



## **Distributional National Accounts for Australia, 1991-2018**

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Matthew Fisher-Post, Nicolas Hérault and Roger Wilkins

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## **Abstract**

We produce estimates of the distribution of the National Accounts measure of national income in Australia for the period 1991 to 2018 by combining household survey data and tax records data, and adjusting individual incomes so as to sum to National Accounts aggregates. We document the shifts in income shares over the period, contrasting changes in the distribution of pre-tax and post-tax national income. We compare Australia to the US and France, and compare our new inequality estimates with traditional household survey based estimates. We additionally exploit the extra richness of survey data to shed further light on the distribution of national income across and within various population groups not usually identifiable in the tax data the tax data on which Distributional National Accounts are generally based.

**Key words:** income inequality, national accounts.

## 1. Introduction

A recent literature led by researchers affiliated with the World Inequality Database (Atkinson and Morelli 2018; Bozio et al. 2018; Garbinti et al. 2018; Piketty et al. 2018, 2020; Piketty et al. 2019) has attempted to provide a more complete picture of the distribution of income through allocating all of the income as measured in National Accounts to individual members of society. The guiding principle for these ‘Distributional National Accounts’ is to allocate the entirety of national income to individuals in line with their ‘beneficial receipt’ of the income—that is, according to how much of the income effectively accrues to them.

By doing so, a more accurate picture of the distribution of income is possible compared with traditional inequality studies using household survey or tax records data, which typically only capture cash incomes, thereby missing important components such as in-kind benefits from government-provided goods and services, imputed rents on owner-occupied housing, and retained earnings of companies. By accounting for these additional income components, the Distributional National Accounts approach therefore generates estimates of individuals’ incomes that are on average larger than obtained from household surveys or income tax data and which should more accurately reflect the distribution of all (cash and in-kind) income.

In this paper we attempt to produce statistics on the distribution of income in Australia as measured by the National Accounts.<sup>1</sup> Our approach is guided by Alvaredo et al. (2020), which details the income concepts and methods of implementation adopted by the World Inequality Database (WID). The guidelines are, however, not completely prescriptive because of the substantial variation across countries in institutional features and data availability. Our approach is therefore considerably influenced by the particular institutional features of Australia and the nature of the available data, including the relative strengths and weaknesses of alternative data sources.

Four main national income concepts are identified in Alvaredo et al. (2020) as being of interest: pre-tax factor income; pre-tax post-replacement income; post-tax disposable income; and post-tax national income. Pre-tax factor income approximately corresponds to total income accruing to capital and labour, where all of national income is attributed to capital and labour. Pre-tax post-replacement income is the same as pre-tax factor income, but with an adjustment made to account for the public pension system by allocating pension payments to recipients and deducting the contributions used to fund them (such that it still sums to national income). Post-tax disposable income deducts all taxes attributable to individuals and adds cash transfers. Consistent with the principle of distributed income aggregating to National Accounts totals, the total value of taxes

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<sup>1</sup> This paper builds on earlier work by Fisher-Post (2020).

deducted equals the total value of taxes collected by government (not just income taxes). However, government expenditure is not allocated to individuals and thus the sum of post-tax disposable income is less than national income. Post-tax national income addresses this deficiency by distributing all of government expenditure, inclusive of items not readily attributable to individuals, such as national defence.

We construct measures of all four income concepts, but the results we present are primarily for pre-tax post-replacement income and post-tax national income on the basis that these are the main pre-tax and post-tax income concepts of interest, respectively corresponding to measures of the distributions of market income and ‘post-government’ income (the latter corresponding to ‘beneficial receipt’ of income).

We are not the first to attempt to describe the distribution of income in Australia adopting a National Accounts income concept. The Australian Bureau of Statistics (ABS) has, on four occasions since 2014, released distributional information on national income by combining information from its biannual household income survey with the National Accounts data (most recently in 2021; see ABS 2021a). The methods have been refined over time. In the most recent release, for each of nine years between 2003-04 and 2019-20, statistics are presented on the distribution of various components of the national household income account across households.

While complementary to the analysis we undertake, the ABS approach is somewhat different to that advocated by Alvaredo et al. (2020). The income concept differs slightly because of the failure to deduct consumption of fixed capital (hence, it is a National Accounts concept of *gross* income that is distributed to households).<sup>2</sup> More importantly, the distributional information produced by the ABS is limited, presenting only the total, mean and share of each income component of the household income account for broad groupings of households: by main source of income (five groups), by equivalised income quintile, by household type (seven groups), by age group of the household ‘reference’ person (six groups) and by wealth quintile.

