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Publication Date
2016-08-15

DOI
10.1186/s12960-016-0145-4

Peer reviewed
Examining the influence of country-level and health system factors on nursing and physician personnel production

Allison Squires¹, ²*, S. Jennifer Uyei³, Hiram Beltrán-Sánchez⁴ and Simon A. Jones⁵

Abstract

Background: A key component to achieving good patient outcomes is having the right type and number of healthcare professionals with the right resources. Lack of investment in infrastructure required for producing and retaining adequate numbers of health professionals is one reason, and contextual factors related to socioeconomic development may further explain the trend. Therefore, this study sought to explore the relationships between country-level contextual factors and healthcare human resource production (defined as worker-to-population ratio) across 184 countries.

Methods: This exploratory observational study is grounded in complexity theory as a guiding framework. Variables were selected through a process that attempted to choose macro-level indicators identified by the interdisciplinary literature as known or likely to affect the number of healthcare workers in a country. The combination of these variables attempts to account for the gender- and class-sensitive identities of physicians and nurses. The analysis consisted of 1 year of publicly available data, using the most recently available year for each country where multiple regressions assessed how context may influence health worker production. Missing data were imputed using the ICE technique in STATA and the analyses rerun in R as an additional validity and rigor check.

Results: The models explained 63 % of the nurse/midwife-to-population ratio (pseudo $R^2 = 0.627, p = 0.0000$) and 73 % of the physician-to-population ratio (pseudo $R^2 = 0.729, p = 0.0000$). Average years of school in a country’s population, emigration rates, beds-per-1000 population, and low-income country statuses were consistently statistically significant predictors of production, with percentage of public and private sector financing of healthcare showing mixed effects.

Conclusions: Our study demonstrates that the strength of political, social, and economic institutions does impact human resources for health production and lays a foundation for studying how macro-level contextual factors influence physician and nurse workforce supply. In particular, the results suggest that public and private investments in the education sector would provide the greatest rate of return to countries. The study offers a foundation from which longitudinal analyses can be conducted and identifies additional data that may help enhance the robustness of the models.

Keywords: Human resources for health, Socioeconomic development, Policy, Global health, Complexity theory, Nurse-to-population ratio, Physician-to-population ratio

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Background
A key component to achieving good patient outcomes is having the right number and type of healthcare professionals with the right resources. Research increasingly demonstrates where and how the healthcare workforce impacts patient outcomes, quality of care, and health inequalities in both primary and acute care settings [1–12]. Political and social determinants of health also contribute to patient outcomes to varying degrees [13].

The production and retention of human resources for health (HRH) at levels that can sustainably achieve good health outcomes seems like it should be the simple case of having enough schools, teachers, and resources to prepare the future healthcare worker to address the healthcare needs of a country’s population, followed by appropriate management and governance that creates supportive practice environments. If it were that simple, Latin America and former Soviet Union countries, for example, would not overproduce physicians and underproduce nurses [14]. As regional examples of this trend, the country of Georgia has 3.2 nurses and midwives per 1000 population and 4.8 physicians while Colombia has 0.6 nurses per 1000 population and 1.5 physicians per 1000 population. Navarro et al. [15] produced one of the few studies examining this dynamic, but focused their analysis on health outcomes and not health worker production. Countless other examples from around the world suggest that resources and infrastructure are only a small part of the healthcare human resource production equation.

Therefore, this study seeks to explore if country-level contextual factors have an impact on HRH production and, if so, to what extent. We define country-level contextual factors as those broader institutional structures that affect, directly or indirectly, the healthcare system, population health, and health worker supply and demand. We take the health worker-to-population ratio as a measure of production instead of availability because a measure of health worker availability means the worker is trained and employed as health workers [16]. Since we know health workers may maintain a license or credential even if they are not working in the country (due to domestic un- or underemployment or because they are working abroad) and many countries do not have accurate and updated health worker records (e.g., workers who are alive vs. dead), for the purposes of this paper the measure is used as a marker of production since good quality data about health worker availability is not consistently available. Our focus for this paper is limited to physicians and nurses/midwives since they represent the largest composite group of professional healthcare workers around the world and have the most reliable data sources. We hypothesize that the identified indicators are differentially linked with physician and nurse supply and study these groups separately.

