.83% GN6 Gunea 2013 31.97% HN6 Honduras 2015 76.24% HT7 Haiti 2012 81.07% HT7 Haiti 2012 31.97% HT7 Haiti 2012 31.97% HT7 Haiti 2012 31.07% HT7 Haiti 2012 30.15% IA6 India 2012 30.15% ID7 India 2012 50.86% ID7 India 2012 56.54% ID6 Jordan 2012 56.54% ID7 Jordan 2013 65.86% V60 Jordan 2013 65.86%	0.66% 53.63%	7.03%	9.53%	-2.50%	9.51%	-2.47%
ET7 Ethiopia 2016 54.12% GA6 Gabon 2012 50.00% GA6 Gabaa 2013 5.75% GM6 Gambia 2013 2.533% GN7 Guina 2013 2.233% GN6 Gambia 2013 2.233% GN7 Guina 2013 2.233% GN6 Guina 2013 31.97% HN6 Honduras 2015 76.24% HT7 Haiti 2012 38.68% HT7 Haiti 2017 30.15% IA6 India 2017 30.15% ID7 India 2017 30.15% ID7 India 2017 30.15% ID6 Indonesia 2012 56.54% IO7 Jordan 2013 65.86% VEA Vanno 2013 65.86%	8.31% 17.93%	40.38%	15.10%	25.28%	14.42%	25.96%
GA6 Gabon 2012 50.00% GH6 Gambia 2013 35.75% GM6 Gambia 2013 35.75% GN7 Guinea 2013 35.75% GN7 Guinea 2013 32.83% GN7 Guinea 2013 31.97% HT6 Hait 2012 31.07% HT7 Hait 2012 31.68% HT7 Hait 2012 31.68% HT7 Hait 2012 31.68% IA6 India 2017 30.15% ID6 Indonesia 2017 30.15% ID7 Indonesia 2012 56.54% IO7 Jordan 2013 65.86% VEC Vecann 2013 65.86%	4.12% 21.12%	33.00%	11.96%	21.04%	11.58%	21.42%
GH6 Ghana 2014 35.75% GM6 Gambia 2013 22.83% GN7 Guinea 2013 22.83% GU6 Guyana 2015 76.24% HT6 Haiti 2015 76.24% HT6 Haiti 2012 81.07% HT7 Haiti 2012 81.07% HT7 Haiti 2012 38.68% HT7 Haiti 2012 38.68% HT7 India 2012 38.68% ID6 Indonesia 2012 58.66% ID7 Jordan 2012 56.54% JO7 Jordan 2013 65.86% VE6 Vanna 2013 65.86%	0.00% 28.98%	21.02%	22.20%	-1.19%	20.81%	0.20%
GM6 Gambia 2013 22.83% GN7 Guinea 2013 21.97% GU6 Guinea 2015 76.24% HN6 Honduras 2015 76.24% HT6 Haiti 2012 81.07% HT7 Haiti 2012 81.07% HT7 Haiti 2012 31.66% HT7 Haiti 2012 31.68% HT7 Haiti 2017 30.15% IA6 India 2017 30.15% ID6 Indonesia 2012 58.16% ID7 Indonesia 2012 56.54% JO6 Jordan 2013 65.86% VEA Vanno 2013 65.86%	5.75% 22.16%	13.60%	19.79%	-6.20%	18.68%	-5.08%
GN7 Guinea 2018 31.97% GU6 Guyana 2015 76.24% HN6 Honduras 2015 76.24% HT6 Haiti 2012 81.07% HT7 Haiti 2012 81.07% HT7 Haiti 2012 38.68% HT7 Haiti 2017 30.15% IA6 India 2017 30.15% ID6 Indonesia 2012 58.66% ID7 Indonesia 2012 56.54% JO6 Jordan 2012 56.54% JO7 Jordan 2013 65.86% VEAND 2014 2012 56.54%	2.83% 6.68%	16.15%	17.59%	-1.44%	17.05%	-0.90%
GU6 Guyana 2015 76.24% HN6 Honduras 2012 81.07% HT6 Haiti 2012 81.07% HT7 Haiti 2012 36.8% HT7 Haiti 2017 30.15% IA6 India 2017 30.15% ID6 Indonesia 2012 51.98% ID7 India 2012 58.06% ID7 Indonesia 2012 56.54% JO6 Jordan 2013 65.86% VEC Value 2013 65.86%	31.97% 10.83%	21.14%	17.07%	4.07%	16.38%	4.76%
HN6 Honduras 2012 81.07% HT6 Haiti 2012 81.07% HT7 Haiti 2017 30.15% HT7 Haiti 2017 30.15% IA6 India 2016 51.98% ID6 Indonesia 2012 58.06% ID7 Indonesia 2012 56.54% JO6 Jordan 2012 56.54% JO7 Jordan 2013 65.86% VEC Vacano 2013 65.86%	6.24% 38.89%	37.35%	7.93%	29.42%	7.75%	29.60%
HT6 Haiti 2012 38.68% HT7 Haiti 2017 30.15% IA6 India 2016 51.98% ID6 Indonesia 2012 58.06% ID7 Indonesia 2017 56.54% ID7 Indonesia 2012 58.66% JO6 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VF6 Varmo 2014 65.86%	31.07% 48.13%	32.93%	6.81%	26.13%	6.62%	26.32%
HT7 Haiti 2017 30.15% IA6 India 2016 51.98% ID6 Indonesia 2012 58.06% ID7 Indonesia 2012 56.54% ID7 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VF6 Variant 2018 65.86%	8.68% 23.33%	15.35%	24.64%	-9.29%	23.45%	-8.09%
IA6 India 2016 51.98% ID6 Indonesia 2012 58.06% ID7 Indonesia 2017 56.54% JO6 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VFC Value 2014 65.86%	0.15% 23.99%	6.15%	25.42%	-19.26%	24.60%	-18.44%
ID6 Indonesia 2012 58.06% ID7 Indonesia 2017 56.54% JO6 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VF Varian 2018 65.86%	51.98% 37.34%	14.64%	10.29%	4.35%	9.66%	4.98%
ID7 Indonesia 2017 56.54% JO6 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VF Ventor 2018 65.86%	8.06% 42.42%	13.38%	6.49%	6.89%	6.44%	6.94%
JO6 Jordan 2012 49.57% JO7 Jordan 2018 65.86% VFG Varian 2018 65.86%	6.54% 43.16%	15.64%	6.10%	9.54%	5.91%	9.73%
JO7 Jordan 2018 65.86% VE6 Venue 2014 64.08%	9.57% 47.44%	7.60%	9.05%	-1.45%	8.71%	-1.12%
VEG Vanito 7014 64.0806	5.86% 58.26%	2.14%	13.82%	-11.69%	13.31%	-11.17%
NEO 2014 01:00 0	4.08% 38.71%	25.37%	12.31%	13.06%	11.81%	13.56%
KH5 Cambodia 2010 58.19%	8.19% 37.37%	33.34%	13.39%	19.95%	12.90%	20.44%
KH6 Cambodia 2014 60.68%	0.68% 27.34%	20.82%	8.51%	12.31%	0.97%	19.84%
KM6 Comoros 2012 30.43%	0.43% 14.49%	15.95%	21.84%	-5.89%	20.12%	-4.17%
KY6 Kenya 2012 41.54%	1.54% 26.56%	14.98%	11.73%	3.24%	11.18%	3.80%