Compared with the ABS outputs, we therefore present distributional information that is based on income concepts more in line with the WID guidelines, and which is more detailed and better suited to international comparisons. Furthermore, we present distributional information for a larger array of demographic groups than is produced by the ABS.

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<sup>2</sup> In practice, this is not an especially significant issue. For example, in 2017-18, total gross income in the household income account was \$1.53 trillion and (net) national income was \$1.47 trillion (ABS, 2021b). (Dollar amounts refer to Australian dollars throughout this paper.)

## 2. Distributing national income to individuals

In building the Distributional National Accounts (DINA) for Australia, we follow approaches taken to produce DINA estimates for France and the US (Garbinti et al. 2018; Piketty et al. 2018) as well as the Distributional National Accounts Guidelines (Alvaredo et al. 2020).

The goal is to distribute to individuals all of the National Accounts measure of income, defined as GDP plus net foreign income minus consumption of fixed capital. Following the DINA Guidelines, we construct four measures of income that are distributed to individuals, although only three of these sum to a National Accounts aggregate. In the following we describe the methods and data used to produce each income distribution.

### 2.1. Pre-tax factor income

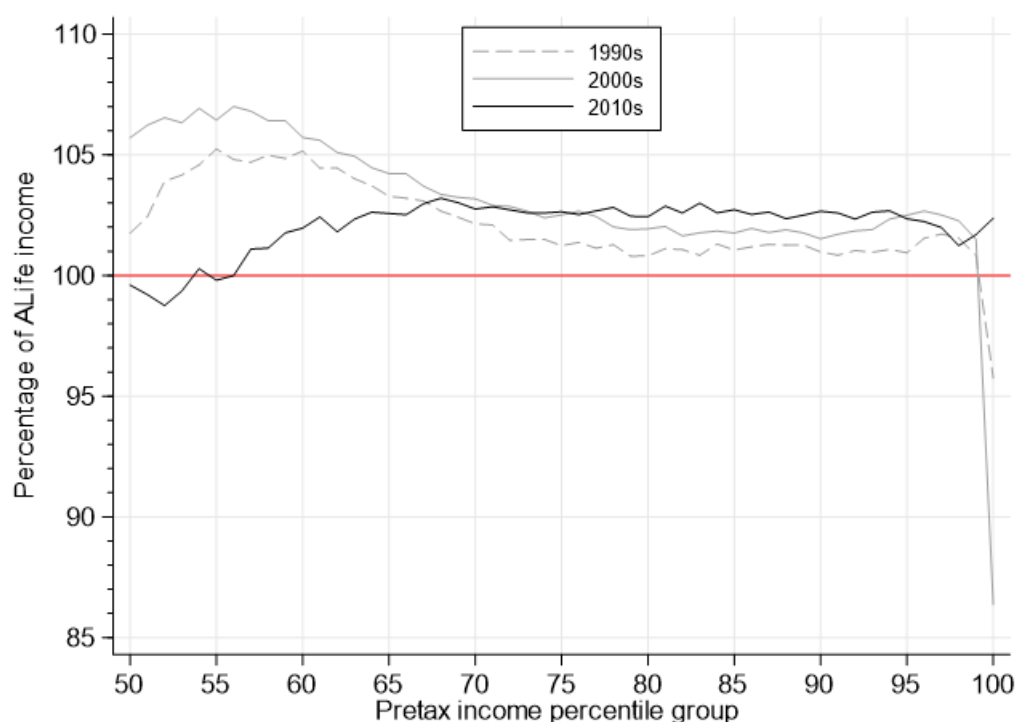
#### 2.1.1. *Pre-tax cash incomes*

For pre-tax cash incomes of individuals, based on exploratory work with both ALife (a 10% random sample of tax returns currently covering 1991-2018) and the Australian Bureau of Statistics' Survey of Income and Housing (SIH, covering 1994 to 2018 but with some gaps), we determined that the best approach was to primarily base cash income estimates on the SIH, but with ALife tax data used to adjust incomes for the top 1%. This is because the tax data appear inferior in income capture for most of the distribution (see Figure 2.1). Although non-labour income is higher in ALife than in the SIH for people with above-median incomes (see Figure 2.3), it is not enough to compensate for the undercoverage of labour income evident in Figure 2.2.<sup>3</sup>

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<sup>3</sup> The SIH provides the longest time-span of coverage for income survey data in Australia, the main other survey source being the Household, Income and Labour Dynamics in Australia (HILDA) Survey, a panel study that commenced in 2001. The ABS has also conducted surveys that collected household income data (which is still available in unit record form) in 1975, 1982, 1986 and 1990. Unfortunately, unit-record tax data is not available prior to 1991. Further extension back to the mid 1970s may be possible using tabulations of tax data and the household survey data that is available. Prior to the mid 1970s, the only broad-based distributional information comes from income tax tables. At this stage, we leave DINA estimation prior to the 1990s as a task for future research.