To date, only a few studies have looked at the context of HRH production at a macro-level [17–19]. Squires and Beltrán-Sánchez [20, 21], for example, conducted two studies that demonstrated a link between educational variables at the State level and the nurse-to-population ratio in Mexico, with average years of education correlating strongly with the overall ratio ($R = 0.69$, $p = 0.001$) and to the ratio of nurses with specialty training (e.g., operating room, intensive care, obstetrics) ($R = 0.54$, $p = 0.020$). Generally, however, the literature quantifying HRH production is scarce, comes largely from high-income countries in the form of policy statements, and examines the phenomenon mostly through a supply side lens. At the micro-level, research examines the career intentions of students [22–24], the impact on worker production of teacher shortages [25–27], and education system dynamics that affect supply [28–31]. Research about demand for health workers from epidemiological and demographic forecasts take a broader view but do not consider how a profession itself creates “demand” for individuals to attempt to gain entry into it (via the public image, perceived employment stability, salary, etc.) and other factors that may influence their ability to do so. Thus, exploring production issues just at the micro-level ignores the complex contextual picture that is involved with HRH production. It does not consider how economic indicators, political factors, and social inequality may influence the production process. This analysis will explore the influence of those factors on HRH production using physicians and nurses as examples.

Methods
The design of this observational study draws from multiple disciplines, including the social sciences, health services research, and public health. The theoretical grounding for the study rests in complexity theory combined with a realist perspective. The combined perspectives “understand reality as comprising multiple, nested, open system in which change is generative, context dependent, and time irreversible,” (Westhorp, [32], p. 406), further positing that causation occurs through mechanisms that are context sensitive, and capture how micro-level interactions translate into macro-level outcomes [32, 33]. In the case of healthcare human resource production, this theory provides the structure for understanding how socioeconomic development, political systems, and other national social factors influence HRH production because it is a context-dependent process involving the interactions between multiple systems.

Variable selection drew from a critical review of interdisciplinary literature from healthcare and the social sciences and expert consultation. They represent population characteristics, health system factors, economic variables known to directly affect health systems, a measure
of political stability, and variables accounting for social inequality in a society. In particular, to attempt to account for the effects of gender-sensitive social inequality, we use the United Nations Development Program's (UNDP) Gender Empowerment Measure (GEM). We selected that variable with the idea that professions that were biased toward one gender or another may have their production affected due to the country's specific dynamic in that area. The combination of these variables attempts to account for the gender- and class-sensitive identities of physicians and nurses. Our variable selection and rationales for choosing them are found in Table 1.

We use nurse/midwife-to-population ratio (NMPR) and the physician-to-population ratio (PPR) as indicators of HRH production. Combining nurses and midwives into a single aggregated indicator is standard practice for data reporting by the World Health Organization (WHO) because in some countries, nurses are dually trained as midwives or midwives may not have formal training and are categorized as community health workers [34]; therefore, we opted for the aggregated indicator because we felt it was a better representative.

For each indicator, we then conducted a series of negative binomial regression models at the country level while controlling for factors shown in Table 1 and using the country’s population size as an offset. The latter step allows us to control for differences in population size across countries. We began the analysis by dividing the variables into five groups as shown in Table 1: population characteristics, health system, political system, economic, and social inequality. Logs were generated for the economic variables “external debt” and “Gross Domestic Purchasing Power Parity (GDPPP)” as a way to reduce variability, a standard practice in economic analyses. For the polity variable, we used the three categories recommended by the Polity IV project for analysis: democracy, anocracy, and autocracy [35]. Figure 1 from the Polity IV project illustrates the 2011 categorizations of global states, which have remained stable since that year.

We began with univariate and bivariate analyses and then proceeded to the multivariate models. Multicollinearity checks based on the variance inflation factor suggested no collinearity issues since all measures were less than ten. The first model included population and health system characteristics, as they represent the variables