670

Diffsurvey Cuntry Year CP_a Unmet Need, New Def. Unmet Need, (Def.1) Diff. Diff. cound Lase Liberia 2013 32.41% 20.26% 19.43% Diff. Diff. Diff. Lase Lise Lescolio 2013 35.24% 10.59% 27.64% B8.4% Diff.	Imet Need Difference (New (Def. 1) - Def. 1) 27,83% -15,68% 10.94% 8.49% 18.84% -130% 22.29% -1.30%	Unmet Need Differ	vence (New
Liberia 2013 32.41% 20.26% 1.15% 27.83% 27.83% Liso Mili 2013 32.41% 20.26% 12.15% 27.83% 27.84% 10.94% 27.84% 10.94% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.94% 27.84% 10.34% 13.85% 27.84% 13.86% 27.84% 13.24%	27,83% -15,68% 10.94% 8.49% 8.49% 2.229% -1.30% -1.30%	(Det. 2) –	- Def. 2)
Liso Lesotho 2014 67.98% 48.56% 19.43% 10.94% MLG Mali 2013 38.24% 10.59% 7.61% 26.81% 2.2.29% ML7 Mali 2013 38.51% 14.51% 7.61% 26.81% 2.8.84% MV7 Maldrices 2017 2.2.29% 14.51% 2.2.29% 2.2.29% MV7 Maldrices 2010 55.57% 3.5.41% 10.34% 2.2.29% MV7 Maldrice 2011 4.6.89% 15.42% 3.1.47% 13.77% MV6 Nigeria 2011 4.6.89% 15.42% 3.1.47% 13.77% MV6 Nigeria 2013 34.6.4% 13.4.9% 13.2.4% 13.2.7% N16 Nigeria 2013 48.1.5% 13.6.5% 3.4.56% 13.4.6% 15.1.3% N16 Nigeria 2011 61.6.6% 3.3.5.3% 2.1.6% 15.1.2% N16 Nigeria 2011 61.6.9% 3.3.5.	10.94% 8.49% 8.49% 18.84%	26.36%	-14.21%
ML6 Mali 2013 38.24% 10.59% 27.64% 18.84% ML7 Mality 2013 35.51% 14.53% 20.99% 22.29% MV7 Maldives 2011 25.52% 45.57% 23.11% 18.85% MV7 Maldive 2010 55.92% 45.57% 23.11% 18.85% MV7 Maldive 2011 45.95% 55.41% 10.34% 18.35% MV7 Malawi 2013 24.65% 15.98% 24.68% 13.34% NG6 Nigeria 2013 44.65% 15.98% 54.9% 13.34% NG6 Nigeria 2013 44.65% 15.98% 51.3% 15.35% NG6 Nigeria 2013 54.6% 13.34% 15.35% 15.35% NM6 Nepal 2013 56.39% 34.15% 34.50% 87.79% 15.75% NM6 Nepal 2013 56.34.3% 34.30% 16.75% 16.75%	18.84% 8.80% 22.29% -1.30%	10.59%	8.84%
ML7 Mali 2018 3551% 14,53% 2009% 22.29% MV7 Malawi 2017 52.22% 14,53% 7.61% 2.631% MW7 Malawi 2016 58.52% 35.41% 10.34% 13.38% MW7 Malawi 2016 58.52% 35.41% 10.34% 13.38% MW7 Malawi 2016 58.52% 35.41% 10.34% 13.38% MW7 Nigeria 2013 44.65% 15.42% 31.47% 13.38% NG6 Nigeria 2013 44.65% 13.49% 13.47% 13.38% NG6 Nigeria 2013 44.65% 13.49% 13.47% 13.57% NG6 Nambla 2011 61.36% 34.53% 14.59% 15.53% NG6 Nambla 2013 59.45% 34.53% 11.56% 15.53% NG6 Nambla 2011 61.36% 34.53% 12.21% 15.53% NF6 Ph	22.29% —1.30%	18.35%	9.29%
MV7 Maldives 2017 22.22% 14.61% 7.61% 26.81% 2.88% MW5 MMV5 Madawi 2010 55.92% 45.57% 2.111% 18.85% MY7 Madawi 2010 55.92% 45.57% 2.111% 18.85% MY6 Nigeria 2011 46.89% 15.42% 31.47% 13.34% NG6 Nigeria 2013 44.65% 15.42% 31.47% 13.34% NG7 Nigeria 2013 44.65% 15.42% 31.47% 18.59% 13.54% NG6 Nigeria 2013 44.65% 15.42% 31.47% 18.59% 11.53% NM6 Nigeria 2013 48.15% 13.65% 34.50% 11.55% 11.55% NM6 Nepal 2013 68.91% 30.41% 13.56% 11.55% 11.55% NM6 Nepal 2013 68.91% 30.41% 13.56% 11.65% 15.36% 11.67% 15.56% 15.76	10 200	18.09%	2.90%
MW5Malawi201055.92% 45.57% 23.11%18.58%MW7Malawi201655.92% 45.57% 23.11%18.85%MZ6Mozambique201146.9%15.47%13.47%18.77%NG7Nigeria201148.15%15.98%28.66%15.13%18.37%NG6Nigeria201148.15%13.49%21.11%15.38%NG6Nigeria201248.15%13.49%21.11%15.38%NF6Nigeria201349.15%50.44%13.56%34.50%14.69%NF6Nambia201161.36%31.65%34.50%18.50%15.76%NF6Nambia201369.94%33.63%11.57%15.56%11.57%NF6Philippines201349.91%34.30%33.63%10.23%20.68%PH7Philippines201359.45%33.63%11.57%11.57%11.57%PH6Philippines201345.06%31.54%12.33%20.08%PH7Philippines201359.45%33.63%20.28%20.08%PH6Philippines201359.75%33.53%20.28%20.28%PH6Philippines201359.45%33.63%20.28%20.28%PH6Philippines201359.75%33.53%10.57%20.98%PK7Pakistan201359.75%33.63%20.38%21.26%PK6Raunda2013 <t< td=""><td>-19.20%</td><td>24.25%</td><td>-16.64%</td></t<>	-19.20%	24.25%	-16.64%
MW7 Malawi 2016 58.52% 35.41% 10.34% 13.38% MZ6 Nigeria 2011 46.89% 15.42% 31.47% 15.13% NG6 Nigeria 2011 46.89% 15.42% 31.47% 15.13% NG6 Nigeria 2011 46.89% 15.42% 31.47% 15.13% NG6 Nigeria 2011 46.89% 50.41% 15.45% 15.13% NM6 Namibia 2011 66.91% 50.41% 13.49% 21.15% 15.13% NP7 Nepal 2011 61.36% 50.41% 13.54% 11.67% 15.55% NP6 Nepal 2011 61.36% 34.15% 10.57% 11.57% NP6 Nepal 2013 59.45% 38.24% 20.47% 11.57% NP6 Philippines 2013 59.45% 34.15% 12.34% 20.68% PK7 Pakistan 2013 57.16% 11.57% 12.33% 12.57%<	18.85% 4.26%	18.30%	4.80%
MZ6 Mozambique 2011 46.89% 15.42% 31.47% 18.77% NG6 Nigeria 2013 44.65% 15.98% 28.66% 13.24% 13.13% NG7 Nigeria 2013 44.65% 15.98% 28.66% 13.24% 13.55% 31.51% 13.13% NM6 Namibia 2013 44.65% 13.65% 34.50% 8.71% 18.53% 14.69% 15.76% 11.56% 14.69% 8.71% NM6 Namibia 2013 68.91% 50.41% 18.50% 8.71% 18.52% 14.69% 8.71% 18.52% 11.57% 15.76% 11.57% 18.52% 11.57% 18.52% 21.21% 18.52% 11.57% 18.52% 21.66% 12.33% 11.57% 18.29% 20.68% 21.66% 12.33% 11.57% 18.52% 11.57% 18.29% 21.66% 12.33% 11.57% 18.29% 18.29% 20.68% 21.66% 21.26% 21.66% 21.26% 21.66% 21.26%	13.38% —3.04%	- 12.76%	-2.42%
NG6 Nigeria 2013 44.65% 15.98% 28.66% 13.24% NG7 Nigeria 2013 44.65% 15.98% 28.66% 13.24% NG7 Nigeria 2012 48.15% 13.49% 21.13% 15.13% NH6 Namibia 2012 48.15% 50.41% 18.50% 19.58% NP7 Nh6 Namibia 2011 61.36% 40.89% 21.11% 15.76% 11.67% NP7 Nh7 Nepal 2011 61.36% 33.65% 20.47% 18.52% PH6 Philippines 2013 59.45% 38.24% 21.21% 18.58% PK7 Pakistan 2013 59.14% 34.30% 19.57% 11.67% PK7 Pakistan 2012 53.83% 34.30% 19.53% 20.68% 11.67% PK7 Pakistan 2013 37.16% 33.54% 12.33% 10.22% RW6 Sterra Leone 2013 37.16% 3	18.77% 12.70%	17.93%	13.54%
NG7 Nigeria 2018 34.64% 13.49% 21.15% 15.13% NI6 Niger 2012 48.15% 13.65% 34.20% 15.13% NM6 Namibia 2013 6.8.19% 50.41% 13.65% 34.10% 146.9% NP6 Nepal 2011 61.36% 40.89% 21.21% 19.58% NP7 Nepal 2011 61.36% 34.15% 11.56% 8.21% NP6 Nepal 2013 69.91% 34.15% 11.58% 11.58% PH7 Philippines 2017 45.96% 33.24% 20.47% 18.52% PK7 Pakistan 2017 45.96% 31.54% 14.52% 11.57% PK7 Pakistan 2013 57.16% 31.54% 10.23% 10.57% S16 Sitra Leone 2013 57.16% 23.38% 20.48% 20.48% S16 Sitra Leone 2011 36.59% 36.3% 27.95% 21.26%	13.24% 15.43%	12.59%	16.07%
NIG Niger 2012 48.15% 13.65% 34.50% 14.69% NM6 Namibia 2013 68.91% 50.41% 18.50% 8.71% 18.50% 8.71% NM6 Nepal 2013 68.91% 50.41% 18.50% 8.71% 18.50% 8.71% NP7 Nepal 2011 61.36% 49.91% 31.54% 15.76% 11.58% PH7 Philippines 2013 49.91% 33.63% 21.21% 19.58% PK7 Pakistan 2013 49.91% 33.63% 12.33% 11.67% PK6 Pakistan 2012 53.83% 34.56% 33.63% 12.33% 11.67% PK6 Pakistan 2012 45.96% 33.63% 12.33% 10.22% PK6 Rwdd Rwda 2013 37.16% 23.54% 23.68% 20.03% S16 Sierra Leone 2013 37.16% 23.53% 21.30% 19.29% S16 Sierra	15.13% 6.02%	14.44%	6.71%
NM6 Namibia 2013 68.91% 50.41% 18.50% 8.71% NP7 Nepal 2011 61.36% 40.89% 21.121% 19.58% NP7 Nepal 2016 59.45% 38.24% 20.47% 18.52% PH6 Philippines 2013 49.91% 34.15% 15.76% 11.67% PH7 Philippines 2012 45.96% 33.63% 20.47% 18.52% PK7 Pakistan 2012 45.96% 31.54% 14.52% 11.67% PK7 Pakistan 2013 37.16% 23.63% 20.68% 10.22% RW6 Rwanda 2013 37.16% 23.64% 20.68% 10.22% SL6 Sierra Leone 2013 50.71% 23.53% 10.25% 10.25% SL6 Sierra Leone 2013 50.71% 23.53% 21.36% 29.53% SL6 Sierra Leone 2013 50.71% 22.33% 21.36% 21.36%	14.69% 19.81%	14.31%	20.19%
NP6 Nepal 2011 61.36% 40.89% 21.21% 19.58% PH6 Philippines 2013 49.91% 33.24% 20.47% 18.52% PH7 Philippines 2013 49.91% 34.15% 15.76% 11.58% PH7 Philippines 2013 49.91% 34.15% 15.76% 11.58% PK6 Pakistan 2013 45.96% 33.63% 12.33% 11.67% PK7 Pakistan 2013 53.63% 34.30% 12.23% 11.57% PK7 Pakistan 2013 37.16% 31.54% 14.52% 18.29% S16 Sierra Leone 2013 37.16% 23.84% 20.03% 10.22% S16 Sierra Leone 2013 37.16% 23.84% 20.03% 10.22% S16 Sierra Leone 2013 37.16% 23.84% 20.03% 10.22% S16 Sierra Leone 2013 37.16% 23.53% 27.9% 18.36%	8.71% 9.79%	8.41%	10.09%
