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The Dynamics of Relative Poverty in China in a Comparative Perspective

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Abstract

I use household panel data to study the dynamics of relative poverty in China, Germany, the UK, and the US. Compared to the three Western countries, not only is relative poverty more common in China, it is also deeper and more severe. Transient poverty accounts for less than half of the total poverty in Germany or the US, but about two thirds of that in China or the UK. Over three waves, 87 percent of Germans, 78 percent of Brits, 71 percent of Americans, but only 46 percent of Chinese were never poor. Using a multinomial logistic regression model, the determinants of poverty are found to be very similar across the four countries. But the variance explained of that model is much smaller for China than for the three Western countries. The results also raise questions about some regularities reported in previous research on poverty dynamics.

Keywords: relative poverty, panel data, cross-national comparison, China

JEL Codes: 132, 057

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1 Introduction

China has a very impressive record of poverty reduction. Using the \$1-aday poverty line, 65 percent of the Chinese population were poor in 1981. This dropped to 10 percent in 2004, meaning that half a billion people were lifted out of poverty (World Bank, 2009). And since 'the absolute number of poor in the developing world as a whole declined from 1.5 to 1.0 billion over the same period ... but for China there would have been no decline in the numbers of poor in the developing world over the last two decades of the 20th century' (World Bank, 2009, p. iii).

Against the backdrop of this clear policy achievement, there is much about poverty in China that is not well understood. This paper contributes to research in this field in three ways. First, existing research is mainly concerned with absolute poverty. This was quite appropriate when China was a very poor country. But after four decades of sustained economic growth, 'extreme poverty, in the sense of not being able to meet the most elementary food and clothing needs, has almost been eliminated in China' (World Bank, 2009, p. iv). Given this, I focus on relative poverty in this paper.

Secondly, previous studies mostly draw on repeated cross-sectional data (for exceptions, see Jalan and Ravallion, 1998, 2000). Although very informative about the aggregate poverty trend, they are silent on the poverty dynamics of individuals. To elaborate, suppose that half of the population were found to be poor on two occasions. This pattern of no overall change might have arisen because everyone's poverty status was stable over time. Alternatively, everyone's poverty status might have changed: those who were poor initially had become non-poor, and vice versa. The welfare implications of these two scenarios and the policy interventions that they call for are quite different (Atkinson, 2019, p. 63). To determine the poverty dynamics of individuals, I analyse household panel data in this paper.

Thirdly, I carry out parallel analyses for China, Germany, the UK, and the US. Large-scale and nationally representative household panel data are hard to come by. Consequently, there are just a few papers that compare poverty dynamics across countries; and they all have a Western focus (Duncan *et al.*, 1993; Layte and Whelan, 2003; Fouarge and Layte, 2005; Valletta, 2006). As high quality Chinese household panel data have recently become available, I am able to compare China with Germany, the UK, and the US. As it turns out, this comparison brings out just how distinctive the Chinese case is. It also revises some of our understanding of poverty dynamics in general.

The rest of this paper is structured as follows. In Section 2, I review past research on absolute poverty in China and also research on poverty dynamics in Western societies. Section 3 introduces the data sets and explains the key operational decisions that I have made. Section 4 reports the results in six parts. Finally, in Section 5, I summarise the results and discuss their implications for understanding both poverty in China and the dynamics of relative poverty in general.

2 Context and previous research

2.1 Absolute poverty in China

Scholars, government agencies, and international organisations have used different poverty lines in their research. So there is a range of estimates of the level of poverty in China (see e.g. Li *et al.*, 2013, p. 74, Table 2.16). But all studies point to a large decline over the past forty years, irrespective of which poverty line is used (Ravallion and Chen, 2007; Zhang *et al.*, 2014).

Appleton *et al.* (2010) argue that the main driver of poverty reduction in China is economic growth, not redistribution or anti-poverty programmes. And since absolute poverty in China is 'predominantly a rural phenomenon' (Gustafsson and Zhong, 2000, p. 984), it was '[g]rowth in the primary sector (primarily agriculture) [that] did more to reduce poverty' (Ravallion and Chen, 2007, p. 2). The critical policy change that spurred rural growth in the 1980s was the decollectivisation of agriculture and the return to family farming (Oi, 1989).

It is well-known that income inequality in China has risen very sharply since the mid-1980s (Xie and Zhou, 2014; Chan *et al.*, 2019). It might be thought that a higher level of inequality is the necessary price to pay for the economic growth that is needed to reduce poverty. But there is no evidence for this. As Ravallion and Chen (2007, p. 3) observe, '[t]he periods of more rapid growth did not bring more rapid increase in inequality. Nor did provinces with more rapid rural income growth experience a steeper increase in inequality. Thus provinces that saw a more rapid rise in inequality saw less progress against poverty, not more.'

The progress in poverty reduction was uneven, with '[h]alf of the decline in the number of poor came in the first half of the 1980s' (Ravallion and Chen, 2007, p. 2). After the mid-1980s, the poverty headcount continued to fall, but there were reverses and the rate of decline slowed down considerably. Spatially speaking, there is more poverty in Western China than in the central provinces which, in turn, are poorer than the provinces on the Eastern coast (Gustafsson and Sai, 2009). Li *et al.* (2013, p. 76) observe that '[b]y all measures, China's poor is heavily concentrated in the West.'

Of even greater importance than the regional difference is the urban-

rural divide. Ravallion and Chen (2007, p. 8) show that '[f]or all years and all measures, rural poverty incidence exceeds urban poverty and by a wide margin.' Similarly, Li *et al.* (2013, p. 75) report that '[u]sing absolute poverty measures, more than 95 percent of the poor were rural.'

The concentration of poverty in rural areas is a direct consequence of the Maoist strategy of fuelling industrialisation by squeezing the countryside (Oi, 1989). Before the market reform, urbanites enjoyed secure employment and a wide range of benefits and services that were provided through their work units (*danwei*). Although meagre by Western standards, income and consumption were far higher in Chinese cities than in the countryside.

This created a strong incentive for peasants to move to the cities. To restrict rural-urban migration, a system of household registration (hukou) was introduced in the 1950s which has for decades 'effectively bound peasants to the soil' (Whyte, 2005, p. 10). Restriction on migration has gradually been relaxed and there are now over 200 million internal migrants in China (Liang *et al.*, 2014). But *hukou* still exists as a legal category. Migrant workers living in cities do not have access to health care, education, or other public services that are available to those with urban *hukou* (Chan and Zhang, 1999). Furthermore, as migrants tend to work in unskilled, low-wage jobs, they are typically found at the bottom of the urban economic hierarchy.