<table>
<thead>
<tr>
<th>Table 1 Variable selection rationale and data sources</th>
<th>Data source</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Population characteristics</strong></td>
<td></td>
</tr>
<tr>
<td>Schooling: Average years of school of a country’s population. Educational access is a key to producing health workers.</td>
<td>CIA.gov</td>
</tr>
<tr>
<td>Migrate: Migration rates of a country (in and out). Migration is known to affect health workforce supply.</td>
<td>CIA.gov</td>
</tr>
<tr>
<td>Urbpop: Urban population percentage. Health workers are known to cluster in urban areas.</td>
<td>CIA.gov</td>
</tr>
<tr>
<td><strong>Population health status</strong></td>
<td></td>
</tr>
<tr>
<td>Beds-per-1000 people: Reflects health system capacity for worker employment.</td>
<td>WorldBank.org</td>
</tr>
<tr>
<td>Public health exp: Healthcare human resources are the largest expenditure of a health system.</td>
<td>WorldBank.org</td>
</tr>
<tr>
<td>Private health exp:</td>
<td>WorldBank.org</td>
</tr>
<tr>
<td><strong>Economic</strong></td>
<td></td>
</tr>
<tr>
<td>Gini: The Gini index is considered the best measure of economic inequality in a country. Economic inequality affects health outcomes because it may affect access to health workers.</td>
<td>WorldBank.org</td>
</tr>
<tr>
<td>WB Income Level: The World Bank classifies countries by four income levels: low income, low middle-income, high middle-income, and high income. It is the best way to control for between country variations in economic status.</td>
<td>WorldBank.org</td>
</tr>
<tr>
<td>Debtext: External debt affects what a country can spend on healthcare.</td>
<td>CIA.gov</td>
</tr>
<tr>
<td>GDPPPP: Gross Domestic Purchasing Power Parity is a proxy measure of national income that better accounts for inequality than GDP per capita.</td>
<td>CIA.gov</td>
</tr>
<tr>
<td><strong>Political System</strong></td>
<td></td>
</tr>
<tr>
<td><strong>Social inequality</strong></td>
<td></td>
</tr>
<tr>
<td>GEM: The Gender Empowerment Measure is how the United Nations Development Program attempts to measure gender inequality. It is a composite measure of women’s relative economic income, participation in high-paying positions with economic power, and participation in governance.</td>
<td>UNDP.org</td>
</tr>
</tbody>
</table>

*The most recently available years of data were used in the analyses*
closest to the actual healthcare worker. We then tested the impact of political and economic variables on the model in the second iteration. The final model included a social inequality indicator variable, the gender empowerment measure (GEM), which is commonly used in United Nations analyses of gender.

The majority of our independent variables had less than 4% missing data, but three had substantial data missing (polity 14%, Gini 40%, and GEM 49%). All variables were imputed using chained equation (ICE), a technique available in STATA software, and models were run with and without imputation to check the degree to which the results were influenced by imputation. A pseudo $R^2$ result was produced to explain the variation in a model because in the case of this analysis, no direct ways of estimating $R^2$ exist. Overall, results were robust to the imputation approach with consistent coefficients in magnitude and direction with and without imputation. All analyses were conducted in STATA, then re-run in R as a validity check. Results from additional checks related to multicollinearity and the imputation approach can be found in the "Additional File 1" online.

**Results**

Figures 1 and 2 illustrate the bivariate relationships between the independent and dependent variables for nurses and physicians, respectively. All variables were significantly correlated with the NMPR, except for the net migration rate (Fig. 1). Private sector health expenditure and the Gini coefficient were negatively correlated with the NMPR at $R = -0.15$, $p = 0.04$, and $R = -0.26$, $p = 0.03$, respectively. For the PPR, net migration rate was positively and significantly linked but only slightly correlated ($R = 0.16$, $p = 0.03$) while private sector health expenditure was not correlated ($R = -0.07$, $p = 0.29$) (Fig. 2). Average years of schooling ($R = 0.60$, $p = 0.00$), GDP ($R = 0.53$, $p = 0.00$), and beds-per-1000 population ($R = 0.49$, $p = 0.00$) were the top three variables positively correlating with the NMPR. Similarly, the PPR also had average years of schooling ($R = 0.71$, $p = 0.00$) and beds-per-1000 population ($R = 0.65$, $p = 0.00$) in the top three, but urban population percentage ($R = 0.60$, $p = 0.00$) beat the GEM’s correlation ($R = 0.46$, $p = 0.00$) by 14 points. Gini coefficients had identical negative and significant correlations with the NMPR.
Further, external debt rates, while both significant, were twice as correlated to the PPR ($R = -0.26$, $p = 0.03$) as the NMPR ($R = -0.26$, $p = 0.03$).