NP7 Nepal 2016 59.45% 38.24% 20.47% 18.52% PH6 Philippines 2013 49.91% 34.15% 15.76% 11.58% PH7 Philippines 2017 45.96% 33.63% 15.76% 11.58% PK6 Pakistan 2017 45.96% 33.63% 12.33% 11.67% PK7 Pakistan 2010 60.12% 23.63% 31.54% 14.52% 16.7% PK7 Pakistan 2010 60.12% 29.62% 30.50% 10.22% S16 Sterra Leone 2010 60.12% 29.62% 33.50% 10.22% S16 Sterra Leone 2010 60.12% 29.62% 33.50% 10.22% S16 Sterra Leone 2011 35.59% 8.63% 21.32% 20.03% S16 Chad 205 20.03% 33.34% 21.32% 21.05% S16 Chad 20.13 35.59% 20.53% 21.36% 20.03%	19.58% 1.63%	19.38%	1.83%
PH6Philippines2013 49.91% 34.15% 15.76% 11.58% PH7Philippines2017 45.96% 33.63% 12.33% 11.67% PK6Pakistan2012 53.83% 34.30% 19.53% 11.67% PK7Pakistan2012 53.83% 34.30% 19.53% 11.67% RW6Rwanda2010 60.12% 23.63% 30.26% 10.22% SL6Sitera Leone2013 37.16% 23.84% 20.03% 10.22% SL7Sitera Leone2013 37.16% 23.84% 20.03% 10.22% SN6Sitera Leone2013 37.16% 23.84% 20.03% 10.22% SN6Sitera Leone2013 37.16% 23.84% 20.03% 21.26% TD6Chad2013 37.6% 3.53% 27.95% 21.26% TG6Togo2013 35.5% 19.27% 22.39% 25.08% TG6Togo2013 35.87% 19.27% 22.39% 25.08% TG6Togo 2013 35.87% 19.27% 22.39% 25.08% TG6Timor Leste 2010 35.87% 19.27% 22.39% 17.21% TL5Timor Leste 2010 54.55% 29.07% 20.59% 17.21% TG6Uganda 2010 54.55% 29.07% 20.59% 17.21% TC7Uganda 2016 56.00% 25.23% 20.59% 17.21% <	18.52% 1.95%	18.24%	2.23%
PH7Philippines 2017 45.96% 33.63% 12.33% 11.67% PK6Pakistan 2012 53.83% 34.30% 19.53% 20.68% PK7Pakistan 2012 53.83% 34.30% 19.53% 20.68% PK6Rwanda 2012 53.83% 34.30% 19.53% 20.68% RV6Rwanda 2010 60.12% 23.62% 30.50% 10.22% SL6Sierra Leone 2013 37.16% 23.84% 28.18% 20.03% SL7Sierra Leone 2013 37.16% 22.53% 13.32% 10.22% SN6Senegal 2011 36.59% 8.63% 27.95% 21.26% SN6Senegal 2011 36.59% 8.63% 27.95% 21.26% TG6T0go 2014 41.67% 19.27% 22.39% 25.08% TG6T0go 2014 41.67% 19.27% 22.39% 25.39% TG6T0go 2012 40.58% 19.27% 22.39% 25.39% TG6Timor Leste 2010 35.87% 19.27% 22.39% 17.21% TL5Timor Leste 2010 54.55% 29.07% 50.98% 17.21% TC7Tanzania 2016 56.00% 25.23% 20.59% 17.21% UG6Uganda 2016 56.00% 25.23% 20.59% 17.21% UG7Uganda 2016 56.00% 22.28% 20.59% 20.5	11.58% 4.17%	10.61%	5.15%
PK6Pakistan 2012 53.33% 34.30% 19.53% 20.68% PK7Pakistan 2012 53.33% 34.30% 19.53% 20.68% PK7Pakistan 2018 46.06% 31.54% 14.52% 18.29% SL6Sierra Leone 2013 37.16% 29.62% 30.50% 10.22% SL6Sierra Leone 2013 37.16% 23.84% 2818% 2003% SL6Sierra Leone 2013 37.16% 23.34% 2818% 20.03% SN6Senegal 2011 36.59% 8.63% 27.95% 21.26% SN6Chad 2011 36.59% 8.63% 27.95% 21.26% TG6Togo 2014 41.67% 19.27% 22.39% 21.32% TG6Togo 2014 41.67% 19.27% 22.39% 25.08% TJ5Timor Leste 2010 35.87% 19.27% 22.39% 25.96% TL5Timor Leste 2010 35.87% 19.27% 22.39% 17.21% TL5Timor Leste 2010 54.55% 29.07% 50.98% 17.21% UG6Uganda 2016 54.55% 29.28% 20.59% 20.59% UG7Uganda 2016 51.04% 29.28% 20.59% 20.59% UG7Uganda 2016 57.30% 22.28% 20.59% 20.59% UG7Uganda 2016 51.04% 20.28% 20.59% <t< td=""><td>11.67% 0.66%</td><td>10.41%</td><td>1.92%</td></t<>	11.67% 0.66%	10.41%	1.92%
PK7Pakistan2018 46.06% 31.54% 14.52% 18.29% RW6Rwanda2010 60.12% 29.62% 30.50% 10.22% SIFra Leone2013 37.16% 23.84% 28.18% 20.03% SL6Sierra Leone2013 37.16% 23.84% 28.18% 20.03% SN6Senegal2011 36.59% 8.63% 23.32% 19.79% SN6Chad2011 36.59% 8.63% 27.32% 21.26% TD6Chad2014 41.67% 19.27% 22.39% 21.26% TG6Togo2014 41.67% 19.27% 22.39% 25.03% TL5Timor Leste2010 35.87% 19.27% 22.37% 18.52% TL7Timor Leste2010 55.87% 19.27% 22.37% 18.52% TC6UG6 20.00% 3.99% 15.01% 15.21% 20.4% TC7Tanzania 2010 54.55% 29.07% 50.29% 17.21% UG6Uganda 2016 51.04% 22.28% 20.59% 20.59% UG7Uganda 2016 57.30% 22.28% 20.59% 20.59% UG7Uganda	20.68% -1.16%	- 20.61%	-1.08%
RW6Rwanda2010 60.12% 29.62% 30.50% 10.22% S16Sierra Leone2013 37.16% 23.84% 20.03% 19.79% S17Sierra Leone2013 37.16% 23.84% 20.03% 19.79% S16Sierra Leone2013 37.16% 23.84% 20.03% 29.03% S16Sierra Leone2011 36.59% 8.63% 27.95% 21.05% S16Chad2011 36.59% 8.63% 27.95% 21.26% TD6Chad2011 36.59% 8.63% 27.95% 21.26% TD6Togo2014 41.67% 19.27% 22.39% 21.36% TL5Timor Leste2010 35.87% 19.18% 22.39% 25.98% TL7Timor Leste2010 54.58% 15.22% 20.4% TZ7Tanzania2016 54.58% 15.22% 20.4% UG6Uganda2016 51.04% 22.28% 20.59% UG7Uganda2016 51.04% 22.28% 20.59% UG7Uganda2016 55.30% 22.02% 20.59% UG7Uganda2016 55.30% 22.05% 20.59% UG7Uganda2016 55.30% 22.05% 20.59% UG7Uganda2016 55.30% 20.70% 20.59% UG7Uganda2016 55.30% 20.70% 20.59%	18.29% — 3.77%		-3.54%
SI6 Sierra Leone 2013 37.16% 23.84% 28.18% 20.03% SI7 Sierra Leone 2013 37.16% 23.34% 28.18% 20.03% SI7 Sierra Leone 2019 50.71% 22.53% 13.32% 19.79% SN6 Senegal 2011 36.59% 8.63% 27.95% 21.26% SN6 Senegal 2014 41.67% 19.27% 22.39% 21.26% TG6 Tapikistan 2014 41.67% 19.27% 22.39% 25.08% TI6 Tapikistan 2012 40.58% 19.18% 21.40% 15.91% TL7 Timor Lest 2010 35.87% 13.50% 22.39% 25.30% TL7 Timor Lest 2010 35.87% 13.50% 15.21% 15.22% TZ7 Tanzania 2016 54.55% 29.07% 50.98% 17.21% UG6 Uganda 2016 55.30% 22.39% 20.48% 17.21% <	10.22% 20.27%	9.79%	20.70%
SL7Sierra Leone2019 50.71% 22.53% 13.32% 19.79% SN6Senegal2011 36.59% 8.63% 27.95% 21.26% TD6Chad2013 30.59% 8.63% 27.95% 21.26% TD6Chad2014 41.67% 19.27% 22.39% 25.08% TJ6Tjgikistan2012 40.58% 19.18% 21.40% 15.91% TL5Timor Leste2010 35.87% 13.50% 22.39% 25.08% TL5Timor Leste2010 35.87% 13.50% 22.37% 15.91% TL7Timor Leste2010 54.55% 29.07% 50.98% 17.21% TZ7Tanzania2016 76.00% 25.02% 22.48% 16.04% UG6Uganda 2016 51.04% 29.28% 32.43% 20.59% UG7Uganda 2016 51.04% 29.28% 32.43% 20.59% UG7Uganda 2016 51.04% 29.28% 32.43% 20.59%	20.03% 8.15%	18.59%	9.59%
SN6 Senegal 2011 36.59% 8.63% 27.95% 21.26% TD6 Chad 2015 20.00% 3.98% 16.02% 18.38% TG6 Togo 3.014 41.67% 19.12% 23.09% 25.08% TJ5 Timor Leste 2012 40.58% 19.18% 21.40% 15.91% TL5 Timor Leste 2010 35.87% 13.50% 22.33% 28.52% TL7 Timor Leste 2010 35.87% 13.50% 22.37% 18.59% TL7 Timor Leste 2010 35.87% 15.68% 17.21% TZ7 Tanzania 2016 76.00% 25.02% 20.48% UG6 Uganda 2016 55.30% 22.16% 17.21% UG7 Uganda 2016 55.30% 22.16% 18.94% UG7 Uganda 2016 55.30% 22.16% 18.94%	19.79% - 6.47%		-5.37%
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	21.26% 6.70%	20.56%	7.40%
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	18.38% —2.36%	- 17.63%	-1.61%
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	25.08% —2.69%	- 24.28%	-1.88%
TL5 Timor Leste 2010 35.87% 13.50% 22.37% 18.52% TL7 Timor Leste 2016 30.90% 15.68% 15.22% 20.48% TZ5 Tanzania 2010 54.55% 29.07% 50.98% 17.21% TZ7 Tanzania 2016 76.00% 25.02% 25.48% 16.04% UG6 Uganda 2011 51.04% 29.28% 32.43% 20.59% UG7 Uganda 2016 55.30% 22.38% 21.76% 18.94% UG7 Uganda 2016 55.30% 22.38% 20.59%	15.91% 5.49%	15.37%	6.03%
TL7 Timor Leste 2016 30.90% 15.68% 15.22% 20.48% TZ5 Tanzania 2010 54.55% 29.07% 50.98% 17.21% TZ7 Tanzania 2016 76.00% 25.02% 25.48% 16.04% UG6 Uganda 2011 51.04% 29.28% 32.43% 20.59% UG7 Uganda 2016 55.30% 22.28% 32.43% 18.94% UG7 Uganda 2016 55.30% 22.87% 20.59% 20.59%	18.52% 3.85%	18.35%	4.01%
TZ5 Tanzania 2010 54.55% 29.07% 50.98% 17.21% TZ7 Tanzania 2016 76.00% 25.02% 25.48% 16.04% UG6 Uganda 2011 51.04% 29.28% 32.43% 16.04% UG7 Uganda 2016 55.30% 22.28% 32.43% 18.94% UG7 Uganda 2016 55.30% 22.28% 32.43% 18.94%	20.48% —5.27%	12.83%	2.39%
TZ7 Tanzania 2016 76.00% 25.02% 25.48% 16.04% UG6 Uganda 2011 51.04% 29.28% 32.43% 20.59% UG7 Uganda 2016 55.30% 22.28% 32.43% 20.59% UG7 Uganda 2016 55.30% 22.87% 21.76% 18.94%	17.21% 33.77%	16.22%	34.75%
UG6 Uganda 2011 51.04% 29.28% 32.43% 20.59% UG7 Uganda 2016 55.30% 22.87% 21.76% 18.94% T77 0.04.00000000000000000000000000000000	16.04% 9.44%	15.30%	10.18%
UG7 Uganda 2016 55.30% 22.87% 21.76% 18.94%	20.59% 11.84%	20.18%	12.25%
7.4.7 0.1.1.1.1.1.1.1.1.1.1.1.1.1.1.1.1.1.1.1	18.94% 2.82%	18.70%	3.07%
ZA/ South Africa 2010 b0.24% 4/.25% 15.01% 10.40%	10.40% 2.67%	9.94%	3.13%
ZM6 Zambia 2014 57.83% 34.46% 23.37% 14.60%	14.60% 8.77%	14.26%	9.11%
ZM7 Zambia 2018 49.48% 35.18% 14.29% 13.31%	13.31% 0.98%	12.54%	1.75%
ZW6 Zimbabwe 2011 71.93% 41.25% 30.68% 8.92%	8.92% 21.76%	8.63%	22.06%
ZW7 Zimbabwe 2015 75.06% 48.91% 26.15% 7.51%	7.51% 18.64%	6.29%	19.86%

