Home Sweet Home?

Macroeconomic Conditions in Home Countries and the Well-Being of Migrants*

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Abstract

This paper examines whether the subjective well-being of migrants is responsive to fluctuations in macroeconomic conditions in their country of origin. Using the German Socio-Economic Panel for the years 1984 to 2009 and macroeconomic variables for 24 countries of origin, we exploit country-year variation for identification of the effect and panel data to control for migrants' observed and unobserved characteristics. We find strong evidence that migrants' well-being responds negatively to an increase in the GDP of their home country. That is, migrants seem to regard home countries as natural comparators, which grounds the idea of relative deprivation underlying the decision to migrate. The effect declines with years-since-migration and with the degree of assimilation in Germany.

Key Words: Migrants, well-being, GDP, unemployment, relative concerns/deprivation.

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1 Introduction

The behavior of migrants regarding labor market decisions, the timing of return to the home country or the incentives behind "circular" migration are probably better understood if one look to both the process of assimilation and to its natural counterpart, i.e. the process of "disintegration" from their home countries (Nekoei, 2013). The latter, which describes how migrants' home country ties weaken over time, is less studied in the economic literature. Migrants may keep non-economic links with their home land (culture, altruism, patriotic feelings during soccer games) but may also experience adverse or competing feelings if the home country is taken as a natural comparator regarding economic performances.

We suggest investigating this particular dimension using subjective well-being (SWB) data. Self-reported measures of life satisfaction have been increasingly used as proxies for utility during the last decade (see the review Clark et al., 2008). This literature has established the importance of relative or positional concerns, notably the influence of a person's relative income compared to a reference group on her welfare (see Easterlin, 1995, McBride, 2001, Senik 2004, Ferrer-i-Carbonell, 2005, Luttmer, 2005, Clark and Senik, 2010, among others). Admittedly, it is difficult to identify the relevant reference point for a given population. However, migrants offer an interesting case study. They are indeed possibly confronted with multiple and switching reference groups between home countries and regions of destination. This question is related to the migration decision itself, and to the close concept of "relative deprivation" often cited in the migration literature (e.g., Stark and Taylor, 1991). Indeed, migration is often undertaken to improve a person's income relative to members of her reference group in the source country. To our knowledge, the literature has not yet studied relative deprivation (and the net gains from migrating) using SWB measures, or whether home countries are relevant reference points for international migrants.¹

In this paper, we test whether migrants are sensitive to the economic performances of both their home country and destination locations using the German Socio-Economic Panel (GSOEP) over 26 years and for 24 origin countries. Time and home country variation is used to identify the effects of macroeconomic fluctuations on migrants' well-being.² While the approach suggested in

¹An exception is Gelatt (2013) who uses data on Latino and Asian Americans to test the location of immigrants' reference groups and the relationship between various measures of subjective social standing and SWB. Akay et al. (2012) also study the role of positional concerns of migrants *within* a country (China).

²Another recent study (Nekoei, 2013) exploits 16 years × 73 origin-countries to study the effect of exchange rate volatility on migrants' labor supply in the US. Other studies check how movements in GDP, unemployment or inflation directly affect individual happiness, e.g. Clark and Oswald (1994), Di Tella et al. (2001, 2003) and Wolfers (2003). We relate especially to the Di Tella et al. papers, and to Becchetti et al. (2013), who study the correlation between citizens' (not solely migrants') SWB and their country's macroeconomic fluctuations. DiTella et al. (2003) use 17 years × 13 countries to capture enough regional and time variation in macroeconomic conditions. They report that GDP (unemployment and inflation) is positively (negatively) associated with citizens' well-being and

this paper could be replicated for other countries, we believe that Germany is interesting for at least two reasons. First, it has one of the highest immigrant populations in Western countries, with 7.72 million persons (9.5% of the total population) coming from 194 countries.³ Second, the GSOEP is a large representative dataset including SWB measures, very detailed individual and household information, a panel dimension and excellent representativeness of migrants. Our main application consists in estimating migrants' SWB on a large set of individual determinants of well-being (household income, health status, etc.) and the macroeconomic variables of home countries. We also control for migrants' family circumstances in both the host and home countries, for (overall and country-specific) time trends and, using panel information, for migrants' time-invariant unobservables.

We originally show that home countries indeed act as a natural comparator for migrants. We find a marked and statistically robust effect of the home countries' macroeconomic conditions on migrants' well-being. It is fully in line with the relative concerns/deprivation hypothesis: migrants' well-being decreases with home country GDP per capita. We extensively check the robustness of our results as well as the validity of alternative interpretations (in particular the role of remittances and a correction for possible non-random selection into return migration). We also examine heterogeneous effects of GDP on migrants' well-being, along dimensions like years-sincemigration (YSM hereafter) and objective and subjective measures of the degree of assimilation in Germany. We unveil that competing feelings towards home countries decrease after some years in the host country. Consistently, less assimilated migrants keep strong translational ties, and origin countries are likely to remain their key reference group. Our conclusions are reinforced through finding an effect of opposite direction regarding local economic performances, i.e., migrants' wellbeing increases along with the GDP of the German counties in which they live. Interestingly, this 'signal effect' also declines with YSM, as if gradually replaced by relative concerns towards the local environment. These results are consistent with the existence of multiple reference points and a possible switch over time and with the assimilation process. We derive important labor market and migration policy implications from these results.

The paper is organized as follows. Section 2 presents the data and the empirical methodology. Section 3 reports the main results, robustness checks and additional results using migrants' heterogeneity. Section 4 concludes.

explain this correlation with feelings of national prestige (for GDP), corroding purchasing power (for inflation) and loss of self-esteem, depression, anxiety and social stigma (for unemployment). Becchetti et al. (2013) show that neighboring countries can be reference groups and generate negative feelings if they experience higher economic success.

³Figures extrapolated to the recent years (before the refugees' crisis) on the basis of the 2011 microcensus by the Federal Statistical Office (www.destatis.de).

2 Empirical Approach

2.1 Data and Selection

Our analysis is based on the German Socio-Economic Panel (GSOEP), a well-known survey of individuals in households living in Germany. It has been used in important analyses in the SWB literature (see, e.g., van Praag et al., 2003; Frijters et al., 2004a, 2004b; Ferrer-i-Carbonell, 2005). It is a representative survey of the entire German population and an exceptionally long panel, of which we are using 26 years from 1984 to 2009. It contains a wealth of information at the individual or household level, including data on education, health, labor market conditions and incomes, as well as various subjective measures of well-being. The dataset was started in 1984 in West Germany (with around 10,000 respondents per wave) and has covered the entire reunited Germany since 1990 (with around 14,000 respondents per wave after 1990 and more than 20,000 after 2000).⁴

We select all the waves of the GSOEP, keeping all adult first-generation immigrants aged 17 or older and living in West or East Germany.⁵ Although more than a hundred nationalities are reported, we restrict our study to the main migration groups, resulting in 24 different countries of origin. These correspond to the largest groups in terms of their population size in Germany and countries for which we have at least 100 observations in the data. Our dependent variable (subjective well-being of individual *i* of country *h* at year *t*, SWB_{iht}) derives from the question "How satisfied are you with your life as a whole, all things considered"? The answer is reported on an 11-point scale (0 signifies "completely dissatisfied" and 10 means "completely satisfied"). Life satisfaction is highly correlated with other subjective measures of well-being like self-reported happiness or aggregated answers about mental health such as the GHQ-12 (see Clark and Oswald, 1994). It has been shown that SWB information is a solid proxy for individual well-being, notably because of the strong correlation with other, more objective measures of well-being (see Oswald and Wu, 2010).⁶ We combine SWB and other individual characteristics with macroeconomic variables

⁴Sample weights are provided and used to guarantee the representativeness of the sample. Representativeness of the migrant population is excellent according to the detailed assessment of Lelkes and Zolyom (2010). Attrition in GSOEP is discussed in Spiess and Kroh (2004) and, in relation with SWB estimations, in Frijters et al. (2004b). Non-random attrition due to return migration is addressed in our analysis below.

⁵We select first generation migrants using information from the "migration background" module of the GSOEP. The migration status of an individual is obtained by combining information on his/her country of birth, citizenship, migration history and parental information. We also have an exact information about the arrival year in Germany, which is used to define the year-since-migration variable and arrival-cohort dummies.

⁶In addition, Krueger and Schkade (2008) provide extensive evidence about the robustness of SWB measures compared to more usual data used by economists. Di Tella et al. (2003) report a high regularity in SWB equation regressions across different nations (as we do for the different migrant groups in our data) while Clark et al. (2008) show that changes in SWB are good predictors of behavior responses. All these checks convey that SWB is not

for the migrants' 24 countries, drawn from annual time series data of the World Bank indicators. We focus on the main variables of interest, including log real GDP per capita of country h in year t (denoted GDP_{ht} hereafter),⁷ growth in real GDP per capita (denoted ΔGDP), price levels measured by the GDP deflator (P_{ht}) and unemployment rates U_{ht} .⁸ The resulting sample includes a total of 51,171 individual×year observations obtained over 26 years of data and migrants from 24 origin countries. We lose a small fraction of this dataset due to missing information so our final sample contains 47,557 individual×year observations.