Despite that, the impact of migration on urban poverty seems quite small. Park and Wang (2010) show that in terms of housing conditions and other non-monetary welfare indicators migrants are indeed worse off than nonmigrants. Migrants also earn lower hourly wages. But as they tend to work longer hours and have a lower dependency ratio in their household, the gap in disposable income between migrant and non-migrant households is smaller than might be expected. Overall, Park and Wang (2010, p. 55) conclude that the difference in poverty rate among migrants and non-migrants is very small and that 'including migrants has a negligible impact on the overall urban poverty estimates'.

In the countryside, remittances from migrants certainly help raise household income. But Du *et al.* (2005, p. 706) point out that '[t]he poorest rural households with few laborers and poor human capital are unable to allow members to migrate.' Thus 'the overall impact of migration on [rural] poverty headcount has been modest.'

Although poverty in China is concentrated in the countryside, urban Chinese have been facing greater poverty risks since the mid-1990s. In particular, state-owned enterprise (SOE) reforms had led to 28 million workers (about a quarter of the SOE workforce) being laid off (Appleton *et al.*, 2014). This contributed to the doubling of the urban unemployment rate from 6 to 12 percent between 1993 and 2000 (Meng and Gregory, 2007).

Other reforms also put economic pressure on city dwellers. In particular, the Chinese state used to subsidise urbanites' food consumption through a coupon system. The value of the coupons distributed to each household was a function of its size and the age of its members. The coupon system was abolished in 1993 and food price control was lifted. Although urban wages had gone up, giving urbanites greater spending power to cope with food price inflation, larger households with few working members lost out in this reform (Meng and Gregory, 2007). Also, urbanites now need to pay for many public services, e.g. education and health care, that were previously free or heavily subsidised.

In response to the growing economic hardship in the cities, the Chinese state piloted a minimum living standard guarantee programme (*dibao*) in Shanghai in 1993, which was then rolled out across urban China in 1999. In 1999, 2.7 million urbanites were enrolled on *dibao*, rising to 23.4 million in 2008 (Gustafsson and Deng, 2011). A *dibao* programme for rural China was launched later and became nationwide in 2007. By 2012, it covers 53 million people, about 8 percent of the rural population (Li and Sicular, 2014).

As with many social programmes in China, the implementation of *dibao* is very decentralised. Municipal and provincial authorities have a lot of leeway in setting the *dibao* line, determining the eligibility criteria, and so on (Ravallion, 2014). The World Bank (2009, pp. 123–124) reports that in 2004 the income threshold for rural *dibao* ranges between 120 yuan and 1,560 yuan per person per year. In any case, due to the restricted coverage of the *dibao* programmes, their impact on poverty reduction is limited (Chen *et al.*, 2006; Li and Sicular, 2014). Overall, we do know a good deal about absolute poverty in China. But, as I will show below, relative poverty is a different story altogether.

2.2 Cross-national research on relative poverty dynamics

There are only a few cross-national comparative papers on the dynamics of relative poverty. Layte and Whelan (2003) and Fouarge and Layte (2005) analyse panel data from eleven countries that took part in the European Community Household Panel.¹ They show that 'the pattern of poverty persistence is congruent with welfare regime theory' (Layte and Whelan, 2003, p. 167), with 'social democratic regimes reducing the level of persistent and recurrent poverty. Liberal and Southern European regime countries have

¹The eleven countries are Belgium, Denmark, France, Germany, Ireland, Italy, Greece, the Netherlands, Portugal, Spain, and the UK.

Table 1: Summary of Duncan <i>et al.</i> 's (1993) observations of cross-national
association between relative poverty level and poverty entry and exit rates
(top panel), and the mechanism mediating these associations (bottom panel)

	association			
	exit rate	entry rate		
higher poverty level	(1) lower	(2) higher		
	mechanism: typical distance to poverty lin			
	poor person	non-poor person		
higher poverty level	(3) larger	(4) smaller		

both higher rates and longer duration of poverty' (Fouarge and Layte, 2005, p. 407).

With a more microscopic lens, Duncan *et al.* (1993) analyse panel data from Canada, France, Germany, Ireland, Luxembourg, the Netherlands, Sweden, and the US. They make several observations that are especially relevant to this paper. First, if the relative poverty line is defined with reference to each country's median income, 'the resulting poverty estimates reflect the degree of inequality of the distribution of size-adjusted family income' (Duncan *et al.*, 1993, p. 217). That is to say, relative poverty tends to be higher in countries with greater income inequality (see also Atkinson, 2015, pp. 25–26).

Secondly, there are 'striking [cross-national] differences in the prevalence of transitions out of poverty' (Duncan *et al.*, 1993, p. 221). These differences are related to how much relative poverty there is. 'Countries with larger fractions of their populations below the poverty line have lower escape rates' (Duncan *et al.*, 1993, p. 221). Escape rates also depend on how far below the poverty line the typical poor person is, '[s]ince crossing a poverty line is generally easiest when a family's income is close to the line' (Duncan *et al.*, 1993, p. 218). Thus, for example, Duncan *et al.* (1993, Tables 1 and 2) show that, compared to their European counterparts, the poor in America (especially African American poor) are further away from the poverty line, and their escape rate is, correspondingly, the lowest of all in the sample.

Thirdly, Duncan *et al.* (1993, pp. 224–225) report 'a decidedly inverse relationship between the fraction of the population at risk of entering poverty and the fraction who actually fall below the line.' Put differently, poverty entry rates are lower in countries with less relative poverty. They offer a similar explanation for this association: 'the greater the fraction of the population above poverty, the greater is the likely distance between the typical nonpoor family and the poverty threshold and the less likely that family is to make the transition' (Duncan *et al.*, 1993, p. 225). Table 1 summarises these suggested associations and mechanisms (see also Jäntti and Danziger, 2000, pp. 355–358). As I will show below, the cross-national comparative analyses of this paper suggest that some revisions to the second and third claims above are necessary.

3 Data

The data that I analyse are collected in four household panel surveys, namely the China Family Panel Studies (CFPS), the German Socioeconomic Panel (GSOEP), Understanding Society (USOC) for the UK, and the Panel Study of Income Dynamics (PSID) for the US.