Table 2 illustrates the results from the negative binomial regressions for nurses and physicians. The final model explained 63% of the NMPR ($R^2 = 0.627$, $p = 0.0000$) and 73% of the PPR ($R^2 = 0.729$, $p = 0.0000$) based on the maximum likelihood pseudo $R^2$ [36]. Notably, for every additional year spent in school, a country’s NMPR increases by 18% ($p = 0.001$) and PPR increases by 19% ($p = 0.001$). These results were consistently significant across the models. The beds-per-1000 population were also significant predictors of the total nurse/midwife (11% increase, $p = 0.001$) and physician (14.5% increase, $p = 0.001$) numbers and consistent across the models. Emigrant population rate was also significant for both types of providers, indicating that relative to countries with no migration, those with negative migration flows have significantly higher nursing ($p = 0.05$) and physician ($p = 0.001$) populations. Unsurprisingly, status as a low-income country relative to high income reduces nurse and physician supply by 30% and was significant for both ($p = 0.001$).

In addition, one variable was only significant for physician supply. Urban population percentage was a significant production predictor across the models for physicians ($p = 0.01$), but not for nurses. Also notable, a country categorized as an “anocracy,” relative to democracy, increases its nurse supply numbers by 42% ($p = 0.050$) but has no effect on physician supply.

With the exception of low-income country status, none of the other economic variables selected for this study were significant predictors of nurse or physician supply in the fully adjusted models; however, the models suggest some important trends to study further. A one-point increase in GDPPP translates into 5% more physicians but adds nothing to nursing supply. External debt has an opposite effect for nurses and physicians suggesting a negative effect on the nursing supply, reducing nurses by 4%, but a positive one on physician supply adding 1.4% more. Economic inequality, as measured by the Gini coefficient, appears to have a minute effect on decreasing physician supply but no effect on nursing numbers.
The strength of the models and their ability to explain 63 to 73 % of the relationship between context and production of the two largest health workforce cadres offers striking new insight into the complex set of dynamics involving health workforce production. For some countries, a grim picture may be present in these results, but the findings do provide some clearer direction for policy initiatives. They also suggest that targeted investments aimed at meeting the sustainable development goals (SDGs) may translate both directly (e.g., quality education, gender equality, good health, and well-being) and indirectly (e.g., no poverty, zero hunger, clean water, and sanitation) into improved health workforce production. More skilled and competent health workers are essential for meeting many of the SDGs.

In consideration of the models, notably comparing pseudo $R^2$ for models with different outcomes should be carried out with caution. To partially mitigate this risk, two other methods for estimating pseudo $R^2$ were used [37, 38] and a comparable difference was still observed. The 10 % difference in variance explained between physicians and nurses in the model, we believe, cannot be explained by existing data because of the amount of cross-national inconsistencies in information systems [39, 40]. Per Riley et al. [40], even if new data were collected, it may not be consistent, of good quality, nor vetted by internationally accountable authority.

In light of the aforementioned data availability and quality issues, an additional variable that may help align the NMPR model to results that are equivalent to the PPR model, along with enhancing its robustness, is the number of nursing and medical schools in each country. Publicly available data about the number of educational institutions would also help to further refine the

### Table 2
Coefﬁcient estimates expressed as incidence rate ratios (IRR) from a negative binomial model associating nurse/midwife-to-population ratio (NMPR) and physician-to-population ratio (PPR) with macro-level variables at the country level