IADLE 3 ESUMAIEU IGEAI COMU	acepuve use and attribut	adie variation dy selection criteria,	giodai analysis	
	(1)	(2)	(3)	(4)
	Ideal	Proportion of sample without	Autributable ideal contraceptive use variation	Percentage of total
	contraceptive use	factor	from factor adjustment	variation
Highest quintile	0.082***	0.801	0.066	41.12%
Currently married	0.0027	0.314	0.0024	1.50%
Sexually active in last month	0.208***	0.336	0.070	43.61%
High education	-0.192 - 0.224 -0.0054 0.037 0.036	0.894	-0.0048	-2.99%
Knows a FP method	-0.02/-0.020 0.333*** 0.331_0.335	0.042	0.014	8.72%
Distance to facility is not a problem	0.037*** 0.037*** 0.031–0.043	0.348	0.0129	8.04%
Observations R ²	970,943 0.151	Total attributable variation to ideal contraceptive use	0.161	
NOTE: The table presents results from a multiva below. The regression includes survey (country-) particular factor for which the coefficient estimat contraceptive use that would be gained after havi estimate obtained from Column 1 and the propon variation estimates from each of the five factors. variation from that factor as a percentage of the t	riable logistic regression model using year) fixed effects, and coefficient stan. te is calculated; for example, 89.4 perc ing adjusted for the particular factor. rition of the analytic sample without th and Column 4 presents the proportion total attributable variation. *P < 0.1, *	the full analytic sample of women. Estimates of the 1 dard errors are clustered at the survey level. Column eent of women in the analytic sample do not have a tte The attributable variation in ideal contraceptive use 1 at factor from Column 2. The total attributed to that 1 and the total variation that can be attributed to that $1 \approx 0.05$. ***P < 0.01.	predicted probabilities are presented in Column 12 presents the proportion of the full analytic san ertiary level of education. Column 3 presents the for a given factor is calculated by taking the proc ation in ideal contraceptive use is calculated by a particular factor, which is calculated by taking th	with 95% confidence intervals pple that does not have the attributable variation in ideal uct of the factor's coefficient umming up the attributable e ratio of the attributable