CFPS is run by a team based at the Institute of Social Science Survey of Peking University. Its sample is drawn from 25 provinces or cities. Together these provinces and cities account for over 94 percent of the Chinese population.² So the survey can be considered as (nearly) nationally representative. The CFPS was launched in 2010, in which 14,960 households and 42,590 individuals were interviewed using computer-assisted personal interviews.³

GSOEP began in 1984 with a sample of about 6,000 households and over 12,000 individuals. USOC is the successor to the British Household Panel Survey. It began in 2009, with a sample of about 30,000 households and over 47,000 individuals. PSID began in 1968 with a sample of about 5,000 households and over 18,000 individuals. GSOEP and USOC are annual panel surveys, but CFPS and PSID are biennial surveys.⁴ To maintain comparability across the four surveys, I use the German and British data from every other wave. For PSID, GSOEP, and USOC, the data are drawn from 2009, 2011, and 2013, whereas for CFPS, the data are from 2010, 2012, and 2014.⁵

Because households are formed and dissolved over time, they do not have a stable identity which can 'be followed over time in a consistent manner' (Jenkins, 2011, p. 36). For this reason, the unit of analysis in this paper is the individual, *not* the household. To factor out life-cycle effects on poverty, I restrict my analysis to individuals aged 31 to 55. In other words, child poverty and pensioner poverty are beyond the scope of this paper. These are

²The provinces or regions not included in CFPS are Hainan, Hong Kong, Macau, Ningxia, Qinghai, Tibet, and Xinjiang.

³The response rate and attrition rate of CFPS are very good by international standards. For example, the response rate at wave 1 at the individual level was 84.1% and the attrition rate between waves 1 and 2 was 80.6% (Xie and Hu, 2013, pp. 10–11). For further details, see http://www.isss.edu.cn/cfps/EN/ and Xie *et al.* (2015).

⁴PSID was an annual panel survey until it turned biennial in 1999.

⁵The fieldwork of each wave of Understanding Society takes two years. Thus wave 1 data were collected in 2009/10, wave 2 was from 2010/11, and so on.

very important issues in their own right and deserve more focused treatment in separate papers.

The main variable of interest is real equivalised household disposable income, i.e. the total post-tax, post-transfer household income after adjustments for inflation and household size. I use the relevant consumer price index to adjust for inflation.⁶ To take household size into account, I divide the total net household income by the square root of household size. If individuals move between households from one interview to the next, their income is calculated on the basis of the household of which they are currently a member.

Finally, following a convention in research on income mobility and poverty dynamics, I drop the top one percent and the bottom one percent of the income distribution from the analysis (Gottschalk and Moffitt, 2009, p. 10). This reflects the fact that the very rich and the very poor are not well captured by household surveys. Such data trimming 'inoculate[s] estimates against the adverse effects of outlier income values and changes' (Jenkins, 2011, p. 123).

4 Results

4.1 Relative poverty lines

I use two relative poverty lines in this paper, namely 60 percent and 50 percent of the median equivalised household income, determined separately for each country and each wave. These are the two most commonly used poverty thresholds in the literature (see e.g. Smeeding *et al.*, 2002; Layte and Whelan, 2003; Fouarge and Layte, 2005). Indeed, 60%-of-median-income 'is the principal poverty line used in Britain's official income distribution statistics ... [it] is also the threshold focused on in official EU reports on poverty and social exclusion' (Jenkins, 2011, p. 207). As the two poverty lines give very similar results, the discussion in the main text will be based on 60%-of-median-income. But I will reproduce all Figures and Tables using the 50%-of-median-income poverty line in the online appendix.

Table 2 reports the poverty line of the four countries (and also for rural and urban China separately) at wave 1. The rightmost column (which is in purchasing power parity dollar) shows that the US poverty line is eight times higher than the Chinese poverty line; and the German and the UK poverty lines are about seven times higher. By implication, these ratios apply to the

⁶The consumer price indices that I use are taken from the OECD, see https://data.oecd.org/price/inflation-cpi.htm.

Table 2: Poverty line (60%-of-median-income) of the four countries at wave 1 in local currency and purchasing power parity (PPP) dollar*

	local currency	PPP \$
DE	€ 12,067	14,879
UK	$\pounds 10,350$	14,744
US	\$ 17,789	17,789
CN	¥ 7,275	$2,\!199$
CN(U)	¥ 9,546	2,886
CN(R)	¥ 5,576	$1,\!686$
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* Source: data.oecd.org/conversion/purchasing-power-parities-ppp.htm

median income of the four countries too. China's market reform has made it the second largest economy in the world. Shorrocks *et al.* (2019, p. 45) report that in 2019 China has '100 million members of the global top 10% [of the wealth distribution], overtaking for the first time the 99 million members of the United States.' Be that as it may, Table 2 reminds us that the typical Chinese is still very much poorer than their Western counterparts. Reflecting the very large gap in living standards between urban and rural China, Table 2 also shows that the poverty line and median income for urban China are 71% higher than those for rural China.

4.2 The incidence, depth, and severity of relative poverty

Figure 1 reports the poverty headcount of the four countries. In wave 1, one in ten Germans, one in eight Brits, one in five Americans, and a quarter of Chinese were poor.⁷ The fact that relative poverty is more common in the US than in Europe is well documented (Smeeding *et al.*, 2002). But it is quite striking that China's relative poverty rate is even higher than that of the US. Furthermore, although absolute poverty in China is 'predominately a rural phenomenon' (Gustafsson and Zhong, 2000, p. 984), relative poverty is as much an issue in Chinese cities as in the countryside. Broadly speaking, the same ranking also holds for waves 2 and 3.

The headcount rates of Figure 1 refer to the percentage of people who are poor. They do not take into account how far below the poverty line the poor are. To gauge the depth and severity of poverty, I use the well-known FGT index (Foster *et al.*, 1984). Suppose we observe the income, y_i , of a sample

 $^{^{7}}$ Within China, the headcount rates for urban and rural areas, at 27% and 26% respectively, are very similar. This is because the poverty lines are determined separately for rural and urban China.

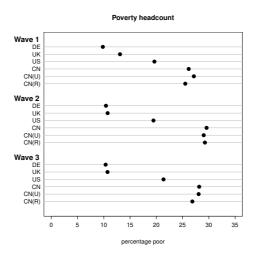


Figure 1: Headcount poverty rate by country and by wave (based on the 60%-of-median-income poverty line)

of N individuals, and that z is the poverty line, the FGT index is given by the following expression:

$$\mathsf{FGT} = \frac{1}{N} \sum_{y_i < z} \left(\frac{z - y_i}{z} \right)^k.$$

The key parameter is k. With k = 0, everyone with income below the poverty line contributes the same amount (i.e. 1) to the FGT index. Indeed, this gives the poverty headcount. If k = 1, individuals with income further below the poverty line contribute more to the index, as the difference, $z - y_i$, is greater for them. This is often referred to as the poverty gap index. With k = 2, we get the squared poverty index which gives even greater weight to those with the lowest income. The poverty headcount index, the poverty gap index, and the squared poverty index can be taken as measuring the incidence, depth, and severity of poverty respectively (Ravallion, 2016).