<table>
<thead>
<tr>
<th>Variable</th>
<th>NMPR Model 1</th>
<th>NMPR Model 2</th>
<th>NMPR Model 3</th>
<th>PPR Model 1</th>
<th>PPR Model 2</th>
<th>PPR Model 3</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Population characteristics</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Schooling (years)</td>
<td>1.216***</td>
<td>1.183***</td>
<td>1.185***</td>
<td>1.221***</td>
<td>1.192***</td>
<td>1.189***</td>
</tr>
<tr>
<td>Migrate (ref = zero migration)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Emigrant</td>
<td>1.460*</td>
<td>1.441*</td>
<td>1.458*</td>
<td>2.121***</td>
<td>1.848***</td>
<td>1.757***</td>
</tr>
<tr>
<td>Immigrant</td>
<td>1.434</td>
<td>1.341</td>
<td>1.321</td>
<td>1.641**</td>
<td>1.449</td>
<td>1.456</td>
</tr>
<tr>
<td>Urban population (%)</td>
<td>1.003</td>
<td>0.999</td>
<td>0.999</td>
<td>1.019***</td>
<td>1.010**</td>
<td>1.010**</td>
</tr>
<tr>
<td><strong>Health system characteristics</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Beds-per-1000 people</td>
<td>1.104***</td>
<td>1.108***</td>
<td>1.110***</td>
<td>1.167***</td>
<td>1.151***</td>
<td>1.145***</td>
</tr>
<tr>
<td>Public health exp</td>
<td>1.009</td>
<td>1.011</td>
<td>1.006</td>
<td>0.941</td>
<td>0.934*</td>
<td>0.953</td>
</tr>
<tr>
<td>Private health exp</td>
<td>0.954</td>
<td>0.993</td>
<td>0.992</td>
<td>1.03</td>
<td>1.054</td>
<td>1.061</td>
</tr>
<tr>
<td><strong>Political system (ref = democracy)</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Autocracy</td>
<td>1.045</td>
<td>1.042</td>
<td>0.943</td>
<td>0.94</td>
<td>0.94</td>
<td>0.94</td>
</tr>
<tr>
<td>Anocracy</td>
<td>1.407</td>
<td>1.421*</td>
<td>1.17</td>
<td>1.17</td>
<td>1.17</td>
<td>1.17</td>
</tr>
<tr>
<td><strong>Economy</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gini</td>
<td>1</td>
<td>1</td>
<td>0.995</td>
<td>0.996</td>
<td>0.996</td>
<td>0.996</td>
</tr>
<tr>
<td>WB income (ref = high)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Middle</td>
<td>0.699</td>
<td>0.709</td>
<td>0.902</td>
<td>0.837</td>
<td>0.837</td>
<td>0.837</td>
</tr>
<tr>
<td>Low</td>
<td>0.322***</td>
<td>0.333***</td>
<td>0.339***</td>
<td>0.303***</td>
<td>0.303***</td>
<td>0.303***</td>
</tr>
<tr>
<td>External debt</td>
<td>0.958</td>
<td>0.958</td>
<td>1.009</td>
<td>1.014</td>
<td>1.014</td>
<td>1.014</td>
</tr>
<tr>
<td>GDPPPP</td>
<td>1.003</td>
<td>1</td>
<td>1.052</td>
<td>1.058</td>
<td>1.058</td>
<td>1.058</td>
</tr>
<tr>
<td><strong>Social inequality</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>GEM</td>
<td>1.243</td>
<td></td>
<td>0.461</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sample size</td>
<td>173</td>
<td>184</td>
<td>184</td>
<td>173</td>
<td>184</td>
<td>184</td>
</tr>
<tr>
<td>Maximum likelihood $R^2$</td>
<td>0.566</td>
<td>0.625</td>
<td>0.627</td>
<td>0.651</td>
<td>0.721</td>
<td>0.729</td>
</tr>
</tbody>
</table>

*Missing data in income was imputed using the ICE method

**Discussion**

The strength of the models and their ability to explain 63 to 73 % of the relationship between context and production of the two largest health workforce cadres offers striking new insight into the complex set of dynamics involving health workforce production. For some countries, a grim picture may be present in these results, but the findings do provide some clearer direction for policy initiatives. They also suggest that targeted investments aimed at meeting the sustainable development goals (SDGs) may translate both directly (e.g., quality education, gender equality, good health, and well-being) and indirectly (e.g., no poverty, zero hunger, clean water, and sanitation) into improved health workforce production. More skilled and competent health workers are essential for meeting many of the SDGs.

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*Squires et al. Human Resources for Health (2016) 14:48*
explanation of the strong correlation and significance of the “average years spent in school” of a country’s population and the health workforce ratios. Furthermore, Owen and You [41] found that the strength of institutions, like education, helps reduce gender inequality. Equitable access to education contributes to HRH production, primarily by ensuring women’s ability to obtain it. Even though the University of Copenhagen maintains a list of the world’s medical schools [42], the same data are not available for nursing nor other healthcare professions; neither WHO, the International Council of Nurses, nor comparable organizations have this information publicly available.

An additional education-sensitive, model-enhancing factor specific to nursing personnel would be accounting for the educational variation within the profession. Since nursing education around the world can range from post-secondary through the doctoral level, more precise data about degree composition within the nursing profession at the country level would strengthen the precision of the model, not to mention patient outcome analyses and workforce planning. It would be very helpful for quantifying faculty shortages which will affect production. Data of this type could also provide strategic direction for the best types of degree programs to open in a country in order to facilitate production that can efficiently meet population health needs through universal health coverage and generate optimal outcomes. This model could help provide the evidence for those types of policy changes.