2.2 A First Look at the Data

Table A.1 in the Appendix provides some aggregate statistics by country of origin, including the main macroeconomic indices (log real GDP per capita expressed in PPP-adjusted 2005 international dollars, nominal GDP per capita and unemployment rates) and migrants' SWB (average SWB over all migrants of a country for the period 1984-2009). We also provide the ratio of real GDP per capita for each country compared to Germany. This reflects the huge variation in development levels across immigration countries, and the convergence process (18 countries out of 24 have caught up with Germany over the period). A lot of variation can also be observed concerning reported well-being. On the 0-10 scale, migrants' SWB scores 7.1 on average over all years and origin countries. Using the country average over 1984-2009, we see that SWB varies from 5.8 for Iranian migrants to 7.6 for Dutch migrants, which partly reflects the large variation in living conditions (as proxied by GDP_{ht}) across nations. This is illustrated by the cross-country correlation between average migrant SWB and absolute real GDP (respectively unemployment rate), which amounts to .46 (-.40). The second country approach is a superior of the second country and the country of the country average over 1984-2009, we see that SWB unemployment correlation between average migrant SWB and absolute real GDP (respectively unemployment rate), which amounts to .46 (-.40).

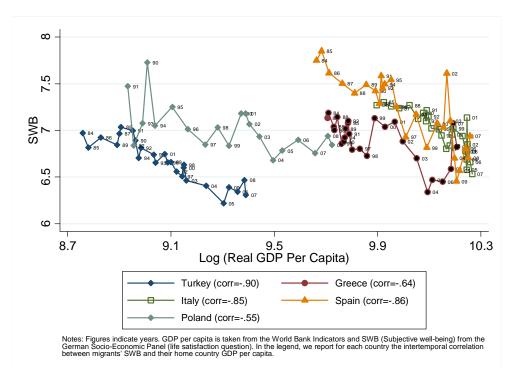
mere statistical noise but rather contains meaningful information. Nonetheless, we keep in mind the possible lack of interpersonal comparability in the perception of (and answers about) well-being. We treat this as a measurement error, namely by using large samples and by controlling for individual fixed effects in our regressions. Notice that we are not interested in SWB scales per se but in the effect of home country macroeconomic performances, or in their relative effect. The latter, the trade-off between these performances and individual income, can be calculated as an "equivalent income" measure of relative concerns, as explained below.

⁷In all the estimations hereafter, we use the log of real GDP per capita divided by 10,000, for comparability with Di Tella et al. (2003).

⁸See http://data.worldbank.org/indicator.

⁹For instance, the ratio is as little as 30% (respectively 28%) of the German real (nominal) GDP per capita for Iran and up to 113% (99%) for the Netherlands.

¹⁰However, differences in income levels do not perfectly explain the well-being gap. The relationship between income and well-being may not be linear: beyond a certain income level, income differences have smaller effects on perceived well-being (this pattern is found in Easterlin, 1995, but questioned more recently by Stevenson and Wolfers, 2008, who do not reject linearity). For instance, the correlation between mean SWB and real GDP per capita is smaller when GDP is expressed in logs (.36). Moreover, if we focus on Western European countries and the US, this correlation drops to .07.



SWB versus GDP Across Time for Selected Ethnic Groups

In Table A.1, we also report correlation over time between yearly migrants' average SWB_{ht} and home country GDP_{ht} (or unemployment U_{ht}). Interestingly, for GDP (respectively unemployment), the time correlations are negative (positive) in the majority of countries, as if increases in GDP per capita (unemployment) were associated with a decline (rise) in the well-being of the corresponding migrants. This unexpected result is illustrated in Figure 2.2 for the five largest migrant groups (those from Turkey, Greece, Italy, Spain and Poland). We plot log real GDP per capita, GDP_{ht} , against yearly migrants' average SWB, SWB_{ht} , for all our panel years (years are indicated next to the data points). The negative relationship between home country GDP and migrants' SWB seems to characterize the whole period (with a few exceptions) and most immigration countries.¹¹ We do the same for unemployment rates (Figure A.1 in the Appendix): the pattern is not as pronounced as for GDP, yet it seems as though increases in unemployment rates are associated with an increase in SWB.

These preliminary results directly align with the interpretation in terms of relative concerns or relative deprivation suggested in this paper. With the Easterlin paradox (Easterlin, 1995), the fact that a country like Germany has experienced GDP growth but a flat trend in SWB over the past 30 years often pertains to the classic explanation in terms of "positionality". That is, after

¹¹This result is not only driven by the periods of economic growth. While not visible in Figure 2.2, we observe in source data that, for instance, the downturns of 1993-1994 and 2000-2001 in Turkey and the 2008-2009 recession in Italy are associated with an increase in SWB among migrants from these countries.

some point, well-being would depend more on relative income than on absolute income, so that absolute increases in national wealth would not improve well-being over time. First, this argument does not mean that relative concerns kick in only above a certain level of income, just that they tend to overcome absolute income effects at this point. Status indeed plays a considerable role in the context of poorer countries as well (see Clark and Senik, 2015 ed., for recent evidence). Second, for migrants (from poorer or other rich countries in the case of Germany), one could in fact expect an even more radically opposed association between SWB_{ht} and GDP_{ht} . Indeed, if home countries act as reference points and if most countries "catch up" with Germany, the relative position of migrants declines over time and their SWB can be negatively affected. This is exactly what Figure 2.2 illustrates. In the following, we attempt to better characterize this effect by means of regressions while controlling for migrants' characteristics. We shall demonstrate that these co-movements in SWB and home GDP are causally linked by the fact that origin countries serve as a reference point against which migrants assess their own well-being.

2.3 Modeling the Well-being of Migrants

Using our selected panel of migrants living in Germany, we estimate the well-being SWB^* of migrant i from home country h at time t as follows:

$$SWB_{iht}^* = X_{it}\alpha + \gamma Macro_{ht} + \theta_t + (t \times \delta_h) + \varphi_i + \varepsilon_{iht}. \tag{1}$$

Latent well-being SWB^* is considered as a proxy for the unobserved utility of a migrant, for which we observe an ordinal metric $SWB_{iht} = j$ on an ordered scale of well-being categories j = 1, ...J. The model combines both characteristics of migrant i at year t, X_{it} , and macroeconomic variables of her home country h at year t, $Macro_{ht}$. Individual time-varying variables in X_{it} include the usual determinants of SWB, i.e., log household income, work status, marital status and family circumstances, health status, education, other characteristics related to the home country (children and spouse in the home country, refugee status, remittance receipt), and German states $(L\ddot{a}nder)^{12}$. We also control for year dummies θ_t (they pick up the effect of German GDP as well as of any global shocks that are common to all migrants' countries in each year), country-specific linear time trends $t \times \delta_h$ (δ_h denotes country fixed effects), individual effects φ_i and a usual i.i.d error term ε_{iht} . Country time trends may capture, for instance, cultural attitude toward changes in well-being or country-specific unobservable assimilation patterns of migrants of country h. 13

¹²State effects account for possible migration patterns within Germany. Evidence in GSOEP shows, however, that geographical mobility of migrants is extremely limited (see Akay et al., 2013).

¹³Together with flexible time trends θ_t , they also represent deterministic functions of time that are used to render the data stationary (Di Tella et al., 2003, stress that for usual unit-root reasons, untrended SWB should not be regressed on trended macroeconomic indices like GDP).

Our baseline estimation strategy consists of linear panel estimations with fixed effects (FE), denoted by φ_i . Alternatively, we shall experiment with the Mundlak "quasi-fixed effects" (QFE) model, which combines both between and within variation. This model allows for the inclusion of variables which cannot be introduced in FE estimations, notably country effects and immigrant arrival cohorts. Hence, the overall individual effect is based on a slightly more structural specification where $\varphi_i = \delta_h + Z_i + Age_{it} + YSM_{it} + u_i$, with home country effects δ_h (for unchanging cultural influences of origin country on reported well-being), time-invariant characteristics Z_i (gender and cohort effects), two time variables (age and YSM, which are not identified when using FE time-demeaning panel estimation with year effects), and the Mundlak QFE u_i . Finally, we consider that J = 10 is large enough to treat reported well-being as a continuous variable so that (1) can be estimated linearly. Yet, we also provide checks where we acknowledge the ordinal nature of the dependent variable, allowing for unobserved individual effects in this nonlinear context by using the QFE ordered probit and the "Blow-Up and Cluster" FE ordered logit estimators (see Baetschmann et al., 2015).

3 Results

3.1 Main Results

We first present our main results, namely the estimation of model (1) on panel data. It relates the macroeconomic conditions of home countries to individual SWB, conditional on various individual and family circumstances in both the host and home countries.

Effect of GDP: Baseline Estimations. In this section, we shall present summary tables in which we report estimates of the coefficient γ only. Our main result is in the two first columns of Table 1.¹⁷ We report panel estimations of the effect of log real GDP per capita on SWB while

¹⁴Migrants may vary in unobservable characteristics depending on the year they arrived in Germany (Borjas, 1999). Therefore, migrants are grouped into 9 cohorts taken 5 years apart (9 dummy variables starting from pre-1960 arrivals until the last cohort corresponding to the last 10 years). These cohort dummies aim to capture cohort-specific unobserved characteristics affecting migrants' well-being. Grouping is necessary for identification.

¹⁵Following Mundlak's approach, the latter combines a normally distributed term and within-means of relevant time-variant variables (we use household income, household size, age, amount of remittances sent to the home country, education and working hours).