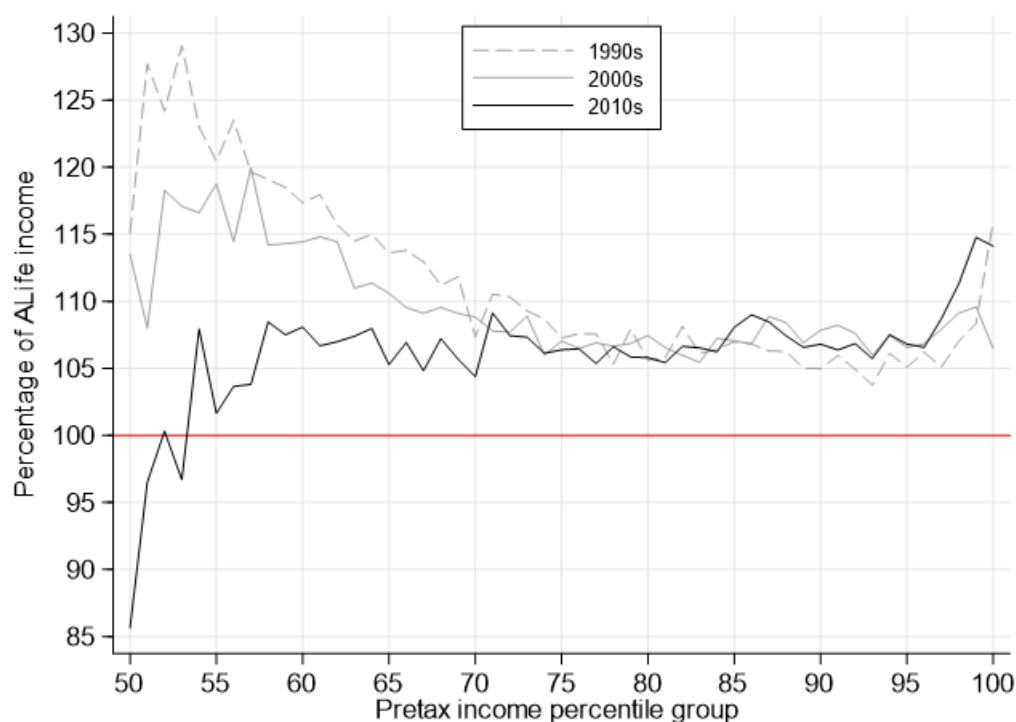
**Figure 2.1: SIH survey data relative to ALife tax data income by percentile – Pre-tax income**



Source: Authors' calculations based on ALife and SIH data.

Stata graph *DINAvsSIH\_check\_pretax\_deca*

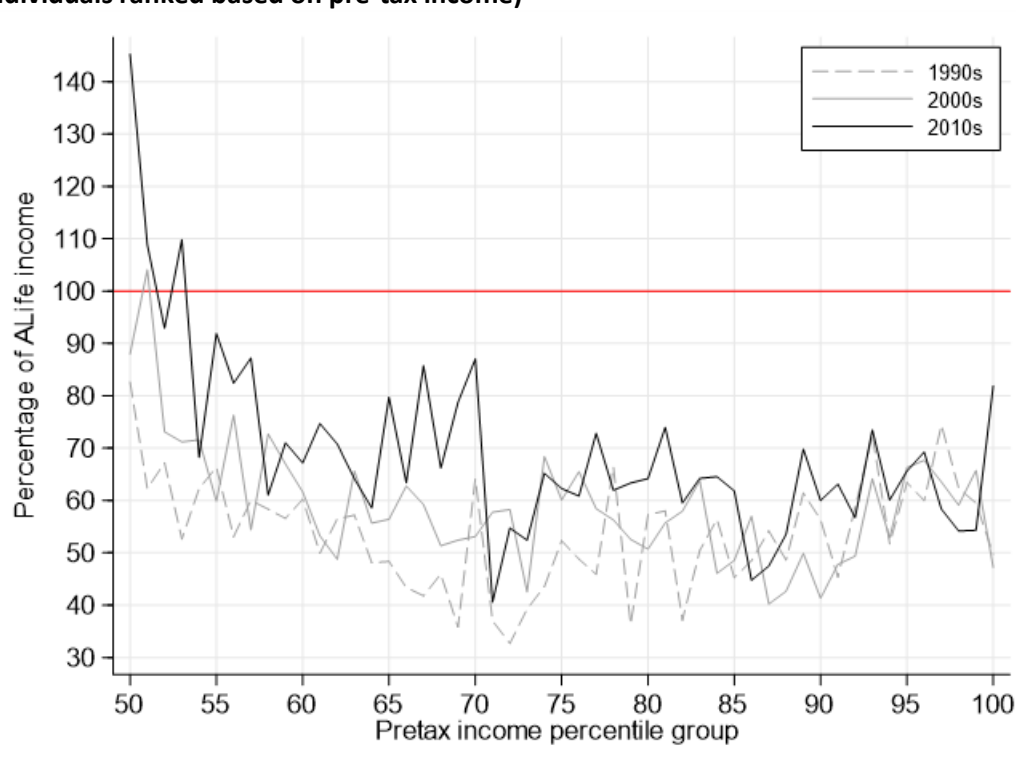
**Figure 2.2: SIH survey data relative to ALife tax data by percentile - Labour income (with individuals ranked based on pre-tax income)**