Thus, for enhancing HRH data quality across the professions (and subsequently the quality of future analyses), it would be ideal to capture the total number of schools in a country, faculty and their educational preparation, categories of education (e.g., vocational/practical, bachelors, masters), entry-level education requirements, graduation rates, and employment rate post-graduation. The presence of a licensure exam or similar credentialing process would also be useful. These data would enhance not only labor market analyses but sociological ones and health outcomes analyses.

Despite the challenges of the data and considering other potential explanations for the findings, the dynamics of professional ecology may help explain the differences in the results between physicians and nurses. Medicine has created itself into a socially elite profession around the world. The positions of power and authority physicians hold in many countries ensure that elites and social class mobile individuals will study medicine there, even if they do not stay. Nursing, however, presents an entirely different story. Squires et al. [43] posit that a negative public image of nursing—resulting from a combination of low salaries, poor work environments, female gender association, and, in some places, a lack of career mobility that deters the twenty-first century applicant—creates a cyclic effect that hampers the production of nurses. Social mores around acceptable careers for women and their work schedules (i.e., night shift and its perceived safety for female workers) may further contribute to production issues for nursing personnel. Cross-national data reflecting public opinion of the nursing profession would help to determine the validity of that hypothesis and may help to further explain the ten-point gap in the results of the models between the two professions.

Other aspects of the models allude to the challenges of health system management. Regardless of location, health workers need places to work and bed availability for the population appears to be a proxy indicator of employment opportunity for both physicians and nurses. These results require further study and refinement, however, since nurses may staff primary healthcare systems at greater rates in low-income countries. Furthermore, the total number of listed beds a country reports may not all be in service due to health system financing issues. Future studies that better account for the number of beds-in-use and public primary care clinics in a country may help refine the result.

The significance of the economic and inequality variables in the model suggests that systematic national policies aimed at reducing social, gender, and economic inequality could positively affect health workforce production. It may also explain the finding about the relationship between the NMPR and migration rates since men are more likely to migrate than women [44], women are more likely to follow their migrating husbands than initiate it themselves [45], and low- and middle-income source countries for international nurse migration are dominated by a small number of countries [46]. This may prove especially important for nursing workforce production because of its female gender dominance. Previously, most studies of gender issues in the health professions, besides the aforementioned educational access ones, centered largely on pay disparities between men and women in each occupational group and largely focus on high-income countries [47–51]. Our results indicate that nurse/midwife production may be more sensitive to broader gender inequality issues than physicians. In some ways, this may seem like a “common sense” finding, but research had not previously quantified it.

Finally, when conducting the analyses, we found complexity theory a useful framework for considering variable relationships and selection. Since we aimed to avoid a “kitchen sink” approach to conducting the multivariable regressions, the theory provided direction for how to prioritize variables given our collective, interdisciplinary backgrounds and have the models reflect the
complex, interconnected systems that are health services
delivery and HRH production.

From a policy perspective, to the best of our know-
ledge, ours is among the first papers to quantify the sig-
nificance of how national education policies may impact
a country’s healthcare worker production. We have
known for many decades that education matters for
population health, and now, the significance of the rela-
tionship is quantified. Our results illustrate the import-
ance of concurrent policy efforts between education and
healthcare sectors for stable HRH production and sup-
port statements by Frenk et al. [52]. Results from this
paper can also be used to help meet international mid-
ifery production goals, as described in the landmark
Lancet series published in June of 2014 [53]. The results
suggest that policymakers that increase investments in
primary and secondary education are likely to see pay-
offs in increased health worker production.

This analysis is also among the first to highlight how
different political regimes and governance issues influ-
ence health workforce production. Our results support
findings from the social sciences that have shown
positive effects of egalitarian regimes on broadly redu-
ing inequality, both economic and gender-based [54].
Finally, the results further help to delineate the nature of
political actors who shape healthcare policy, as recom-

dended by van Olmen et al. [55], along with supporting
Varghese and Kutty’s [56] work on improving govern-
ability in public health systems. Findings may also in-
form the movement toward global health governance by
illustrating the impact of States’ political systems and
governance practices on health workforce production.