¹⁶The advantage of the linear approach is that it makes the required extensions to panel estimations much more transparent and allows including unobserved individual heterogeneity in a flexible way. Notwithstanding, Ferreri-Carbonell and Frijters (2004) show that results are typically similar using both linear and ordinal models; the present study shares this conclusion.

¹⁷The complete set of SWB estimates is shown and discussed in the Appendix (Table A.2). Models I and II relate to models 1 and 2 in Table 1, while model 0 is a variant without home country GDP (see detailed discussion

controlling for year effects, state effects and time-invariant unobservables (FE). We obtain an estimate of -.303, which is significant at the 1% level. The next column additionally controls for country-specific time trends to filter out the spurious correlation between macroeconomic indices and SWB. The magnitude of the effect is basically unchanged (-.212) but the effect is less precisely estimated, even if still significant at the 10% level. This finding suggests that macroeconomic movements in the home countries feed through into migrants' feelings of well-being. This may be seen as an unexpected result if one believes that migrants are bounded to homelands by a sense of pride, identity and patriotic ties. Yet it is likely that this altruistic or emotional link pertains to non-economic aspects.²⁰ As far as economic conditions are concerned, our results do consolidate previous findings in the literature showing that people's well-being is evaluated against natural comparison points (e.g. Luttmer, 2005) – and we show that home countries are an important one. This also relates to the fact that mean income in home countries is a marker with respect to which migrants can gauge the success of their migration experience. Migrants from countries characterized by better macroeconomic performances experience lower gains from migration and, other things being equal, lower levels of well-being. Arguably, this effect may be attenuated when migrants decide to stay forever in Germany or become assimilated enough for their reference point to shift from home countries to other comparators within Germany. We investigate this point below.

Magnitude. To gauge the magnitude of the effect, we suggest alternative metrics and a brief comparison with other studies. We base our calculation on the FE model with country-specific time trends. First, a one standard deviation increase in the home country's (log) GDP per capita is associated with a decline of 2% of a standard deviation of SWB (or a 0.5% decrease in mean SWB). While this may seem modest, it is very much in line with measures of relative concerns or socio-economic status in the literature. For instance, Di Tella et al. (2010) find that a one standard deviation change in status (i.e., an individual's relative standing to others measured

in the Appendix).

¹⁸In all specifications, clustering is made at the year and home country level to account for possible bias due to repeated observations for the same country of origin (and to control for the correlation between errors in the same country). Alternatively, we have clustered standard errors at the individual level due to the panel nature of the data. The standard errors only slightly increased in both cases.

¹⁹Alternatively, we have also used the Hodrick-Prescott filter to detrend macroeconomic variables before estimations (detailed results available from the authors). By doing so, we obtain an effect of −.303 (standard deviation of .145) for detrended GDP per capita in levels and −.256 (.137) for GDP per capita in logs. Hence, the results are still significant in this case and the log GDP effect is of similar magnitude as in the baseline.

 $^{^{20}}$ We perform separate estimations of the effect of battle-related deaths (log number of people) and life expectancy (number of years) on migrants' SWB, using the same controls as in the baseline model. The former is significant (estimates of -0.016 with a standard error of .006), suggesting that there may be feelings of sympathy towards home countries when it comes to non-economic domains.

by job prestige) explains 3.1% of the standard deviation in SWB. An alternative way to gauge the effect is to take the ratio of the coefficient on log GDP per capita over the coefficient on log household income.²¹ We obtain a ratio of -.553, which can be interpreted as an *equivalent income* variation, i.e. a 1% increase in the home country's real GDP per capita is equivalent to a .55% decrease in household income. Drawing from estimates of absolute and relative income effects in the literature, we find equivalent income measures of a similar order of magnitude, for instance -.58, -.76 and -.82 in Akay and Martinsson (2011), Ferrer-i-Carbonell (2005) and Luttmer (2005) respectively.

Alternative Estimators and Specifications. Our baseline results are obtained with FE linear estimations and treating SWB as a continuous variable. We check the sensitivity of our results with respect to alternative estimators. Table A.4 in the Appendix reports a series of estimates, starting with the FE model without and with country-specific time trends. Acknowledging the ordinal nature of SWB data, we also show estimates of the "Blow-Up and Cluster" FE ordered logit. The coefficient is still negative and significant. We could not calculate marginal effects but we can check the equivalent income measure, -.972, which turns out to be only slightly larger than the linear FE estimation without country time effects. Then we move to QFE estimates showing very similar results compared to the baseline (-.281 and -.224 for QFE models without)and with country time effects, respectively). Equivalent incomes are also almost identical. The penultimate model augments QFE with information on personality traits based on the so-called "Big Five" model. Psychological traits are increasingly used as a time-invariant and potentially important determinant of well-being (Boyce, 2010). "Big Five" traits are reported in waves 2004 and 2009 only, so cannot be used for all individuals in the panel. Despite the resulting drop in sample size, the coefficient of -.321 is close to the baseline. Finally, we estimate an ordered probit with QFE: the coefficient of -.215 is not directly comparable but the equivalent income is again very similar to the baseline.

Timing and Adaptation. Turning back to Table 1, we provide additional results, starting with the timing of the effect. It may be the case that migrants are affected by the *dynamics* of their country's economic performances more than its actual level. We introduce the potential role of ΔGDP alone or together with GDP_{ht} (columns 3 and 4 of Table 1). The negative sign on the former term indicates that an increase in home country growth negatively affects the well-being of migrants, yet it is not significant. If introduced simultaneously, the GDP effect remains significant and close to the baseline. A more flexible way to account for dynamics is to introduce lagged GDP. Macroeconomic fluctuations may be perceived with a delay or their impact on SWB could

²¹The latter is 0.38 in the baseline, which is of the same order as in related studies. For instance, Akay and Martinsson (2011), Ferrer-i-Carbonell (2005) and Di Tella et al. (2010) report .36, .25, and .20 respectively.

depend on longer-term trends rather than on current economic conditions. Lagged macroeconomic variables can also relate to adaptation effects (Di Tella et al. 2003, 2010), stemming from the idea that migrants may adjust to the home country GDP after a period of time and only thereafter derive negative positional feelings from increases in GDP. Columns 5 and 6 show results with 1-year and 2-year lags of GDP respectively. Di Tella et al. (2010) interpret the sum of lagged effects as the amount of adaptation. We observe that lagged GDP effects change sign. Only the 2-year lag is significant but an F-test of whether the joint effect of all GDP variables (i.e., current and lagged) is zero can be rejected. With one lag (two lags), 17% (9%) of an initial increase of GDP is lost over the ensuing year(s), leaving a long-lasting effect of -.337 (-.426) on SWB, which is very similar to our baseline result. We draw two lessons from these results. First, it is obviously not possible to identify the precise timing due to the high correlation between GDP_{ht} , GDP_{ht-1} and GDP_{ht-2} . This is no impediment to our analysis, as cumulated effects do not change our conclusions. Second, we find no evidence of an adaptation effect to individual positional concerns towards the home country.²²

Price Effects and Exchange Rates. In place of real GDP, it would make sense to include log nominal GDP per capita, denoted GDP_{ht}^{nom} , to check if migrants are to some extent victims of money illusion (Boes et al., 2007). That is, migrants should be affected by the success of their home country in terms of nominal GDP, but they should also know that a price increase in their home country reduces their relative deprivation as it decreases the relative cost of living in Germany. Since $GDP_{ht}^{nom} = P_{ht} + GDP_{ht}$, with P_{ht} as the log price index (log GDP deflator), we can simply introduce the latter in the SWB regression together with GDP_{ht} . Column 7 in Table 1 shows that the effect of log real GDP per capita is unchanged while the log price level has no significant effect. Even if not a definitive proof, this is suggestive evidence that real GDP is what truly matters for well-being, i.e. migrants do not suffer from money illusion. Regardless, other interpretations should be mentioned. In particular, migrants from countries with lower relative prices could take advantage of the higher relative purchasing power of their income when they vacation at home. In this case, higher prices in the home country should decrease rather than increase SWB, an effect that may partly counteract the relative concern effect described above. For one thing, migrants could equally go to any other low-price country to take advantage of their German salaries. We nonetheless replicate estimations while including exchange rates between the home country and Germany. In column 8, the GDP effect is slightly larger than the baseline (but not significantly so) while the coefficient on exchange rates is insignificant.

²² If any, this is a very partial adaptation process, which is consistent with the findings in Di Tella et al. (2010) or Ferrer-i-Carbonell and van Praag (2008). These authors show that while people almost fully adapt to changes in absolute living standards, they do not (or only partly) adapt to changes in status.

Table 1: Effect of Home Country GDP on Migrant SWB: Micro Data

Dependent variable: SWB	1	2	3	4	5	6	7	8
GDP	-0.303 ***	-0.212 *		-0.338 ***	-0.406 *	-0.468 **	-0.349 ***	-0.437 ***
	(0.107)	(0.125)		(0.120)	(0.221)	(0.233)	(0.120)	(0.142)
ΔGDP			-0.110	-0.095				
			(0.169)	(0.170)				
GDP (t-1)					0.069	0.378		
					(0.193)	(0.286)		
GDP (t-2)						-0.336 *		
						(0.185)		
Prices (GDP deflator)							-0.012	
							(0.009)	
Exchange Rates								0.004
								(0.005)
Individual effects (a)	FE	FE	FE	FE	FE	FE	FE	FE
State effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Home country linear time trends	No	Yes	Yes	Yes	Yes	Yes	Yes	Yes
GDP (equivalent income)	-0.797	-0.553		-0.889	-1.069	-1.237	-0.918	-1.110
GDP (equiv. inc. cumul.)					-0.887	-1.126		
R2 or pseudo-R2	0.140	0.141	0.139	0.139	0.139	0.138	0.139	0.135
# observations	47,557	47,557	47,557	47,557	47,557	47,557	47,557	45,974

Note: *, ***, **** indicate significance levels at 10%, 5% and 1% respectively. Linear estimations performed on migrants from 24 countries over 26 years. GDP refers to log of real GDP per capita, taken from World Bank indicators. Subjective well-being (SWB) taken from the German Socio-Economic Panel. All models include the time-varying characteristics reported in Appendix Table A.2 as well as fixed effects (FE), state effects (16 federal states of Germany) and year effects.