Figure 2 reports the poverty gap (left panel) and the squared poverty gap (right panel) of the four countries. They show the same ranking as the poverty headcount index, with Germany and the UK at the low end, China at the high end, and the US in the middle. This ranking holds for all three waves, though the cross-national differences have widened over time. To recap, compared to the three Western countries, not only is relative poverty more common in China, it is also deeper and more severe. Also, while absolute poverty is primarily a rural phenomenon, relative poverty afflicts urban China as much as it does the countryside.

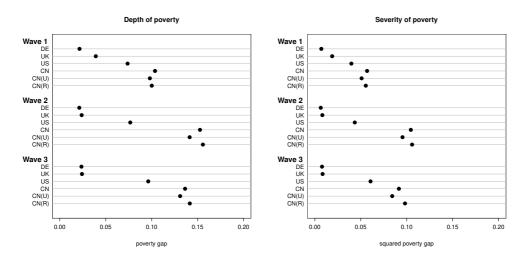


Figure 2: Poverty gap and square poverty gap by country and by wave (based on the 60%-of-median-income poverty line)

4.3Chronic vs transient poverty

So far, I have analysed the data as though they came from repeated crosssectional surveys. As a first step to exploit the panel nature of the data, I use a method proposed by Jalan and Ravallion (1998, 2000) to decompose total poverty into its chronic and transient components. They regard someone as chronically poor if her average income over the relevant period is below the poverty line; and transient poverty is defined as 'the contribution to expected poverty of the variability over time in the individual welfare indicator' (Jalan and Ravallion, 1998, p. 341).

Formally, total, chronic, and transient poverty are defined as follows. For a sample of N individuals observed over T periods, let y_{it} be the normalised income of individual i at time t, and the poverty line is set at 1. Using the squared poverty gap, the poverty experienced by i at t is:

$$p(y_{it}) = \begin{cases} (1 - y_{it})^2 & \text{if } y_{it} < 1\\ 0 & \text{if } y_{it} \ge 1. \end{cases}$$

The total poverty experienced by *i* over *T* periods is $P_i = T^{-1} \sum_t p(y_{it})$, and the total poverty for the population is $P = N^{-1} \sum_{i} P_{i}$. Now the average income of individual *i* is $\bar{y}_{i} = T^{-1} \sum_{t} y_{it}$. The chronic

poverty experienced by i is given by:

$$C_{i} = \begin{cases} (1 - \bar{y}_{i})^{2} & \text{if } \bar{y}_{i} < 1\\ 0 & \text{if } \bar{y}_{i} \ge 1. \end{cases}$$

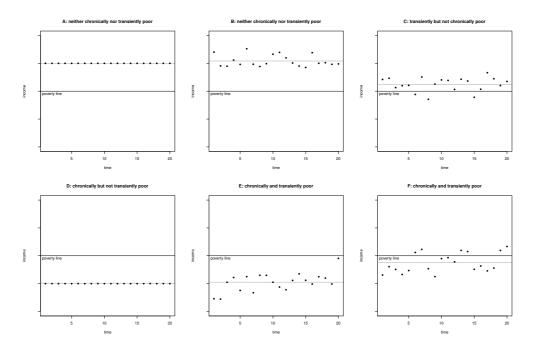


Figure 3: Hypothetical examples of income and poverty status of six individuals over time

The level of chronic poverty in the population is $C = N^{-1} \sum_i C_i$. The transient poverty experienced by *i* is the difference between total and chronic poverty, i.e. $T_i = P_i - C_i$. Similarly, the transient poverty for the whole population is defined as T = P - C.

Figure 3 illustrates the notions of chronic and transient poverty as set out above. Here we observe for 20 periods the income of six hypothetical individuals, labelled A through F. The mean income (represented by the dotted line) of A, B, and C is above the poverty line (the solid line). So there is no chronic poverty for them (shown in the top row of Figure 3). The opposite is true of D, E, and F in the bottom row. As their mean income (dotted line) is below the poverty line (solid line), they are chronically poor.

The income of A and D (the top-left and bottom-left panels respectively) is completely stable over time. A is never poor and so, by definition, never transiently poor. D is always poor. But since D's income never changes, he experiences no transient poverty either. We have already noted that E and F are chronically poor. And since their income varies over time, they experience transient poverty too, even though E's income is *always* below the poverty line. C is not chronically poor. But her income sometimes drops below the poverty line. So the only kind of poverty that C experiences is transient in

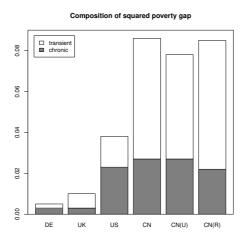


Figure 4: Total, chronic, and transient poverty by country (based on the 60%-of-median-income poverty line)

nature. Finally, although B's income varies over time, it is always above the poverty line. Thus B experiences neither transient nor chronic poverty.

Having explained the meaning of total, chronic, and transient poverty, let us go back to the data. The overall height of the bars in Figure 4 is a measure of the amount of total poverty found over three waves.⁸ Germany is at the low end (.005), followed by the UK (.010), the US (.039), and then China (.086). Thus the US has eight times as much total poverty as Germany. But the level of total poverty in China is double that in the US. This is consistent with the pattern in Figures 1 and 2.

Figure 4 also shows interesting cross-national differences in the composition of total poverty. Transient poverty accounts for less than half of the total poverty of the US (40%) or Germany (44%), but about two thirds of those of the UK (66%) or China (68%).

4.4 Poverty entry and exit

To delve deeper into the cross-national differences in the balance between chronic and transient poverty, let us consider poverty 'entry' and 'exit'. I have used quotes in the previous sentence because what the biennial panel surveys give us are snapshots of the respondents' poverty status *at the time*

⁸Figure 4 is based on the squared poverty gap index. I have repeated the analysis using the poverty gap and the poverty headcount indices. They give very similar results to those reported here. Details are available on request.