the correlation between high education and ideal contraceptive use to be small and negative, highlighting: (1) a potentially inverse association between educational attainment and contraceptive use at high levels of educational attainment, which has been observed in high-income settings (Kravdal 2001); and (2) the relatively small variation in ideal contraceptive use that

Calculating Unmet Need by Matching: An Extension

can be attributed to high educational attainment compared to other factors.

An assumption of the counterfactual approach to calculating unmet need is that the sample of women from an ideal family planning environment should be representative of the full population of women on characteristics that are "nonmodifiable," such as age, religion, or ethnicity, which define the population. Specifically, one would expect that the distribution of age, religion, ethnicity, and other such invariant characteristics should be the same (not statistically different) between women from ideal environments and the general population of women. Given limited sample sizes, however, it may be the case that a comparison of the empirical distributions between women from ideal environments and the general population of women may yield differences. To account for these potential sampling differences, we extend our analysis as follows:

- (1) Within each DHS survey, we pair each woman from ideal family planning environments with a woman from nonideal environments on age and ethnicity using nearest-neighbor exact matching with the propensity score as the distance metric (Rosenbaum and Rubin 1983; Becker and Ichino 2002; Leuven and Sianesi 2003). Matching on these two characteristics would guarantee that the empirical distributions of these two characteristics are identical between groups.
- (2) Following matching, we can proceed to calculate unmet need as before by calculating the difference in contraceptive prevalence between women from ideal family planning environments and the matched subgroup of women who are not from ideal environments.

Table 6 presents estimates of unmet need using the matching approach, and Figure 6 compares the distributions of predicted unmet need between the matched counterfactual methodology against the current DHS methodologies. Similar to the previous results, we find that the distribution of estimates of unmet need under the matched approach is wider than the distributions under the traditional measures of unmet need, implying that the matched approach yields more extreme estimates of unmet need. However, we observe that the variation in the estimates of unmet need under the matched approach is less dispersed than under the first counterfactual approach that was presented in Table 3 and Figure 2, implying that matching would likely have improved the efficiency in the estimator's approach to calculating unmet need.