Unemployment. The effect of home country unemployment rates on migrants' SWB is reported in Table A.5 in the Appendix, using alternative specifications including simultaneous estimation of unemployment and GDP effects. The overall picture is that results are much less pronounced in the case of unemployment. Consistently with our positionality interpretation, the coefficient on unemployment is positive. Yet it is small enough, or the effect imprecisely estimated, so that it becomes insignificant as soon as individual effects (FE or QFE) are introduced.²³ This could be explained by the fact that the unemployment effect also relates to migrants' own labor market prospects in the case of return migration. Another explanation is that informal work, which affects many of the low-income countries sending migrants to Germany, might leave unemployment as a less reliable proxy for their labor market conditions.

3.2 Sensitivity Analysis

Basic checks. First, it may come to mind that such a positional concern vis-à-vis home countries can be mitigated by the fact that some of the migrants' close relatives still live there and may be negatively affected by macroeconomic shocks. For this reason, our estimations control both for the presence of close relatives in the home country and for the level of remittances sent by migrants to help face income shocks (see Appendix Table A.2). Second, the effect could be driven by the fact that household income is partly determined by home country GDP if a migrant has investments in the home country. In the absence of information in GSOEP about the specific nature of investments, we can nonetheless replicate baseline estimations whereby investment income is excluded from household income. In this case, the coefficient on GDP could now capture both the investment income effect (positive) and relative concerns (negative). Results show hardly any difference with the baseline estimates (i.e. coefficient of -.307, std. err. of .112), which conveys that the former effect is certainly marginal. Third, the effect of home country GDP implies that origin countries are different from other countries in terms of migrant comparisons. To check this, we conduct a placebo test whereby each migrant is randomly assigned to another country's GDP. In this case, the estimated coefficient on the placebo GDP figure is insignificant.

Regions of Origin. Next, we investigate the sensitivity of our results to country and year selection. First, Turkish migrants are by far the largest group among all migrants in Germany (25.1% of the total foreign population, see Table A.1). We check if this group drives the results. In columns (a) and (b) of Table 2, we report estimates of the FE model on our sample excluding Turkish migrants and on Turkish migrants alone. The effect of GDP_{ht} is negative and significant in both cases. It is very similar to the baseline in the model without Turkey, conveying that results are not driven by Turkish migrants alone. The coefficient is very large (but less precisely

²³This is true in general and for separate estimations on working age, employed and unemployed migrants.

estimated) when using only time variation among Turkish migrants.²⁴

Next, we check whether the effect varies with geographical distance to Germany. Closer countries are in general richer (so the rate by which they may converge towards German GDP is lower), yet GDP comparisons can be easier to do. Countries located farther away are poorer but make circular migration more difficult (especially in the early years of our panels during which possibilities of air travel were not as developed as today). Columns (c) and (d) in Table 2 show estimates using a threshold of 2 100 kilometers from Germany (the median), which excludes countries like Turkey, Iran, Ukraine and Russia. The effect is larger in the more distant group, but not significantly so, compared to countries in the vicinity of Germany. Finally, we distinguish countries of origin by level of economic development: OECD/rich countries (real GDP above 65% of German real GDP), middle income (35-65%) and poor countries (below 35%). Estimates in columns (e) to (g) display a U-shaped pattern, i.e. stronger effects from less developed countries, an insignificant effect in the middle group, and the largest effect from rich countries. The latter may correspond to the fact that the economic performances of neighboring countries are most visible (see also Becchetti et al., 2013) and generate the most regret among migrants who do not benefit from the positive dynamics at home.

Asymmetrical Effects. We also verify if selected years make a difference. As previously seen in Figure 2.2, most countries in our sample experience economic growth for a majority of the years 1984-2009. We investigate whether our results are driven by these episodes of growth or whether the recession years tell us a similar story. While upturns in home countries are expected to trigger relative concerns among migrants, downturns may have an asymmetrical effect if migrants experience more sympathy toward their nation during bad years. We interact macroeconomic conditions with dummy variables for upward or downward changes in these variables. The results are reported in columns (h) and (i) of Table 2. Both upward and downward changes in the home country GDP affect migrants' well-being. While the effect generated by economic downturns in home countries is smaller, as conjectured above, the difference with upturns is neither large nor significant.

3.3 Alternative Interpretations

Di Tella et al. (2003) discuss the possible endogeneity of national GDP effects on citizens' life satisfaction. They reckon that it is difficult to find believable macroeconomic instruments and therefore suggest instead to experiment with different forms of lag structures. In the present

²⁴This is not necessarily reflecting a larger effect in this country. Indeed, in this case, the estimation is different as coefficient γ accounts for GDP time variation only, in a specification where time dummies θ_t must be dropped (as they would pick up Turkish GDP over time) and a linear time t trend is kept.

Table 2: Effect of Home Country Macroeconomics on Migrant SWB: Sensitivity Check

Dep. variable: SWB	(a)	(b)	(c)	(d)	(e)	(f)	(g)	(h)	(i)	
	All countries	All countries		Distance to Germany \$		Level of Home Country GDP \$\$			Checking for asymmetrical effects:	
	but Turkey On.	Only Turkey	<median< td=""><td>>m edian</td><td>rich countries</td><td>middle income</td><td>poor countries</td><td>upward trends</td><td>downward trends</td></median<>	>m edian	rich countries	middle income	poor countries	upward trends	downward trends	
GDP	-0.313 ***	-0.745 ***	-0.258 **	-0.428 **	-0.718 ***	-0.126	-0.318 ***	-0.297 ***	-0.255 **	
	(0.114)	(0.160)	(0.115)	(0.170)	(0.274)	(0.214)	(0.110)	(0.109)	(0.116)	
GDP (equiv. Income)	-0.889	-2.209	-0.680	-1.127	-1.927	-0.338	-0.854	-0.783	-0.672	
R2	0.133	0.152	0.140		0.140			0.140		
# obs.	31,303	16,254	47,557		47557			47,557		

Note: *, ***, **** indicate significance levels at 10%, 5% and 1% respectively. Linear estimations performed on migrants from 24 countries over 26 years. GDP refers to log of real GDP per capita, taken from World Bank indicators. Subjective well-being (SWB) taken from the German Socio-Economic Panel. All models include the time-varying characteristics reported in Appendix Table A.2 as well as fixed effects, state effects (16 federal states of Germany), year effects and home country linear time trends. Models (c) to (i) obtained by interaction effects. \$ Median 2100 km from German boarders. \$\$ Rich (middle income, poor) countries are defined as those with real GDP above 65% (between 35 and 65%, below 35%) of German real GDP.

context, there is much less concern for endogeneity given the minimal influence of migrants on their home country's GDP. Nonetheless, relative changes in the home country's GDP may affect migrants through three other channels besides positional concerns, namely migration flows, remittances and the option to return home. We now investigate whether migrants responding to country-of-origin conditions through these variables challenge our interpretations.

Inflow of Home Country Peers. A potential effect of bad economic conditions in the home country is that more potential immigrants from that country may be interested in migrating to Germany. Possibly they migrate to the same regions where their co-nationals already live. In this case, an increased flow of new migrants may enhance the well-being of existing migrants (which would reduce our effect) or decrease it (which would explain our effect). Additional, unreported estimations depart from our baseline model by including the proportion of immigrants in local labor markets. They show no effect of the latter, interpreted as a *change* in migrants' proportion in our FE estimations, while the effect of GDP_{ht} is basically unchanged. This is also true when including local labor market conditions (mainly the local unemployment rate). More generally, the formation of enclaves requires long lasting dynamics, probably mixing people of different nationalities. Also, migration inflows cannot respond freely to changes in the home country's economic conditions.

Return Migration. A second channel is return migration, which we treat as a more serious challenger in terms of interpretating our results. Indeed, the potential return decision concerns

each migrant directly. We first empirically check whether return migration depends on changes in the home countries' macroeconomic performances. We suggest the following model:

$$r_{iht} = \mathbf{1}(X_{it}.\eta + \mu.Macro_{ht} + \nu_i + \xi_h + \pi_t + \nu_{iht} > 0), \tag{2}$$

where r_{iht} is an indicator variable taking value 1 if migrant i from country h leaves Germany in year t (and drops from the panel for this reason), and 0 otherwise. The model combines individual characteristics, X_{it} , including cohort and state fixed effects, a macroeconomic index of the home country, $Macro_{ht}$, individual effects (modeled as QFE), ν_i , country and time fixed effects, ξ_h and π_t respectively, and an i.i.d. normally distributed random term, v_{iht} . Unreported results show that μ is positive but insignificant.²⁵ Next, we re-estimate SWB regressions accounting for possible return – and non-random sample attrition due to return migration – as a function of home country macroeconomic conditions. We use the Heckman procedure adapted to panel data by simultaneously estimating selection into return migration and the SWB equation by Maximum Likelihood (for a more structural approach, see Bellemare, 2007). A complete discussion on the instrumentation is provided in the Appendix. The first column of Appendix Table A.6 reports the effect of $GDP_{h,t}$ on migrants' SWB when controlling for selection into return migration. It is very much in line with the baseline results and significant in all cases.