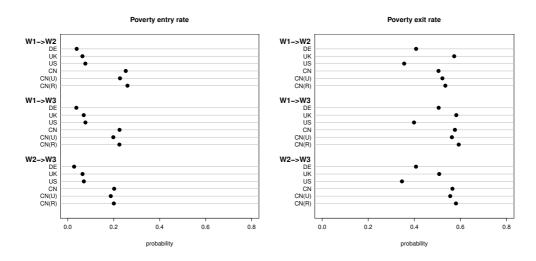


Figure 5: Poverty entry and exit probabilities (based on the 60%-of-medianincome poverty line)

of the interviews, not what happened in-between. Thus someone who was found to be poor in both waves 1 and 2 might have stayed poor throughout the two years between the interviews. Or, he might have escaped poverty and then fallen back into it (i.e. a case of recurrent poverty).⁹ It is because of this limitation of the data (and also because the Chinese panel is still very short) that I refrain from carrying out event history analyses of poverty entry and exit (Bane and Ellwood, 1986).

Having stated this caveat, it is useful to consider the following questions. Among people who were not poor at, say, wave 1, what fraction was poor at wave 2? And among those who were poor at wave 1, what fraction were not poor at wave 2? I use poverty entry rate and poverty exit rate as shorthands to refer to these fractions.¹⁰

The left panel of Figure 5 shows that, between waves 1 and 2, the poverty entry rates are 4 percent for Germany, 6 percent for the UK, 8 percent for the US, and, remarkably, 25 percent for China. The right panel reports the poverty exit rates, which are 41 percent for Germany, 57 percent for the UK, 36 percent for the US, and 51 percent for China. Very similar poverty entry and exit rates are observed if we were to compare the poverty status

⁹Analysing *annual* panel data from waves 1–16 of the British Household Panel Survey, Jenkins (2011, p. 229) reports that '[a]mong a large group of people starting a poverty spell, 45 percent leave after one year ... the median [poverty] spell length ... is between one and two years.'

¹⁰Strictly speaking, I should be speaking of probabilities rather than rates. I use the term rates to make the text less clunky.

of individuals between waves 1 and 3 or between waves 2 or 3.

Taken together, these entry and exit rates help elucidate the results of Figure 4. For example, China has, by some distance, the highest poverty entry rate *and* either the highest or the second highest poverty exit rate. This means that, by international standards, the typical Chinese face quite high risks of becoming poor, but they tend not to stay poor for very long. This is consistent with the observation that total poverty in China is at a very high level, and much of it is transient in nature. In sharp contrast to China, Germany has the lowest poverty entry rate *and* the second lowest poverty exit rate. This is in line with what we saw in Figure 4 where Germany has the lowest level of total poverty, of which just over half is chronic.

We saw in Figure 4 that the level of total poverty in the UK is relatively low (when compared to China or the US) and two-thirds of it is transient. This makes sense given the pattern in Figure 5 where the UK's poverty entry rate is closer to Germany's, and its poverty exit rate is comparable to China's. Finally, the poverty entry rate of the US is slightly higher than those of the UK and Germany, and it has the lowest poverty exit rate of the four countries. Again, this matches with the pattern in Figure 4 where total poverty in the US is high, and three-fifths of which is chronic.

The entry and exit rates of Figure 5 are also relevant to assessing previous cross-national research on poverty dynamics. Recall Duncan *et al.*'s (1993) observation that countries with higher poverty level tend to have lower poverty exit rates (cell 1 of Table 1) and higher poverty entry rates (cell 2). They also propose a simple mechanism that explains these associations. Specifically, the more poverty there is, the greater the likely distance between the typical poor person and the poverty line (cell 3), and this would explain the lower escape rate. Conversely, where there is less poverty, the typical non-poor person would be further above the poverty line (cell 4), which is why she is less likely to fall into poverty.

The left panel of Figure 5 shows that Germany has the lowest poverty entry rate, followed by the UK, the US, and China. As the poverty levels of the four countries follow the same ranking (see Figures 1 and 2), this result is consistent with Duncan *et al.* (1993).

As regards escaping poverty, the right panel of Figure 5 shows that, between waves 1 and 2, the US has the lowest poverty exit rates, followed by Germany, China, and the UK.¹¹ The pattern for the three Western countries is arguably consistent with Duncan *et al.* (1993). Compared to the US, there

¹¹This rank order also holds for the transitions between waves 1 and 3 and between waves 2 or 3, with the exception that the poverty exit rates for China are higher than the UK's in these transitions.

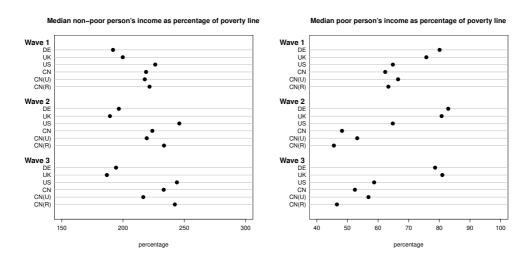


Figure 6: Median non-poor person's (left panel) and median poor person's (right panel) income as a percentage of the poverty line (based on the 60%-of-median-income poverty line)

is less relative poverty in Germany or the UK, and it is also easier to escape poverty in the two European countries.¹² But when we bring in China, the picture looks very different. Although relative poverty is at a very high level in China, poverty exit rates in China are the highest or the second highest of the four countries.

Turning to the proposed mechanism that links relative poverty levels to the poverty entry and exit rates. The left panel of Figure 6 reports the median income of the non-poor of each of the four countries, expressed as a percentage of that country's poverty line. In wave 1, they are 192 percent in Germany, 200 percent in the UK, 226 percent in the US, and 219 percent in China. Thus the non-poor in China or the US are further above the poverty line than their counterparts in Germany or the UK. But, as we have seen, Americans and, especially, Chinese are at greater risks of falling into poverty than are Germans or Brits. This is not consistent with Duncan *et al.*'s (1993) suggestion that 'the greater ... the likely distance between the typical nonpoor family and the poverty threshold and the less likely that family is to make the transition.'

The right panel of Figure 6 reports the median income of the poor of the four countries, again expressed as a percentage of each country's poverty line. In wave 1, they are 80 percent for Germany, 76 percent for the UK, 65

 $^{^{12}\}mathrm{An}$ anomaly here is that the UK has very high poverty exit rates. I will return to this in Section 5.

percent for the US, and 62 percent for China. Thus, consistent with Duncan *et al.* (1993), the typical poor person of the two countries with less poverty (i.e. Germany and the UK) are indeed closer to the poverty line than their counterparts in the countries with more poverty (i.e. China and the US). But, contra Duncan *et al.* (1993), proximity to poverty line does not always imply higher escape rate, e.g. compare Germany with China. Overall, it seems clear that poverty entry (exit) rate is not simply a function of how far the non-poor (the poor) are from the poverty threshold.