CONCLUSIONS

Unmet need has been a key indicator in family planning and reproductive health for more than four decades. It is an indicator that holds significant policy and programmatic weight

numor la montro	Country	iCP ^{MAI} UI	CP_{ct}^{MATCH}	Unmet Need, Matched	Matched – Original	Matched – Def. 1	Matched - Def. 2
A F7	Afghanistan	36.92%	20.56%	16.36%	1.04%	-9.03%	-8.46%
AL7	Albania	49.88%	35.66%	14.22%	6.54%	1.49%	1.83%
lF6	Burkina Faso	63.77%	28.99%	34.78%	13.30%	15.07%	15.69%
8]6	Benin	38.24%	13.73%	24.51%	-0.08%	-0.37%	0.20%
517	Benin	35.52%	14.75%	20.77%	0.21%	-4.76%	-4.12%
D6	DRC	52.36%	21.47%	30.89%	4.61%	9.26%	10.25%
0G6	Congo, Republic	61.90%	52.38%	9.52%	10.01%	-4.78%	-4.43%
216	Cote d'Ivoire	36.36%	20.20%	16.16%	1.41%	-6.11%	-5.40%
3M6	Cameroon	55.81%	36.43%	19.38%	12.34%	1.12%	2.03%
SM7	Cameroon	43.20%	28.53%	14.67%	8.16%	-2.97%	-2.06%
.T6	Ethiopia	58.31%	26.78%	31.53%	8.85%	16.43%	17.11%
T7	Ethiopia	54.12%	33.18%	20.94%	12.07%	8.97%	9.36%
A6	Gabon	52.78%	27.78%	25.00%	-3.98%	2.80%	4.19%
H6	Ghana	35.75%	23.46%	12.29%	1.31%	-7.50%	-6.39%
M6	Gambia	22.83%	8.66%	14.17%	1.98%	-3.42%	-2.88%
'N7	Guinea	31.97%	10.88%	21.09%	0.05%	4.02%	4.71%
iU6	Guinea	76.24%	53.93%	22.31%	15.04%	14.38%	14.56%
IN6	Honduras	81.07%	57.46%	23.61%	9.33%	16.80%	16.99%
46	India	51.96%	49.63%	2.34%	12.30%	-7.95%	-7.32%
07	Jordan	49.57%	48.01%	1.56%	0.57%	-12.26%	-11.75%
E6	Kenya	64.08%	55.51%	8.57%	16.80%	-3.74%	-3.24%
B6	Liberia	32.41%	20.37%	12.04%	0.11%	-15.79%	-14.32%
IL6	Mali	38.24%	13.24%	25.00%	2.64%	6.16%	6.65%
1L7	Mali	35.51%	21.50%	14.02%	6.97%	-8.27%	-4.07%
1W5	Malawi	58.52%	41.48%	17.04%	6.07%	-1.81%	-1.26%
1Z6	Mozambique	46.89%	28.25%	18.64%	12.82%	-0.13%	0.71%
G6	Nigeria	44.69%	30.46%	14.22%	14.44%	0.99%	1.63%
[G7	Nigeria	34.64%	22.10%	12.54%	8.61%	-2.58%	-1.89%
P6	Nepal	59.45%	49.62%	9.82%	11.39%	-9.76%	-9.55%
IP7	Nepal	61.36%	45.20%	16.16%	4.31%	-2.36%	-2.08%
H6	Philippines	49.91%	47.92%	1.98%	13.77%	-9.60%	-8.63%
H7	Philippines	45.96%	44.46%	1.50%	10.83%	-10.17%	-8.91%
K6	Pakistan	53.83%	40.81%	13.02%	6.51%	-7.66%	-7.59%
L6	Sierra Leone	50.71%	30.14%	20.57%	7.61%	0.53%	1.98%
L7	Sierra Leone	37.16%	27.59%	9.58%	3.75%	-10.22%	-9.12%
N6	Senegal	36.59%	17.07%	19.51%	8.44%	-1.75%	-1.04%
D6	Chad	20.83%	12.50%	8.33%	7.68%	-10.05%	-9.29%
G6	Togo	41.67%	18.75%	22.92%	-0.52%	-2.17%	-1.36%
JG6	Uganda	55.30%	28.79%	26.52%	5.91%	5.92%	6.34%
JG7	Uganda	51.04%	40.83%	10.21%	11.55%	-8.73%	-8.49%
CA7	South Africa	60.34%	55.17%	5.17%	7.90%	-5.22%	-4.76%
.M6	Zambia	57.63%	43.58%	14.04%	9.32%	-0.56%	-0.21%



FIGURE 6 Kernel density plots, matched unmet need across definitions

and serves an important role in advocacy, resource allocation, and agenda setting in family planning. At the same time, it is recognized to be a biased measure that is difficult to operationalize. As a result, a number of definitions of unmet need have been used over time and have resulted in estimates that are not comparable with each other and have limited scope for unbiased inference.

In this study, we use a counterfactual approach to derive a simplified operational definition of unmet need that can be consistently applied over time and across countries. In conducting a number of empirical analyses of our new indicator, we find significant mean differences between our approach and existing DHS approaches to calculating unmet need, whereby our country-level unmet need estimates are on average 5–6 percentage points higher than the estimates that are calculated with the current DHS algorithm. In addition, we observe significant variation across countries when comparing the estimates that are generated by the different approaches, which should be noted when interpreting what our higher estimates could mean for the design of policies and programs aimed to address unmet need.

Our indicator is preferable to existing measures due to its conceptual appeal and grounding in revealed preference theory through observed behavior; its independence from biases that are generated through the use of reported preferences and other problematic assumptions that typically form the foundation of such indicators, instead exclusively relying on observable characteristics or behavior; its simplicity in its derivation, which is based on only six routinely collected survey items that do not require additional preference-specific modules; and its flexibility to be both generalizable for cross-country comparisons as well as tailored 7284465, 2022, 4, Downloaded from https://anlinelibrary.wiley.com/doi/10.1111/sifp.12216, Wiley Online Library on [1404/2023]. See the Terms and Conditions (https://onlinelibrary.wiley.com/terms-and-conditions) on Wiley Online Library for rules of use; OA articles are governed by the applicable Creative Commons License

for context-specific analyses. Our indicator is also conceptually preferable in its estimation as a point prevalence, which clearly defines the measure to be a *population-level* statistic that estimates a population parameter. In being a population statistic, the measure is therefore not appropriate for assessing unmet need at a more disaggregated (i.e., individual) level; in fact, even the current definition of unmet need (which is also a population-level measure) should also only be used for population-level inference and never, for example, as an individual-level dependent variable.

From a perspective of implementation, our method presents a more attractive approach to calculating unmet need for these and a number of other related reasons. The simplicity of the method and ease with which unmet need can be calculated relative to the current algorithm are particular advantages. That our method arguably presents a more intuitive understanding of unmet need as a gap between current contraceptive practice and a conceptually ideal reference point also facilitates its adoption by policy makers and practitioners alike. The fact that our approach compares the current state of contraceptive use to a reference point that is calculated based on proposed determinants of empowerment highlights the central role that reproductive and contraceptive autonomy should play in the determination of unmet need. Specifically, the process for selecting factors to identify women from "ideal" environments reorients researchers, policy makers, and practitioners to recognize that the standard of comparison that should be achieved is a state of the world where women are fully able to make their own contraceptive and reproductive choices.

Previous studies of unmet need have raised a key question: "What is desirable contraceptive coverage in the 'perfect contracepting' society, and what principles should guide the answer to this large question?" (Bradley and Casterline 2014). Our study answers this question by estimating what contraceptive coverage would be in an environment where women have the capability to "perfectly contracept" if they choose. We define an approach for identifying this environment, and the proportion of women who belong to this environment, using observable factors related to reproductive empowerment and well-being as a proxy for this ideal environment. We then propose that contraceptive coverage in such an environment would reflect the ideal level of contraceptive use.