Remittances. Remittances constitute a third channel linking migrants to their home countries. First, remittances sent by migrants can directly affect home country macroeconomic conditions and influence, at the same time, their own well-being. Yet, the latter effect is of significant magnitude only for a limited set of countries and years. Horeover, our GDP measure already includes total annual remittances sent by migrants in Germany and other destination countries. Second, if per capita income in the home country increases, migrants may need to compensate their relatives left behind less and, hence, their SWB would increase. Note however that our baseline estimations already control for the amount of remittances sent by migrants, and we find hardly any difference in the GDP effect whether we include this variable or not. Additionally, we have run estimations of the probability to send remittances on individual characteristics and macroeconomic variables. Remitting does not significantly depend on (current or lagged) GDP. Third, even if remitting behavior does not respond much to home country economic conditions, the implicit value of remittances may change with it. If economic conditions improve, migrants' status may decrease (along with their SWB) to the extent that their role as supporting their

 $^{^{25}}$ We obtain the same conclusion with lagged GDP. Only the lagged change in GDP, i.e., $GDP_{h,t-1} - GDP_{h,t-2}$, is found to significantly affect the probability of return in year t. Note that this variable does not affect migrants' SWB in the main equation.

²⁶This concerns especially Turkey, given the size of its migrant community in Germany. For instance in 2002, remittances sent by Turkish migrants living in Germany accounted for 0.4% of the total GDP of Turkey. We have checked above that this country does not drive the results.

extended family in the origin region becomes less prominent. In fact, replicating our estimations on migrants who do not send remittances provides results that are very similar to the baseline. These various checks convey that the channel of remittances does not affect our results nor our interpretation in terms of relative concerns/deprivation.

3.4 Heterogeneity among Migrants and Additional Outputs

We now examine how the migration history of migrants and their connection to the home country may affect the results. To capture migrants' heterogeneity, we first linearly interact GDP with migrants' duration of stay (YSM), then with a set of characteristics on intentions to stay in Germany, objective and subjective measures of assimilation and attachment to host versus home countries.

Duration of Stay. We first check how duration into migration influences the *GDP* effect. We use a flexible specification with four groups of YSM interacted with the GDP coefficient (less than 10 years, 10-20, 20-30 and more than 30 years). FE estimations with year effects do not allow the inclusion of time variables like age or YSM, so our interaction terms would not have a clear interpretation. Therefore, we rely on QFE in this exercise. The results correspond to the blue curve in Figure 1. The effect of the home country GDP per capita is negative and very large (around -.5) in the first 10 years, declines a bit in the following years, then becomes virtually zero after 20 years. That the effect of the home country GDP only affects migrants' SWB in the first two decades after arrival can most likely be interpreted in two ways: (i) as migrants assimilate into the host country, the effect of the home country GDP as a reference group fades away; (ii) migrants who arrived young in the host countries are more assimilated and ignore their home country as a reference point. Alternative, less convincing explanations pertain to the changing composition of the migrant community due to cohort effects²⁷ or to return migration.²⁸

Assimilation and Fading Connection with Home Countries. We further explore the assimilation process that potentially explains the pattern in Figure 1. First, interpretation (ii) above suggests that relative concerns are necessarily smaller for those who arrived as a child in

²⁷For instance, new comers due to family reunification would have different assimilation potentials than first round migrants attracted by bilateral guest-worker programs (Borjas, 1999). Yet, we control for unobservable differences between different migrant cohorts by using arrival cohort fixed effects in our QFE estimations.

²⁸Those experiencing greater relative concerns could also be more likely to eventually return to their home countries. Yet, we have seen that accounting for non-random return migration did not change our result at the mean. Moreover, the short-dashed line in Figure 1 plots the GDP effect estimated on a sub-sample excluding all the observations for those who return to home countries at some point in the panel (930 return migrants over the period of study and 6,118 individual-year observations). The results are basically unchanged.

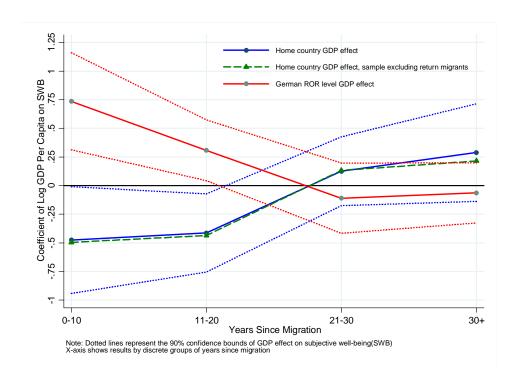


Figure 1: Effects of Home GDP versus Local German GDP on Migrants' SWB according to Years Since Migration

the host country and feel disconnected from the home countries. In unreported estimations, we have interacted GDP with dummies for the age at which migrants arrived in Germany: as a child (under 12), teenager (12-18), young adult (18-39), or older. Results are not inconsistent with this explanation. While those who arrived as children are not affected by home country GDP, the GDP effect remains significant at older ages (12-18 and 18-39). Yet we cannot provide a definite answer to the question of whether stronger assimilation for people who migrated younger is due to (i) duration of exposure or (ii) exposure during a sensitive period for acculturation.²⁹

Second, the assimilation process may have more implications than just "forgetting" home countries. It may also imply a switch in the reference group over time, with the *local* economic environment becoming the new natural comparator for long-term migrants. To check this, we exploit variation in economic performances across German ROR (*Raumordnungsregionen*). ROR are spatially organized units based on various criteria to represent local markets. We match information about 96 German ROR with our micro data and regress migrants' SWB on ROR-level GDP interacted with YSM.³⁰ The red curve in Figure 1 shows that for short-term migrants, local GDP

²⁹Some evidence, provided by Cheung et al. (2011), tends to show that these mechanisms are cumulative: people are better able to identify with a host culture the longer their exposure is to it, but only if this exposure occurs when they are relatively young.

³⁰ROR information is unfortunately limited to 12 years, 1998-2009, which reduces the sample to around 21, 145

has a positive and significant effect. This is consistent with an interpretation in terms of *signal* effect, i.e. urban residents' higher incomes may be informative about migrants' own future income (see also Ravallion and Lokshin, 2000, Senik, 2004, Akay et al., 2012). It may appear as opposed to the (negative) effect of local income when taking the population as a whole (Ferrer-i-Carbonell, 2005). Yet, we observe that this effect also exhausts over time, possibly as migrants assimilate and start to consider their local environment as competitors. Interestingly, the declining (positive) signal effect is symmetrical to the decline of competing feelings vis-à-vis home countries.³¹

Third, we investigate the assimilation process in a more qualitative way. We estimate the potential heterogeneity of the GDP effect among migrants by using different proxies for their connection to their home country. In separate estimations, we use information about the intention to migrate back (wish to stay temporarily or permanently), migrants' attachment to the host country (do you feel like a German?), whether migrants have purchased their dwelling (which may indicate a long-term commitment to stay), objective measures of socio-cultural assimilation (language skills) and on the presence of children and spouses in home versus host countries. Results are reported in Figure 2. The effect of GDP per capita on migrant SWB is ordered, for each of the questions above, from the highest to the lowest connection to the home country. Strikingly, all questions point to the same conclusion: migrants characterized by a strong connection with their home lands show greater relative concerns. Admittedly, the difference with other migrants is not significant when each item is taken separately. Nonetheless, the fact that all measures systematically point to the same direction seems to corroborates our interpretation: those who lose touch with the home land, intentionally or not, also treat it less as a reference point. This is highly consistent with the time pattern discussed above.

Additional Outcomes. Our paper is also related to the burgeoning literature on the effect of migration on well-being. Even if migrants see their economic conditions improve, they may experience a declining SWB due to a fall in their relative position when migrating (Knight and Gunatilaka, 2012). Yet, our results suggest that a shift in reference frame may take time. Stillman et al. (2015) obtain causal effects, thanks to a lottery randomization, which also show that the mere impact of migration on subjective welfare is complex, emphasizing a sensitivity of the SWB impact to the well-being measure at use. Unfortunately our dataset does not include other SWB like mental health. Additional regressions nonetheless show that domains of satisfaction (job, income, health) point to the same result as life satisfaction (a negative effect of log GDP per

migrant-year observations (this is another reason to use QFE rather than FE in this extension).

³¹We believe that such suggestive evidence of a switch in reference groups – theorized by Piore (1979) and Stark (1991) – is original in the literature. The study by Gelatt (2013) suggests that Latino and Asian migrants maintain simultaneous reference groups in both the US and the country of origin. Yet she does not find clear evidence of a shift in reference groups, which may be due to small sample sizes (low test power) or the fact that she cannot capture changes in reference points occurring within migrants' earliest years in the country.