In terms of study limitations, known data quality
issues in health and development datasets may have
affected our results overall due to inconsistencies in
reporting across countries and data coordination failures
known to affect cross-national datasets [39, 40, 57]. The
choice to conceptualize the study on production versus
availability was due to the data issues and has its own
limitations. For nursing supply data in particular, how
countries define nurses and midwives can create fluid in-

terpretations of who qualifies as nursing personnel, and
calculations of nursing supply data are not the same
across countries [39, 58]. The beds-per-1000 population
data also do not account for beds not in service in the
country, which may affect employment opportunities for
both physicians and nurses and reflect the financial
health of the system. Differentiation between hospital-
based and primary care-based employment was also not
possible to obtain through publicly available sources;
those data would also add tremendously to the precision
of the analysis in future studies. Polity, Gini, and GEM
data had substantial data missing and so those variables
required more imputation than others in the study,
which may have affected the results. Nonetheless, coeffi-
cient estimates for the main factors associated with
nurses and physicians (e.g., education, migration, and
beds-per-1000 population) are of similar magnitude and
direction before and after the imputation, suggesting
that our substantive results are not driven by the imput-
ation method.

Conclusions
In conclusion, this study offers a foundation for studying
how macro-level contextual factors influence physician and
nurse workforce supply. Additional analyses that use this
model and add the health worker ratios as independent
variables with patient outcomes as the dependent variables will
also make for useful contributions to the literature. Future
research should also include replicating the study with
pharmacy and dental personnel which would be a logical
next step to check the stability of the model across profes-
sions. When more data becomes publicly available, replica-
tion of these cross-country comparative analyses will allow
researchers to further refine the methodologies and more
precisely account for the impact of potentially confounding
factors related to production. Since HRH data quality is im-
proving because its value for policymakers has increased in
the last decade, panel data analyses will also become feas-
ible. Event analyses, for example, that can account for major
economic and political changes will also allow researchers
to study the impact of those events on HRH production
over time. Future studies may also want to test other vari-
ables in the categories we identified to see if these enhance
the model’s precision. Hence, for strategic workforce plan-
ing initiatives and global health governance, this study
makes a useful contribution to the policy dialogue.

Additional file

Additional file 1: Supplementary results related to multicollinearity and
data imputation. (DOC 25 kb)

Abbreviations
GEM, Gender Empowerment Measure; HRH, human resources for health; ICE,
imputation using chained equations; NMPR, nurse/midwife-to-population
ratio; PPR, physician-to-population ratio; UNDP, United Nations Development
Program; WHO, World Health Organization

Acknowledgements
The authors would like to thank Karina Schless, BSN, RN, and Lismarys Arjona,
BSN, RN, for their countless hours pulling and cleaning data to help create
the dataset for this study.

Funding
This study was funded by a New York University Global Health Challenge
grant and the Pauline Goddard Junior Faculty Fellowship. Funds were used
to pay for research assistant services.

Authors’ contributions
AS conceived the idea for the study, obtained funding for the project,
participated in data analysis, wrote initial paper drafts, and finalized the
manuscript for submission. JU contributed to the study design, conducted
analyses in STATA, and participated in writing the paper. HBS contributed to the study design, conducted analyses in STATA including supplemental regression models, and participated in writing the paper. SJ addressed statistical methods issues in the study, completed the analyses in R, and contributed to writing the paper. All authors read and approved the final manuscript.

**Authors’ information**

AS received her PhD at Yale University in Health Policy, completed a post-doctoral fellowship at the University of Pennsylvania in Health Outcomes Research, and focuses her research on health workforce capacity building strategies and analyses. To date, she has research and programmatic experience in 30 countries, most of them low and middle income. She consults for the World Bank on health labor market analyses with a special emphasis on nursing personnel and is a Fellow with the Migration Policy Institute in Washington, DC, where she works on policy analyses related to international health worker migration. She is currently an associate professor at New York University, College of Nursing and a research assistant professor at the School of Medicine. SJ completed her doctorate in Public Health (epidemiology focus) and post-doctoral fellowship (Comparative Effectiveness Research) at New York University. Her research interests come from a decade of work related to HIV prevention in Sub-Saharan Africa and current projects reflect that work. HBS is finishing his third post-doctoral fellowship at the Center for Demography of Health and Aging at the University of Wisconsin after completing a PhD in demography at the University of Pennsylvania. His two other post-doctoral fellowships were at the University of Southern California and at the Harvard Population Studies Center. His research interests include examining aging and health demographic trends in low- and middle-income countries and developing models to predict health service demand. A native of Mexico, he will assume an assistant professorship at the University of California Los Angeles in the fall of 2015.

**Competing interests**

The authors declare that they have no competing interests.

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**Received:** 23 February 2015 **Accepted:** 28 July 2016

**Published online:** 15 August 2016

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