4.5 Occasional and recurrent poverty

A third way to exploit the panel nature of the data is to count how many times someone is poor over three waves. Figure 7 reports the cumulative experience of poverty: 87 percent of Germans, 78 percent of Brits, and 71 percent of Americans were never poor over three waves. The corresponding figure for China was only 46 percent. Comparing Figure 7 with Figure 1, it is clear that in all four countries poverty is a more common experience when we take a longitudinal perspective. This is consistent with both expectation and past research (Layte and Whelan, 2003). But the stark contrast shown in Figure 7 is that while only a minority of individuals in the West were touched by poverty over three waves, being poor was actually the experience of a small majority of Chinese.

For brevity, I refer to being poor once over three waves as occasional poverty and being poor twice or thrice as recurrent poverty. With these shorthands, Figure 7 shows that 7 percent of Germans, 12 percent of Brits, 12 percent of Americans, and 30 percent of Chinese were occasionally poor; and 6 percent of Germans, 9 percent of Brits, 18 percent of Americans, and 24 percent of Chinese were recurrently poor. Thus the experience of both occasional poverty and recurrent poverty are, by quite a large margin, more common in China.

4.6 Determinants of occasional or recurrent poverty

To find out the determinants of occasional or recurrent poverty, I fit a multinomial logit model to the data, with the threefold distinction of never poor, occasionally poor, and recurrently poor as the dependent variable. As regards the independent variables, they include the respondents' age (at wave 1), sex, family structure (at wave 1, four categories), and urban (vs rural) residence, and three key predictors of income: educational attainment (four categories), employment status, and the additional-earner-to-household-size

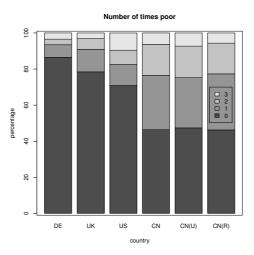


Figure 7: Cumulative experience of poverty over three waves by country (based on the 60%-of-median-income poverty line)

ratio.¹³ Finally, for China only, I include *hukou* as a predictor of poverty. All but two of these variables, as constructed, are time-constant in nature. The exceptions are employment status and the additional-earner-to-household-size ratio. For these two variables, I take the mean of their values over three waves.¹⁴ Descriptive statistics of the variables can be found in Table A1 in the online appendix.

The results are reported in Table 3. Overall, there is a good deal of crossnational similarity in the determinants of poverty. For example, age (within the age range of 31 to 55) and gender tend not to predict poverty status. The main exception is the US where men are less likely to be poor.

People with more education are everywhere less likely to be poor, except for the contrast between occasionally poor and never poor in Germany. Similarly, in all countries, having more earners in the household and being employed protect against poverty. But it is notable that the magnitude of these employment-related parameters are considerably smaller in China. Indeed, when urban and rural China are examined separately, being employed only protects against occasional poverty in urban China.

As regards family structure, compared to the reference category of couples without children, single-parents are more likely to be poor; while couples with

¹³That is, the ratio of the number of earners in the household (not counting the respondents) to the household size.

¹⁴Employment status at each wave is a binary variable, coded 1 for being employed or self-employed, and 0 otherwise. The covariate is the mean of this binary variable over three waves.

children are less likely to be occasionally poor in the US, and they are less likely to be recurrently poor in the UK and China.

In China, Germany, and the US, but not in the UK, people living in rural areas are more likely to be poor: occasionally in Germany, recurrently in the US, and both occasionally and recurrently in China. As expected, Chinese with urban *hukou* are less likely to be poor.

But the most remarkable result of Table 3 is that the pseudo- R^2 of the multinomial logit model is much smaller for China (8 percent) than for the other three countries (18 to 28 percent). Thus while many key determinants of poverty work in similar ways in the four countries, being poor is a much more unpredictable experience for Chinese.

Given the large regional difference in living standards in China, I have, in a further model (not shown), controlled for regions, i.e. the Bundesländer in Germany, the government regions in the UK, the states in the US, and the provinces in China.¹⁵ It turns out that adding region to the model increases the pseudo- R^2 slightly: from less than 1 percent in the UK to 3 percent in China. Thus even with regional differences accounted for, there is still a large gap in the variance explained between China and the three Western countries.

¹⁵Details are available on request.

	Poor once over three waves vs never poor						
	Germany	UK	US	China	China (urban)	China (rural)	
age	.022 .014	.001 .005	008 .010	004 .004	011 .007	.013* .006	
male	.050 .205	097 .071	279* .141	.032 .063	.010 .096	.021 .080	
junior high ^{a}	248 .301	263 * *.095	652 $.407$	355 * * .068	371** .114	564 * *.087	
senior high	.411 .443	504 * *.134	-1.294 * * .323	581 * *.097	699 * * .136	316* .139	
postsecondary	459 345	803 * *.100	-2.047 * * .319	-1.764 * * .201	-1.745 * * .205	-1.924 * *.523	
mean $\# \text{ emp/hh}$	-3.134 * * .496	-3.083 * * .236	-3.691 * * .481	-1.537 * * .223	-1.795 * * .333	-1.360 * *.294	
mean employed	-2.166 * * .244	-1.144 **.088	-1.851 * *.225	364 * * .112	411 * * .149	.223 .167	
single $parent^b$.997**.360	.523**.133	.571 .310	.703* .336	.714 .476	.788 .459	
couple w kids	119 $.272$	160 .100	464* .219	180 .140	040 .199	342 .191	
others	.316 .310	150 .131	.228 .250	087 .141	.074 .202	159 .191	
urban	811 * *.198	083 .083	278 .174	171* .068			
hukou (urban)				658 * *.091	522 * * .105	630 * *.204	
constant	565 .794	.194 .276	3.086 * * .686	1.237 * * .296	1.545 * * .427	286 $.375$	
		Poor twic	e or thrice over	three waves vs r	never poor		
age	.024 .016	.009 .006	010 .009	006 .005	005 .008	004 .007	
male	118 $.256$	028 .096	412 * *.156	.108 .070	008 .108	.170 .093	
junior high ^{a}	127 $.298$	324 * *.115	-1.049 * *.368	496 * *.076	513 * * .121	516 * *.101	
senior high	-1.183* .471	522 * *.165	-2.372 * *.323	787 * *.118	855 ** .156	642 * *.187	
postsecondary	-1.554 **.356	-1.044 * * .126	-3.481 * * .318	-2.881 * * .382	-2.801 * * .361	-1.903 * *.703	
mean # emp/hh	-6.827 * * .821	-4.583 * * .335	-8.891 * *.615	-3.264 * * .257	-3.333 * * .367	-3.189 * * .359	
mean employed	-3.822 * *.317	-2.075 * *.107	-3.174 * * .239	324 * *.120	297 .163	051 .180	
single parent ^{b}	2.115 * * .456	.779 * * .165	1.234 * * .328	.473 .310	1.243** .449	.401 .432	
couple w kids	061 .384	464 * *.150	.122 .271	763 * *.146	.003 .220	-1.134 * * .198	
others	1.473 * * .374	.120 $.157$.748* $.289$	617 * *.147	169 .225	960 * *.197	
urban	363 .248	016 .106	630 * *.173	491 **.078			
hukou (urban)				-1.021 * * .114	-1.088 * * .122	-1.336 * * .332	
constant	082 .837	.224 .362	6.007 * *.693	2.270 * *.319	1.885** .484	1.629 * * .416	
N	5,122	10,611	$3,\!659$	11,371	4,652	6,212	
R^2	0.273	0.175	0.279	0.076	0.088	0.034	