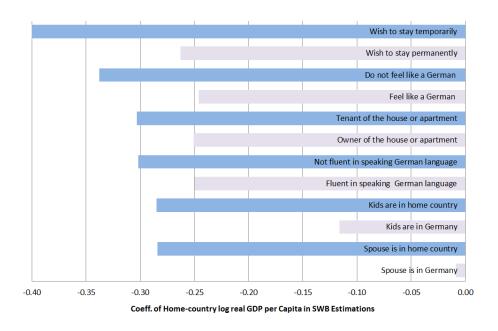


Figure 2: Effect of Home GDP on Migrant SWB: Heterogeneity

capita), yet with insignificant estimates. A question on whether the person is concerned with the economic environment in Germany (3-very concerned, 2-somewhat concerned, 1-not concerned at all) shows a positive significant response to home country log GDP per capita (p-value of 0.09). We have also explored the impact of home country GDP on behavior. While remittances and return migration are discussed above, other dimensions may be of interest. A migrant may, for example, work harder or acquire skills as a response to changes in the home country performances, possibly to improve her relative position to the "average" fellow in the home country. We indeed find that the log GDP per capita increases both hours worked (p-value of 0.01) and education (the effect is significant when restricting estimations to migrants below 40 years old, with a p-value of 0.04).

4 Concluding Discussion

We investigate whether a country's macroeconomic performances matter for those who have migrated. Using various groups of migrants in Germany observed over 26 years, we find a significant and negative effect of home country GDP per capita on migrants' reported well-being. This result is well explained by positional concerns and the idea of relative deprivation of international migrants. Migrants leave their country to improve living conditions, potentially in relation to what they could have achieved in the homeland. We also show that both relative concerns towards home countries and a signal effect from migration regions are stronger upon arrival or for those

with a low degree of assimilation in Germany. Both effects tend to disappear as migrants lose ties with home countries, possibly forming new reference groups among destination regions.

These results bear some interesting implications and suggest further natural developments. First, relative income effects provide an original way to measure assimilation in relation to such policies. Indeed, our approach may allow for the identification of different types of migration dynamics, as discussed in Clark et al. (2008). The modern brain drain view – and the type of workers targeted by migration policy in Canada and the US – would correspond to high-skilled migrants who voluntarily migrate, quickly assimilate and rapidly switch their reference frame. Other migrants from poorer regions and less easily assimilated may keep home countries as the reference point for a longer time. The different types will have different economic and cultural implications for the host country.

Second, the macroeconomic conditions in the home country are one of the most important sources of information to make a cost-benefit calculation not only for initial migration decisions but also for return migration decisions. We could examine how macroeconomic conditions of home countries affect "circular migration", which is an important phenomenon of the last decade (Constant et al., 2013). As noted by Clark et al. (2008), relative concerns can also explain why migrants continue to visit their home countries: this is when they can cash in as relatively high earners compared to those in the home country.

Last, our SWB-based test of the "relative deprivation hypothesis" was only partial. We simply check whether the migrant's relative position with respect to her origin country as a whole – proxied by GDP per capita – may have an effect on her well-being. We could not say anything about how migration improves the relative position of a migrant or her family within the home country income distribution (i.e., the internal relative deprivation hypothesis, as described in the studies of Stark and Taylor, 1991). Interestingly, this hypothesis potentially generates further testable implications. In particular, it implies that the characteristics of the migrants' home country income distribution can influence both the decision to migrate and to return. Further research should attempt to gather more specific information on a migrant's expected labor income in the home country, on her family's position within the home country distribution and on how differential income growth between host and home countries affects this position.

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A Online Appendix

A.1 Descriptive Statistics

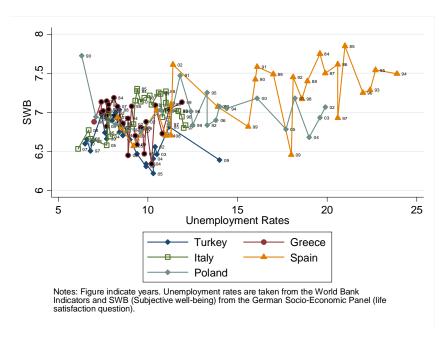


Figure A1: SWB versus Unemployment Across Time for Selected Ethnic Groups

Table A.1: Statistics

Migrants from	(Log) Real GDP	(Log) Nominal GDP	Real GDP: Country / Germany	Real GDP: Country / Germany First Wave	Real GDP: Country / Germany Last Wave	Unempl. rate (%)	SWB (0-10)	correlation SWB & GDP	correlation SWB & unempl.	# Obs. (indiv. x year)
Turkey	9.0	9.5	0.31	0.30	0.35	8.6	6.7	-0.90	-0.40	16,924
	(0.2)	(0.2)	(0.02)			(1.5)	(2.0)			
Greece	9.8	9.9	0.70	0.68	0.80	8.6	7.0	-0.64	-0.34	5,123
Italy	(0.1) 10.1	(0.1) 10.1	(0.05) 0.91	0.93	0.83	(1.3) 10.1	(2.0) 7.1	-0.85	0.34	7,474
italy	(0.1)	(0.1)	(0.02)	0.93	0.63	(1.4)	(1.8)	-0.63	0.54	7,474
Poland	9.4	9.5	0.40	0.32	0.49	14.5	7.0	-0.55	0.04	4,082
	(0.2)	(0.1)	(0.06)			(3.9)	(1.8)			.,
Spain	9.9	9.9	0.76	0.77	0.84	18.4	7.4	-0.86	0.56	3,139
	(0.2)	(0.1)	(0.04)			(3.6)	(2.0)			
Russia	9.2	9.5	0.34	0.49	0.44	8.7	7.3	-0.63	0.47	2,636
	(0.2)	(0.3)	(0.06)			(2.0)	(1.7)			
Kazakhstan	8.8	9.1	0.23	0.18	0.31	9.7	7.3	-0.79	0.68	2,321
Croatia	(0.3) 9.4	(0.4) 9.8	(0.06) 0.43	0.52	0.51	(2.3) 13.3	(1.6) 6.8	0.00	0.59	1,920
Cioana	(0.2)	(0.6)	(0.06)	0.32	0.51	(2.9)	(1.7)	0.00	0.39	1,920
Romania	9.0	9.3	0.28	0.31	0.35	6.9	7.2	-0.14	0.09	1,754
i Comana	(0.2)	(0.2)	(0.04)	0.51	0.55	(0.9)	(1.7)	0.11	0.05	1,751
Bosnia-Herzegovina	8.5	8.6	0.17	0.11	0.22	29.7	6.8	-0.61	-0.28	1,023
	(0.4)	(0.3)	(0.04)			(3.5)	(1.8)			,
Austria	10.3	10.3	1.03	1.00	1.07	4.4	7.4	0.42	0.48	768
	(0.1)	(0.1)	(0.03)			(0.7)	(1.7)			
Czech Republic	9.8	9.9	0.61	0.64	0.69	6.5	6.9	0.12	-0.46	541
	(0.1)	(0.1)	(0.05)			(2.0)	(1.9)			
Ukraine	8.5	8.7	0.16	0.31	0.20	8.9	6.9	-0.35	0.28	515
	(0.2)	(0.4)	(0.03)			(1.9)	(1.8)			
USA	10.6	10.6	1.29	1.24	1.28	5.6	7.5	0.12	-0.21	381
r	(0.1)	(0.1)	(0.04)	0.05	0.00	(1.3)	(1.6)	0.00	0.27	270
France	10.2	10.2	0.93	0.95	0.90	9.7	7.0	0.09	-0.37	379
Netherlands	(0.1) 10.4	(0.1) 10.4	(0.02) 1.09	1.02	1.13	(1.5) 5.0	(1.7) 7.6	-0.17	-0.01	363
recticitatios	(0.1)	(0.1)	(0.04)	1.02	1.13	(2.5)	(1.3)	-0.17	-0.01	303
Hungary	9.6	9.7	0.49	0.48	0.53	6.8	6.9	0.20	0.54	320
	(0.2)	(0.1)	(0.05)		0.00	(2.6)	(2.2)			
Great Britain	10.3	10.3	0.99	0.92	1.01	6.2	7.2	-0.62	0.45	311
	(0.1)	(0.1)	(0.05)			(1.9)	(1.8)			
Macedonia	8.9	9.1	0.24	0.32	0.26	34.1	6.5	-0.12	-0.29	264
	(0.1)	(0.5)	(0.02)			(2.3)	(2.0)			
Slovenia	9.8	10.0	0.65	0.64	0.81	6.9	7.3	-0.51	0.26	248
	(0.2)	(0.2)	(0.08)			(1.2)	(1.5)			
Iran	9.1	9.2	0.28	0.24	0.31	12.3	5.8	-0.06	0.22	200
DI T	(0.1)	(0.1)	(0.03)	0.00	0.40	(2.2)	(2.3)	0.50	0.00	407
Philippines	7.9	8.0	0.09	0.09	0.10	8.6	7.3	-0.52	0.39	187
Portugal	(0.1) 9.9	(0.1) 10.0	(0.01) 0.67	0.63	0.65	(1.4) 6.2	(1.7) 7.5	-0.69	-0.23	170
. 0.11uSm	(0.1)	(0.1)	(0.03)	0.05	0.03	(1.7)	(1.5)	-0.09	-0.43	1/0
Bulgaria	8.9	9.3	0.28	0.29	0.36	12.0	7.3	-0.07	-0.21	128
J	(0.2)	(0.5)	(0.04)			(5.9)	(1.7)			
Mean / total *	9.5	9.6	0.55	0.56	0.60	10.9	7.1	-0.34 [0.46]	0.11 [-0.40]	51,171
	(0.2)	(0.2)	(0.0)			(2.2)	(1.8)			
Germany	10.28	10.29				8.50	6.99			334,308
	(0.1)	(0.1)		are country as		(1.4)	(1.8)			

Note: GDP, unemployment and subjective well-being (SWB) figures are country averages over 1984-2009. GDP (2005 PPP international dollars) and unemployment rate (annual) taken from World Bank Indicators, SWB from the German Socio-Econmic Panel. Standard deviations are reported in parentheses. Correlation between SWB and GDP (or unemployment rate) are calculated over the 26 years using mean SWB for each country-year. The correlations in square brackets in the Mean/total row reflect both time and country variation (24×26 country-year cells).