We note that taking this approach highlights an important conceptual distinction between our measure of unmet need and the current standard. By operationalizing unmet need as the gap between women's observed contraceptive (non)use and their desire to space or limit births, the current standard establishes an inextricable equivalence relation between women's preferences for spacing or limiting pregnancies and their contraceptive preference. Specifically, it considers a functional mapping correspondence where all possible fertility preferences can be identified as a countable set that contains each unique preference as an element, and where every unique fertility preference can be mapped to at most one contraceptive preference, which can be defined generally in terms of preferences over (non)use and over methods. In the standard framework, the set of fertility preferences can effectively be collapsed to the following mutually exclusive elements: (1) a preference for spacing births; (2) a preference for limiting births; or (3) a preference for neither spacing nor limiting, that is, having children soon.

Under the injective function mapping, a woman who prefers to space or limit births would *necessarily* have a demand for contraception to meet this fertility preference;

conversely, a women who prefers to not use contraception would not have a preference for limiting or spacing births for at least two years (that is, she has revealed her preference to have children soon through her nonuse). However, one can identify numerous counterexamples where (1) women who prefer to space or limit births also have well-defined preferences for not using contraception (e.g., due to religious opposition, fear of contraceptive-related side effects, etc.), and (2) women who have a preference for using contraception currently may also have a preference for having children soon. Each of these counterexamples are considered to be violations of the implied function mapping where fertility preferences are outcomes to which contraceptive preferences are expected to align. More generally, it is also possible to conceive of reasons for women's contraceptive use that are neither causes nor consequences of fertility control. For example, women may choose to use barrier methods of contraception specifically for preventing sexually transmitted diseases. If this is true, then the conceptualization of the set of outcomes on which contraceptive preferences are mapped would need to be expanded beyond fertility, to the extent that contraceptive preferences and demand would need to be decoupled from fertility preferences, thereby establishing a case for the examination of preferences for contraception independently of fertility intention. By extension, the use of an unbiased measure of fertility preferences, even if such an indicator could be estimated, to calculate unmet need may therefore be incorrect. Taking these points into consideration, our approach presents a clear advantage over the current methods in that it infers a singular mapping between contraceptive practice and contraceptive preference and makes no problematic assumptions on women's fertility preferences, desires, intentions, or wishes.²

In critically reflecting on our approach, a first order of concern is that over our choice of selecting variables. In theory, we would aim to and be able to select on as many variables that, together, identify the subset of women in a population who likely face no constraints to their family planning and reproductive health decision-making. Although conceptually appealing, one of the challenges to identify these women in surveys like the DHS is that there may be very few women who fit into this highly selective subpopulation, which would limit the statistical precision with which prevalence can be estimated in some surveys. To this end, there is a direct trade-off between the marginal utility from including a characteristic to screen on and select women from "ideal" environments and the resulting size of the subsample of women who belong to these more selective environments. With this said, if obtaining larger samples, with larger potential to identify those women who have greater contraceptive autonomy, were feasible, then we can be more confident that our estimates of ideal contraceptive prevalence, and hence unmet need, would converge to the true value of unmet need.

Our approach to operationalizing unmet need in this exercise is not without its own limitations. From an empirical standpoint, applying our method yield a small number of observations that are selected to estimate the ideal contraceptive prevalence estimate, iCP_{ct} , which in turn would generate less statistically precise estimates of unmet need. This statistical variability may in fact explain some of the large differences that we observe in our estimates

² On a side note, we have generally been confused by the family planning field's synonymous use of the words "desires," "preferences," "intentions," "beliefs," and "wishes." Though theoretically distinct (Casterline and El-Zeini 2007), each of these words seem to describe the same empirically reported phenomenon, which we would term "fertility preferences." We have yet to be provided with an explanation or (more convincingly) with empirical data that demonstrates why the term, for all practical purposes, should not be fixed.

over time and within countries. Given that our estimate is derived using point prevalence measures, we are only able to generate unbiased inference at the moment when women's revealed preferences, as indicated by their current contraceptive use, is reported in the survey. Specifically, our estimate implies that at the time of interview, the contraceptive use for women who are identified to live in ideal environments reflect their true contraceptive preferences at that moment. To this end, our approach, as well as current approaches to measuring unmet need, limits the extent to which we are able to interpret contraceptive use over time, when preferences can vary—this may likely be the key cost that we pay for simplicity and ease of calculation. Our approach is comparable to recent approaches that estimate unmet need as a point prevalence (Moreau et al. 2019); however, we deliberately do not include measures of fertility preference and/or contraceptive intention, both of which are likely to be biased in traditional survey data.

Like other approaches, our metric for unmet need does not account for husband or partner preferences for contraception. To the extent that male preferences can be included using the women's DHS datasets, we run a sensitivity analysis that selects ideal women based on whether they had a partner who supported them in their contraceptive use/nonuse. The inclusion of partner approval of contraception does not seem to significantly change our empirical estimates of unmet need. On a broader conceptual note, however, it is not clear as to how one would in fact calculate unmet need by including male preferences, particular in the case where there is discordance in preferences between men and women (Ashraf, Field, and Lee 2014; Karra and Zhang 2021)—would a couple have an unmet need for contraception if women want to use contraceptive methods but their male partners do not? What about the converse? In a sense, the (lack of) inclusion of male preferences in the unmet need measurement speaks to the tension between women's reproductive rights over contraceptive choice and male involvement in contraceptive decision-making that the field continues to debate.

Given that our approach questions the utility of direct preference elicitation through surveys, our findings also call for a critical review of existing surveys and a reprioritization of survey questions that are currently asked as part of large-scale data collection efforts, like the DHS. Our study specifically calls for the substitution away from the use of problematic fertility preference questions that are known to be biased from the onset and toward a wider and more inclusive range of observable metrics that would serve to capture latent constructs related to contraceptive autonomy, reproductive empowerment, family planning access, and well-being. In the absence of any changes to the current data collection efforts, we encourage future efforts in this domain to continue testing a wider range of factors that capture women's ideal reproductive health environments to determine the extent to which ideal contraceptive use, and by extension unmet need, are sensitive to these choices.

DATA AVAILABILITY STATEMENT

All data that are used for this study are available for free download after registering with the DHS Program at http://dhsprogram.com/data/

CONFLICTS OF INTEREST

I declare that no conflicts of interest exist, and all errors are my own.