Table 3: Multinomial logit model predicting occasionally poor and recurrently poor v never poor (based on the 60%-of-median-income poverty line)

* p < .05, ** p < .01; reference category: ^a primary education, ^b couples without children

5 Summary and discussion

China has made some truly significant progress in tackling absolute poverty. But relative poverty is a different story altogether. Compared to Germany, the UK, and the US, not only is relative poverty more common in China, it is also deeper and more severe. Furthermore, while absolute poverty in China is 'predominantly a rural phenomenon', relative poverty afflicts Chinese cities as much as it does the countryside.

By international standards, a very large share of China's relative poverty is transient in nature. (Transient poverty is also important in the UK. We will come back to this point later.) Direct measures of poverty entry and exit rates also suggest greater instability of people's economic fortune in China. Compared to their Western counterparts, the average Chinese faces significantly higher risks of falling into relative poverty. But the relatively poor in China are also more likely to escape poverty than the poor in Germany or the US.

The combination of high poverty entry rate and high poverty exit rate means that there are more occasional poverty *and* more recurrent poverty in China than in the West. Indeed, while a minority of people in Germany, the UK, and the US were touched by relative poverty over three waves (four years), being poor was actually the experience of a small majority of Chinese.

Relative poverty in China is distinctive in many ways. But we should not regard China as sui generis, as the covariates predicting poverty status are very similar across the four countries. For example, higher educational qualifications, employment, and having more earners in the household all protect against poverty in the four countries. But collectively these covariates are much less predictive of poverty status in China than in Germany, the UK and the US.

The findings of this paper also speak to our understanding of poverty dynamics in general. Duncan *et al.* (1993) posit that countries with more poverty tend to have higher poverty entry rates, and that is because the nonpoor in those countries tend to be closer to the poverty line. We do observe higher poverty entry rates in countries with more poverty. But the non-poor of those countries (e.g. China and the US) are actually further away from the poverty line than their counterparts in countries with less poverty (e.g. Germany and the UK).

Furthermore, Duncan *et al.* (1993) argue that poverty exit rates would be lower in countries with more poverty, as the typical poor person in those countries would be further below the poverty line. Once again, this is only partly true. The median poor person in China or the US are indeed further away from the poverty lines than the median poor person in Germany or the UK. But poverty exit rates are much higher in China or the UK than in Germany or the US.

Some revisions to Duncan *et al.* (1993) is therefore necessary. Mechanically speaking, it is a large income change that pushes people into poverty or plucks them out of it. What counts as a large income change depends partly on how far someone is from the poverty threshold. This is why Duncan *et al.*'s (1993) claims are eminently reasonable. But our findings show that this is not the whole story. Other social forces operating in the labour market, in households, and in the broader policy environment also make large income changes more likely to occur in some countries than in others.

For example, using the same data sets, Chan *et al.* (2019) show that not only is *permanent income* more unequally distributed in China than in the West, there is greater instability in *transitory income* in China too. They decompose the total variance of log-income in the panel data into its between and within components, and report that 'in Germany, the United Kingdom, and the United States the lion's share of income inequality is found between individuals rather than within individuals. But the opposite is true for China, suggesting that the average Chinese face a much higher degree of income instability and uncertainty' (Chan *et al.*, 2019, p. 443). Moreover, while there is regression to the mean in income in all four countries, 'the magnitude of income change is always larger in China than in the three Western countries' (Chan *et al.*, 2019, p. 439). They then argue that broader changes in the labour market and employment practices in China have created a large pool of casually employed workers which, in turn, explains the heightened income instability in China (Gallagher, 2005; Kuruvilla *et al.*, 2011).

Finally, the implications of the findings of this paper go beyond China. In comparative research on welfare capitalism, the UK and the US are often grouped together as exemplars of the liberal regime. Similarly, Atkinson (2015, p. 20) notes that, so far as inequality is concerned, 'the situation in the UK is a pale imitation of what is happening in the US, and that the UK chart can be obtained by simply replacing "S" by "K" in the heading. There is some truth in this.' But the poverty dynamics of the two countries is actually quite different. In particular, rates of poverty exit are much higher in the UK than in the US (see also Valletta, 2006; Office For National Statistics, 2015; Vaalavuo, 2015). This means that there is much less chronic poverty in the UK. This is a particularly intriguing outcome many of the antipoverty policies brought in by the New Labour government (1997–2010), e.g. tax credits, were 'strongly influenced by evidence from US welfare-to-work experiments and, again, had many elements common with them' (Waldfogel, 2010, p. 5). Further work on how the social structure and/or anti-poverty programmes of the two countries had led to such different poverty dynamics

would be very illuminating.

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A Supplementary figures and tables

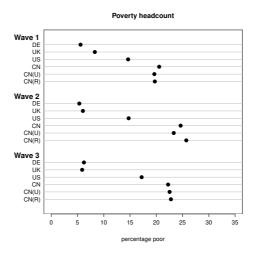


Figure A1: Headcount poverty rate by country and by wave (based on the 50%-of-median-income poverty line)

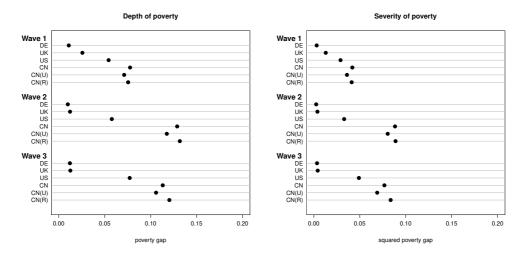


Figure A2: Poverty gap and square poverty gap by country and by wave (based on the 50%-of-median-income poverty line)