A.2 Detailed SWB Estimates

Table A.2 reports the complete set of estimates of equation (1). We distinguish between personal determinants of SWB, individual characteristics related to the home countries and macroeconomic variables (here the log real GDP per capita, $GDP_{h,t}$). The specifications I and II relate to models 1 and 2 in Table 1: FE model with and without country-specific time trends. Specification 0 just checks what happens if we ignore home country GDP. All specifications control for time-varying characteristics, German state and year effects. Results are in line with standard findings in the literature (as surveyed in Clark et al., 2008). Essentially, income, good health and being married are positively related to SWB while being unemployed is negatively correlated. The presence of children in Germany has strong positive effects. Migrants' refugee status affects SWB negatively. The level of remittances is negatively correlated (the loss of resources endured by the migrant dominates the gains from remitting: altruism, investment in social capital in home country, etc.) but insignificant.

Comparing models 0 and I shows that the signs and significance of individual characteristics are not affected much by the inclusion of $GDP_{h,t}$. In model I, we obtain an estimate of the GDP effect of -.303, which is significant at the 1% level. Model II controls for country-specific time trends to clean out the spurious correlation between macroeconomic indices and SWB. The magnitude of the effect is basically unchanged (-.212) but the effect is less precisely estimated, even if still significant at the 10% level.

We have also run separate regressions for each country and find that life satisfaction estimates have a broadly common structure overall (detailed results are available from the authors). The impact of variables like income, health, marital status and children is very comparable and stable across countries of origin. This regularity suggests that SWB data contain reliable and potentially interesting information for welfare measurement (see also Di Tella et al., 2003).

Table A.2: Subjective Well-Being Regressions with Alternative Specifications

Dep. variable: SWB	0	I	H		0	I	II
Personal characteristics				Personal characteristics related to origi	n country		
Log of HH income	0.376 ***	0.380 ***	0.384 ***	One children with the migrant	0.096 ***	0.093 ***	0.092 ***
	(0.034)	(0.034)	(0.023)		(0.033)	(0.034)	(0.027)
Non-employed	-0.001	0.014	0.011	Two children with the migrant	0.125 ***	0.129 ***	0.127 ***
	(0.046)	(0.047)	(0.042)		(0.042)	(0.043)	(0.033)
Unemployed	-0.419 ***	-0.401 ***	-0.402 ***	More than two children	0.208 ***	0.223 ***	0.218 ***
	(0.057)	(0.058)	(0.047)		(0.057)	(0.057)	(0.042)
Old age/retired	0.043	0.035	0.042	Spouse in home country	-0.435 ***	-0.496 ***	-0.486 ***
	(0.076)	(0.076)	(0.060)		(0.140)	(0.144)	(0.092)
In training/education	0.119 *	0.110	0.102	Other relative in home country	0.001	-0.011	0.009
	(0.074)	(0.076)	(0.067)	•	(0.090)	(0.089)	(0.087)
Self-employed	0.033	0.033	0.027	Migrant is a refugee	-0.184 **	-0.143 *	-0.141 *
* *	(0.071)	(0.072)	(0.060)		(0.082)	(0.083)	(0.084)
Log of working hours	0.041 ***	0.044 ***	0.044 ***	Log of remittances	-0.006	-0.005	-0.005
	(0.012)	(0.012)	(0.011)		(0.010)	(0.010)	(0.008)
Separated (1)	-0.394 ***	-0.407 ***	-0.410 ***	Macroeconomic conditions		, ,	, ,
1 (7	(0.098)	(0.100)	(0.065)	GDP		-0.303 ***	-0.212 *
Single (1)	-0.216 ***	-0.237 ***	-0.224 ***			(0.107)	(0.125)
	(0.065)	(0.066)	(0.046)				
Divorced (1)	-0.208 **	-0.219 **	-0.231 ***	Individual effects	FE	FE	FE
	(0.101)	(0.103)	(0.070)	State effects	Yes	Yes	Yes
Widowed (1)	-0.571 ***	-0.578 ***	-0.596 ***	Year effects	Yes	Yes	Yes
	(0.130)	(0.135)	(0.094)	Home country linear time trend	No	No	Yes
Health: poor (2)	0.709 ***	0.705 ***	0.702 ***	R2	0.141	0.140	0.141
	(0.054)	(0.054)	(0.037)	# observations	47,557	47,557	47,557
Health: average (2)	1.290 ***	1.287 ***	1.283 ***				
	(0.056)	(0.056)	(0.036)				
Health: good (2)	1.776 ***	1.779 ***	1.773 ***	Note: *, **, *** indicate significan	æ levels at 10	%, 5% and 1%	6 respectively.
	(0.058)	(0.058)	(0.037)	Estimations performed on migrant		,	
Health: very good (2)	2.255 ***	2.257 ***	2.253 ***	errors dustered at the individual leve		0	1 1 .
	(0.062)	(0.062)	(0.039)	taken from World Bank indicators. 3 German Socio-Economic Panel. (1)	,	0 ()	
Log of HH size	-0.292 ***	-0.313 ***	-0.308 ***	category is 'very poor health'. Unobse			` '
•	(0.053)	(0.053)	(0.037)	using fixed effects (FE). State effect d			
Years of education	-0.010	-0.010	-0.008	- , ,			•
	(0.012)	(0.012)	(0.009)				

A.3 Estimations on Grouped Data

Grouped data estimation is an alternative to estimations on individual migrant observation. We use a sample of 556 country×year points,³² taking the mean SWB over all migrants in a country-year cell as the dependent variable. The model becomes:

$$\overline{SWB}_{ht} = \overline{X}_{ht}\alpha + \gamma Macro_{ht} + \theta_t + t \times \delta_h + \delta_h + \overline{Z}_h + \overline{Age}_{ht} + \overline{YSM}_{ht} + \varepsilon_{ht}$$

where \overline{SWB}_{ht} is the mean subjective well-being over all migrants of origin country h in year t, $Macro_{ht}$ the home country macroeconomic variable (we focus on log real GDP per capita, $GDP_{h,t}$, and unemployment hereafter), \overline{X}_{ht} a set of mean characteristics of migrants from country h observed in year t (the characteristics listed in Table A.2) and $\overline{Z}_h + \overline{Age}_{ht} + \overline{YSM}_{ht}$ the means of gender and cohorts, age, and years-since-migration. The composite error term includes time trends θ_t (for any global shocks that are common to all countries in each year), country-specific time trends $t \times \delta_h$ (cultural attitude toward changes in well-being or country-specific unobservable assimilation patterns of migrants of country h), home country fixed effects h0 (for unchanging cultural influences of origin country on reported well-being), and a usual i.i.d error term, h0. Regressions are weighted by cell sizes to account for the larger representation of some migrant groups in the data and to make them more comparable to regressions on individual data. This grouped data estimation is similar to the micro data estimations when assuming that individual FE h1 average up to h2 average up to h3 average up to h4 average up to h6 average h7 average up to h8 average up to h9. The likely departure from these assumptions will explain the difference with micro estimates.

Effect of GDP. In Table A.3, we simply report estimates for γ , which is the impact of the macroeconomic variables on SWB. Column I reports the coefficient on $GDP_{h,t}$. The parameter estimate is negative and highly significant, with a magnitude of -.565. Hence, it is confirmed that an increase in the home country's GDP per capita is negatively correlated with migrants' well-being, conditional on country and year fixed effects. Column II departs from the assumption

 $^{^{32}}$ We do not have observations in the GSOEP for 1 year (5, 5, 6 and 10 years) in Iran (Portugal, Russia Ukraine and Kazakhstan respectively), which makes 27 country × year observations missing. We have checked that the conclusions of this study hold when excluding these countries completely. In addition, macroeconomic variables are not reported in World Bank indicators for 6 years in Poland, Slovenia, Macedonia, Croatia and the Czech Republic, 1 year for Russia and 10 years for Bosnia, leading to another 41 missing points. Again, we have verified that our results are consistently similar when using linear extrapolation or other sources to fill in the missing GDP or unemployment information. Our baseline nonetheless relies on the original sample. The total of 68 missing points corresponds to 10.9% of the $26 \times 24 = 624$ country × year sample used for grouped estimations below. This proportion is smaller in terms of individual×year observations (7.1%) due to the fact that missing points affect countries that are below the average country size.

of common linear time trends for all countries by adding $t \times \delta_h$.³³ As in micro estimates, the coefficient becomes a little bit smaller but the relationship between $GDP_{h,t}$ and SWB is hardly affected. The coefficient, -.472, is significant and gives a 95% confidence interval of [-.99, .04]. Corresponding regressions on individual migrant data (columns II and V of Table A.4) yield overlapping intervals of [-.457, .033] and [-.479, .031] in the case of FE and QFE respectively.