679

REFERENCES

- Ami, Dominique, Frédéric Aprahamian, and Stéphane Luchini. 2017. "Stated Preferences and Decision-Making: Three Applications to Health." *Revue economique* 68 (3): 327–333.
- Anand, Paul, Graham Hunter, and Ron Smith. 2005. "Capabilities and Well-Being: Evidence Based on the Sen–Nussbaum Approach to Welfare." Social Indicators Research 74 (1): 9–55. https://doi.org/10.1007/s11205-005-6518-z.
- Ashraf, Nava, Erica Field and Jean Lee. 2014. "Household Bargaining and Excess Fertility: An Experimental Study in Zambia." American Economic Review 104 (7): 2210–2237. https://doi.org/10.1257/aer.104.7.2210.
- Bachrach, Christine A. and S. Philip Morgan. 2013. "A Cognitive-Social Model of Fertility Intentions." Population and Development Review 39 (3): 459–485. https://doi.org/10.1111/j.1728-4457.2013.00612.x.
- Bearak, Jonathan, Anna Popinchalk, Leontine Alkema and Gilda Sedgh. 2018. "Global, Regional, and Subregional Trends in Unintended Pregnancy and Its Outcomes from 1990 to 2014: Estimates from a Bayesian Hierarchical Model." *The Lancet Global Health* 6 (4): e380–e389. https://doi.org/10.1016/S2214-109X(18)30029-9.
- Becker, Sascha O. and Andrea Ichino. 2002. "Estimation of Average Treatment Effects Based on Propensity Scores." The Stata Journal 2 (4): 358–377. https://doi.org/10.1177/1536867x0200200403.
- Bhrolcháin, Máire Ní and Éva Beaujouan. 2019. "Do People Have Reproductive Goals? Constructive Preferences and the Discovery of Desired Family Size." In Analytical Family Demography, edited by Robert Schoen, vol. 47, 27–56. The Springer Series on Demographic Methods and Population Analysis. Cham: Springer International Publishing. https://doi.org/10.1007/978-3-319-93227-9_3.
- Bongaarts, John. 1990. "The Measurement of Wanted Fertility." *Population and Development Review* 16 (3): 487–506. https: //doi.org/10.2307/1972833.
- Bongaarts, John, W. Parker Mauldin and James F. Phillips. 1990. "The Demographic Impact of Family Planning Programs." Studies in Family Planning 21 (6): 299–310. https://doi.org/10.2307/1966918.
- Bradley, Sarah E. K. and John B. Casterline. 2014. "Understanding Unmet Need: History, Theory, and Measurement." Studies in Family Planning 45 (2): 123–150. https://doi.org/10.1111/j.1728-4465.2014.00381.x.
- Bradley, Sarah E. K., Trevor Croft, Joy Fishel and Charles Westoff. 2012. Revising Unmet Need for Family Planning. DHS Analytical Studies No. 25. Calverton, MD: ICF International.
- Brown, S. S. and L. Eisenberg. 1995. *The Best Intentions: Unintended Pregnancy and the Well-Being of Children and Families*. Washington, DC: National Academies Press.
- Casterline, John B. and Samuel Agyei-Mensah 2017. "Fertility Desires and the Course of Fertility Decline in Sub-Saharan Africa." *Population and Development Review*, 43 (S1): 84–111. https://doi.org/10.1111/padr.12030
- Casterline, John B. and Laila O. El-Zeini. 2007. "The Estimation of Unwanted Fertility." *Demography* 44 (4): 729–745. https://doi.org/10.2307/30053115.
- Cleland, John, Sarah Harbison and Iqbal H. Shah. 2014. "Unmet Need for Contraception: Issues and Challenges." Studies in Family Planning 45 (2): 105–122. https://doi.org/10.1111/j.1728-4465.2014.00380.x.
- Cleland, John, Kazuyo Machiyama and John B. Casterline. 2020. "Fertility Preferences and Subsequent Childbearing in Africa and Asia: A Synthesis of Evidence from Longitudinal Studies in 28 Populations." *Population Studies* 74 (1): 1–21. https: //doi.org/10.1080/00324728.2019.1672880.
- Darroch, J. E., S Audam, A Biddlecom, G. Kopplin, T. Riley, S. Singh and E. Sully. 2017. Adding It Up: Investing in Contraception and Maternal and Newborn Health, 2017. New York, NY: Guttmacher Institute.
- Darroch, Jacqueline E., Gilda Sedgh and Haley Ball. 2011. "Contraceptive Technologies: Responding to Women's Needs." New York, NY: Guttmacher Institute.
- Hsiao, Cheng. 1985. "Benefits and Limitations of Panel Data." *Econometric Reviews* 4 (1): 121–174. https://doi.org/10.1080/ 07474938508800078.
- Karra, Mahesh, S. V. Subramanian and Günther Fink. 2017. "Height in Healthy Children in Low- and Middle-Income Countries: An Assessment." *The American Journal of Clinical Nutrition* 105 (1): 121–126. https://doi.org/10.3945/ajcn.116.136705.
- Karra, Mahesh and Kexin Zhang. 2021. "User-Centered Counseling and Male Involvement in Contraceptive Decision Making: Protocol for a Randomized Controlled Trial." *JMIR Research Protocols* 10 (4): e24884. https://doi.org/10.2196/24884.

- Kravdal, Øystein. 2001. "The High Fertility of College Educated Women in Norway: An Artefact of the Separate Modelling of Each Parity Transition." *Demographic Research* 5: 187–216.
- Leuven, Edwin and Barbara Sianesi. 2003. "PSMATCH2: Stata Module to Perform Full Mahalanobis and Propensity Score Matching, Common Support Graphing, and Covariate Imbalance Testing." Statistical Software Components, Boston College Department of Economics.
- Lloyd, Cynthia and John A. Ross. 1992. "Methods for Measuring the Fertility Impact of Family Planning Programs: The Experience of the Last Decade." Population Council Working Paper No. 7, Population Council.
- Loewenstein, George, Ted O'Donoghue and Matthew Rabin. 2003. "Projection Bias in Predicting Future Utility." Quarterly Journal of Economics 118 (4): 1209–1248. https://doi-org.ezproxy.bu.edu/10.1162/003355303322552784.
- Moreau, Caroline, Mridula Shankar, Stephane Helleringer and Stanley Becker. 2019. "Measuring Unmet Need for Contraception as a Point Prevalence." *BMJ Global Health* 4 (4): e001581. https://doi.org/10.1136/bmjgh-2019-001581.
- Müller, Maximilian W., Joan Hamory, Jennifer Johnson-Hanks and Edward Miguel. 2022. "The Illusion of Stable Fertility Preferences." *Population Studies* 76 (2): 169–189. https://doi.org/10.1080/00324728.2022.2057577.
- Newman, Karen and Charlotte Feldman-Jacobs. 2015. Family Planning and Human Rights—What's the Connection and Why Is It Important? Washington, DC: Population Reference Bureau.
- Nussbaum, Martha and Amartya Sen. 1993. The Quality of Life. Oxford: Clarendon Press.
- Rosenbaum, Paul R. and Donald B. Rubin. 1983. "The Central Role of the Propensity Score in Observational Studies for Causal Effects." *Biometrika* 70 (1): 41–55. https://doi.org/10.1093/biomet/70.1.41.
- Rosenzweig, Mark R. and Kenneth I. Wolpin. 1993. "Maternal Expectations and Ex Post Rationalizations: The Usefulness of Survey Information on the Wantedness of Children." *The Journal of Human Resources* 28 (2): 205–229. https://doi.org/10. 2307/146201.
- Senderowicz, Leigh. 2020. "Contraceptive Autonomy: Conceptions and Measurement of a Novel Family Planning Indicator." Studies in Family Planning 51 (2): 161–176. https://doi.org/10.1111/sifp.12114.
- Singh, S. 1998. "Adolescent Childbearing in Developing Countries: A Global Review." Studies in Family Planning 29 (2): 117–136.
- Trinitapoli, Jenny and Sara Yeatman. 2018. "The Flexibility of Fertility Preferences in a Context of Uncertainty." *Population and Development Review* 44 (1): 87–116. https://doi.org/10.1111/padr.12114.
- United Nations. 2018. Sustainable Development Goals. New York, NY: United Nations. https://sdgs.un.org/.

ACKNOWLEDGMENTS

I thank Michelle Hindin, David Canning, Iqbal Shah, Amanda Kalamar, Mihira Karra, and Jessica Gipson for their feedback and comments. This work was supported by the Program for Women's Empowerment Research (POWER) at the Boston University Global Development Policy Center through a grant from the William and Flora Hewlett Foundation (Grant No. 2020-1162). The study funders had no role in study design, data collection, analysis, interpretation, or writing of the results.