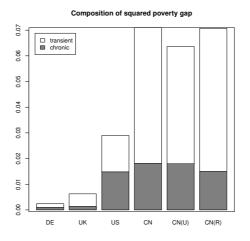


Figure A3: Total, chronic, and transient poverty by country (based on the 50%-of-median-income poverty line)

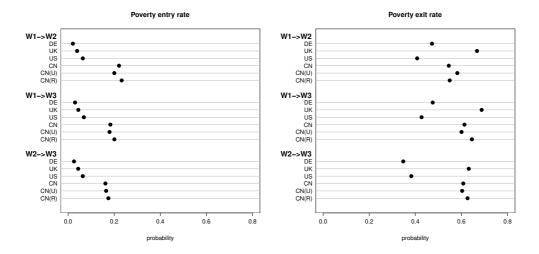


Figure A4: Poverty entry and exit probabilities (based on the 50%-of-medianincome poverty line)

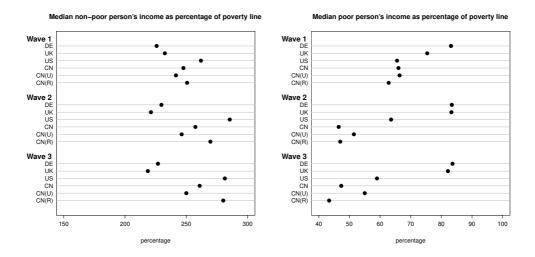


Figure A5: Median non-poor person's (left panel) and median poor person's (right panel) income as a percentage of the poverty line (based on the 50%-of-median-income poverty line)

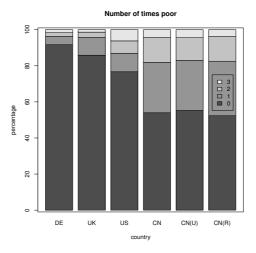


Figure A6: Cumulative experience of poverty over three waves by country (based on the 50%-of-median-income poverty line)

	DE	UK	US	CN	CN(U)	CN(R)
male	47.7	48.9	47.5	50.2	50.1	50.4
primary	8.5	15.4	3.1	44.9	29.7	59.0
junior high	52.4	34.1	6.1	34.3	36.6	31.4
senior high	7.6	9.5	32.4	14.9	22.0	8.5
postsecondary	31.6	41.0	58.5	6.0	11.8	1.1
single parent	7.2	10.3	9.2	1.5	1.8	1.2
couple w/o kids	23.3	19.0	17.9	4.9	5.2	4.7
couple w kids	51.1	56.6	49.0	53.2	56.7	51.2
others	18.4	14.1	23.9	40.5	36.3	42.9
urban	66.9	78.6	82.2	47.1		
hukou (urban)				26.0	50.9	4.3
age (mean)	44.04	43.24	43.40	42.73	42.42	43.18
age (sd)	6.83	6.92	7.18	6.80	6.77	6.78
employed (mean)	0.85	0.82	0.85	0.81	0.76	0.85
employed (sd)	0.31	0.35	0.31	0.29	0.34	0.24
# earn/hh (mean)	0.22	0.24	0.24	0.29	0.27	0.30
# earn/hh (sd)	0.20	0.19	0.19	0.14	0.15	0.14

Table A1: Descriptive statistics (percentages, except for age, mean of being employed, and mean of earners to household size ratio)

Poor once over three waves vs never poor						
	Germany	UK	US	China	China (urban)	China (rural)
age	005 .015	.004 .005	017 .010	.004 .004	014* .007	.006 .005
male	.133 .261	029 .079	156 .150	.020 .062	.025 .095	014 .079
junior $high^a$	044 .349	213* .102	719 $.397$	372 * *.068	257* .109	477 * *.086
senior high	825 .686	373* .144	-1.298 * * .328	469 * *.097	697 * * .136	269 .138
postsecondary	-1.092 * * .416	691 * *.108	-1.874 * * .325	-1.855 * * .215	-1.795 ** .226	-2.138 * *.566
mean # emp/hh	-3.126 * * .685	-2.885 * * .264	-4.229 * *.516	-1.711 * *.221	-1.852 * * .327	-1.262 * * .289
mean employed	-2.594 * * .285	-1.238 * * .090	-2.033 * * .219	215 .110	380* .148	.240 .163
single parent ^{b}	1.521** .456	.361* .145	.378 .308	.754* .307	.516 .434	.862* .411
couple w kids	433 .346	226* .114	573* .237	324* .135	.008 .199	522 * *.183
others	.874* .361	066 .139	.201 .258	238 .136	.017 .203	306 .183
urban	308 .248	111 .094	209 .186	147* .067		
hukou (urban)				698 * * .092	653 * * .106	525* .206
constant	052 .790	370 .310	3.207**.710	.651* .285	1.449** .429	058 $.366$
		Poor twic	e or thrice over	three waves vs n	ever poor	
age	001 .021	.001 .008	013 .010	001 .005	.001 .008	008 .007
male	067 .324	207 .132	241 .173	.127 .074	.017 .120	.202* .101
junior high ^{a}	701 .397	288 .149	-1.210 * *.367	442 * *.080	515 * * .130	472 * *.110
senior high	-1.558 * * .593		-2.545 * * .341	774 * * .133	841 * * .175	659 * *.212
postsecondary	-1.808 * * .464	827 * *.173	-3.424 * * .333	-2.847 * *.514	-2.974 * * .451	-1.541*.706
mean # emp/hh	-6.819 * * 1.009	-4.806 * *.531	-8.713 * *.718	-3.028 * * .269	-3.302 * * .409	-3.388 * * .394
mean employed	-4.334 * * .372	-1.926 * *.145	-3.151 * * .245	208 $.124$	294 .169	051 .191
single $parent^b$	1.019* .508	.514* .207	.759* .366	.475 .294	.756 .401	044 .461
couple w kids	-1.038* .503	563 * *.204	238 .307	804 * *.151	329 .228	-1.206 * *.208
others	1.122* .463	.286 .191	.781* .322	684 * * .151	360 .231	999 * *.205
urban	614* .302	146 .142	695 * *.184	489 * *.084		
hukou (urban)				-1.043 * * .130	941 * * .134	-1.332 * * .387
constant	1.467 1.111	445 .478	5.708 * * .746	1.465 * * .334	1.152* .509	1.513 * *.448
Ν	5,122	10,611	$3,\!659$	11,371	4,652	6,212
R^2	0.326	0.155	0.280	0.071	0.083	0.033

Table A2: Multinomial logit model predicting occasionally poor and recurrently poor (v never poor), 50%-of-medianincome as poverty line

* p < .05, ** p < .01; reference category: ^a primary education, ^b couples without children