Effects of Unemployment. Our relative concerns/deprivation interpretation could apply to other macroeconomic variables and notably to unemployment. Market failures that constrain labor market and earnings opportunities in the homeland may increase the attractiveness of migration both as a potential avenue for effective gains in relative incomes and a source of satisfaction for those who have already migrated. Column III in Table A.3 presents the effect of the home-country unemployment rate. This effect is significantly positive, which is consistent with the interpretation above and the findings regarding GDP. This effect is robust to controlling for home country specific time trends (column IV). When including $GDP_{h,t}$ in the same regression (unreported), both home country log GDP per capita and unemployment effects keep the sign and magnitude that they had in independent estimations.

³³This is a necessary check, as argued by Di Tella et al. (2003). Indeed, as macroeconomic indices such as GDP are time-trended while SWB is usually untrended (Easterlin, 1995), regressing the latter on the former generates concerns of costationarity. In our sample of migrants, we have observed a small downward trend in life satisfaction. We nonetheless account for time trends θ_t in the estimation to reduce this concern. Including country×year effects – and hence accounting for possible differences in slope across source countries – should eliminate it. Note also that the GDP effect could be spurious if country-specific time effects, and in particular the effect of YSM, were misspecified and picked up by the GDP trend. While country-specific time trends eliminate this, we have checked that our results are not sensitive to using flexible specifications of YSM in a model without country-specific time effects.

Table A.3: Effect of Home Country Macroeconomics on Migrant SWB: Grouped Estimations

SWB grouped estimations	I	II	III	IV
GDP	-0.565 ***	-0.472 *		
	(0.202)	(0.263)		
Unemployment rate			0.040 ***	0.030 ***
			(0.010)	(0.011)
Year effects	Yes	Yes	Yes	Yes
Home country fixed effects	Yes	Yes	Yes	Yes
Home country linear time trends	No	Yes	No	Yes
GDP (equivalent income)	-1.45	-1.21		
R2	0.637	0.685	0.587	0.673
# observations	556	556	556	556

Notes: *, ** and *** indicate significance levels at 10%, 5% and 1% respectively. GDP refers to log of real GDP per capita. GDP and unemployment rates taken from World Bank indicators. Subjective well-being (SWB) averaged per country of origin x year, taken from the German Socio-Economic Panel. Linear estimations performed on migrants from 24 countries over 26 years, weighted by country x year cell size. All models include the mean value (for each country x year) of characteristics reported in Appendix Table A.2 (including mean cohort and state effects).

A.4 Estimations on Micro Data: Additional Results

Table A.4: Effect of Home Country GDP on Migrant SWB: Micro Data

SWB micro estimations	I	П	III	IV	V	VI	VII
GDP (coefficient)	-0.303 *** (0.107)	-0.212 * (0.125)	-0.490 *** (0.181)	-0.281 *** (0.104)	-0.224 * (0.130)	-0.321 *** (0.122)	-0.215 *** (0.057)
Individual effects (a)	FE	FE	FE	QFE	QFE	QFE#	QFE
Cohort fixed effects (b)	n.a.	n.a.	n.a.	Yes	Yes	Yes	Yes
State effects (c)	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Home country fixed effects	n.a.	n.a.	n.a.	Yes	Yes	Yes	Yes
Home country linear time trends	No	Yes	No	No	Yes	No	No
Estimation method	linear	linear	ologit	linear	linear	linear	oprobit
GDP (equivalent income)	-0.797	-0.553	-0.972	-0.714	-0.562	-0.860	-0.689
R2 or pseudo-R2	0.140	0.141	0.103	0.284	0.285	0.305	0.085
# observations	47,557	47,557	47,557	47,557	47,557	25,306	47,557

Notes: *, **, *** indicate significance levels at 10%, 5% and 1% respectively. Estimations performed on migrants from 24 countries over 26 years, standard errors dustered at the individual level. GDP refers to log of real GDP per capita, taken from World Bank indicators. Subjective well-being (SWB) taken from the German Socio-Economic Panel. All models include the full set of observed characteristics as reported in Appendix Table A.2 (time-invariant characteristics, age and years-since-migration not used with fixed effects). (a) Unobserved individual effects are taken into account using fixed effects (FE), quasi-fixed effects (QFE) or QFE and big-five personality traits (QFE#). Other individual effects are: (b) 10 arrival cohort effects (used with QFE only) and (c) 16 federal states of Germany.

Table A.5: Effect of Home Country Unemployment on Migrant SWB: Micro Data

SWB micro estimations	A	В	C	D	E	F	G
Unemployment rates	0.011 ***	0.009 **	0.006	0.005	0.002	0.007	0.009
	(0.004)	(0.004)	(0.004)	(0.004)	(0.005)	(0.006)	(0.007)
Unemployment rate (t-1)						-0.002	0.005
						(0.006)	(0.009)
Unemployment rate (t-2)							-0.011
							(0.007)
GDP		-0.374 ***			-0.417 **	*	
		(0.115)			(0.148)		
Individual effects (a)	No	No	QFE	FE	FE	FE	FE
Cohort fixed effects (b)	Yes	Yes	Yes	n.a.	n.a.	n.a.	n.a.
State fixed effects (c)	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Home country fixed effects	Yes	Yes	Yes	n.a.	n.a.	n.a.	n.a.
R-squared	0.289	0.289	0.284	0.139	0.139	0.140	0.139
#Observations	47,557	47,557	47,557	47,557	47,557	47,398	47,231

Notes: *, **, *** indicate significance levels at 10%, 5% and 1% respectively. Linear estimations performed on migrants from 24 countries over 26 years. All models include the full set of observed characteristics as reported in Appendix Table A.2. Unemployment rates and GDP (referring to log of real GDP per capita) are taken from World Bank indicators. Subjective well-being (SWB) taken from the German Socio-Economic Panel. Other controls include: (a) Unobserved individual effects modeled as quasi-fixed effects (QFE) or fixed effects (FE), (b) 10 arrival cohort effects, (c) 16 federal states of Germany.

A.5 Return Migration

We use the Heckman procedure adapted to panel data by simultaneously estimating selection into return migration and the SWB equation by Maximum Likelihood (for a more structural approach, see Bellemare, 2007). Ideally, the selection equation should contain an instrument explaining variation in migrants' likelihood to return but uncorrelated with (conditional) migrants' SWB. There is no obvious variable of the kind, as virtually everything can potentially affect well-being. We use a first series of instruments based on the migrant's declared intention to stay in Germany (contemporaneous, lagged and time change of this intention). We also use the average intention to stay over all the migrant's household members (also as contemporaneous, lagged or time difference), which is expected to be more exogenous but possibly less relevant as an instrument. Column 2 in Table A.6 reports the effect of the different instruments on the propensity to return. All instruments have a significant impact and the expected sign (F-tests pass the threshold of 10 commonly used for checking if instruments are weak). Column 3 reports the effect of $GDP_{h,t}$ on the probability of return: it is positive but insignificant. As discussed in the text, column 1 shows that SWB regressions controlling for selection into return migration yield very similar GDP effects as the baseline. The correlation ρ between the residuals of the two equations is significantly different from zero only when the instrument used is the contemporaneous intention to stay (the migrant's intention or the mean answer for her family), which denotes the possible role of unobservable shocks simultaneously affecting well-being and the current intention to return.

Table A.6: SWB Estimations Corrected for Selection into Return Migration

SWB estimation with Heckman	SWB equation	Propensity to	return equation		
correction for return migration	Coeff. on GDP	Coeff. on instrument	Coeff. on GDP	Rho	# obs.
Instrument: Migrant's intention to stay					
Intention (t)	-0.245 ***	-0.395 ***	0.065	0.085 **	47,568
	(0.095)	(0.019)	(0.134)	(0.036)	
Intention (t-1)	-0.314 ***	-0.336 ***	0.116	0.007	40,961
	(0.099)	(0.021)	(0.146)	(0.036)	
Intention (t) - Intention (t-1)	-0.316 ***	-0.064 ***	0.094	-0.005	40,961
	(0.099)	(0.021)	(0.145)	(0.038)	
Intention (t-1) - Intention (t-2)	-0.254 **	-0.052 **	0.100	-0.021	35,664
	(0.105)	(0.022)	(0.157)	(0.043)	
Instrument: mean intention to stay of migrand	t's household				
Intention (t)	-0.249 ***	-0.460 ***	0.071	0.062 *	47,568
	(0.095)	(0.021)	(0.135)	(0.034)	
Intention (t-1)	-0.316 ***	-0.397 ***	0.123	-0.008	40,961
	(0.099)	(0.023)	(0.146)	(0.035)	
Intention (t) - Intention (t-1)	-0.316 ***	-0.086 ***	0.092	-0.005	40,961
	(0.099)	(0.024)	(0.145)	(0.038)	
Intention (t-1) - Intention (t-2)	-0.254 **	-0.066 ***	0.099	-0.021	35,664
	(0.105)	(0.026)	(0.157)	(0.043)	

Note: *, **, *** indicate significance levels at 10%, 5% and 1% respectively. SWB equation estimated linearly on microdata using baseline specification and additionally accounting for Heckman correction for non-random selection into return migration (ML estimation). Selection based on a dummy variable for return migration. Different rows report results for alternative instruments in the selection equation. Instruments are based on the migrant's intention to stay or her household mean intention to stay. Rho is the correlation between the two equations.