Source: Authors' calculations based on ALife and SIH data.

Stata graph *DINAvsSIH\_check\_labour\_deca*

**Figure 2.3: SIH survey data relative to ALife tax data by percentile – Non-labour income (with individuals ranked based on pre-tax income)**



Source: Authors' calculations based on ALife and SIH data.  
 Stata graph *DINAvsSIH\_check\_nonlabour\_deca*

Up until 2015-16, the SIH unit record data contain measures of both annual income (for the preceding financial year, 1 July to 30 June) and 'current weekly' income. We use the annual income estimates for these surveys. However, in the 2017-18 SIH, only current weekly income is available. We therefore use an annualised measure of this income measure for this survey.

Our approach is something of a departure from existing studies, which have given greater weight to tax records data. However, DINA need to be flexible to national circumstances, and in Australia's case, survey data is preferable to tax records data for all but the top 1%.

Australia is by no means unique in the finding that income survey data is at least as good as tax data for incomes below the top 1%. Burkhauser et al. (2012) found the US CPS matched income tax data up to the 99<sup>th</sup> percentile, and Burkhauser et al. (2018) similarly found the UK HBAI matched income tax data up to the 98<sup>th</sup> percentile. Perhaps requiring some explanation is why the survey data actually captures more income below the 99<sup>th</sup> percentile than the tax data. Two main explanations exist: some forms of income are nontaxable and are even received by high income earners; and there are incentives to minimise income reported to tax authorities that do not apply to statistical agencies. Regardless of the explanation, the fact remains that macroeconomic aggregates are

better captured when income survey data is used for the bottom 99% and tax data is only used for the top 1%.

Aside from better capture of the incomes of the bottom 99%, additional reasons to use the SIH include better flexibility to look at different income concepts (including equivalised disposable cash incomes), income units (including the household unit) as well as information on wealth. That said, we focus on the four income concepts described in the DINA Guidelines.

We distribute incomes of households on an ‘equal-split adults’ basis, meaning each adult household member is assigned an equal share of the total household income, as per the ‘broad equal-split series’ in the DINA Guidelines (p23). Although our baseline estimates are based on these broad equal-split series, we also consider two alternatives. First, we build ‘individualistic series’, which assume no sharing within households and distribute income to each person individually according to individual earnings and ownership. This is a useful comparison point with the ‘broad equal-split series’ when we further breakdown income shares by individual characteristics. Second, we build and use the ‘narrow-split series’ to ensure consistency in the comparison with the US and France. The ‘narrow-split series’ distributes income to all adult individuals by splitting income equally within a couple, but not within the extended household.

While the SIH is our preferred ‘core’ data source, it nonetheless has important limitations which need to be addressed. It is only available from 1994-95, and it has only been conducted every second year from 1997-98 to 2002-03 and from 2003-04 onwards. It also only has wealth data (and hence information on superannuation (private retirement account) balances and home equity required to distribute capital income; see below) in 2003-04, 2005-06 and 2009-10 onwards.

To produce estimates in non-SIH years, we interpolate distributions and adjust according to changes in the components of the National Accounts in those years. We use the national income price index to either inflate the distribution from the closest earlier year or to deflate it from the closest later year. If both an earlier and a later year are available, we apply both methods separately and compute the final DINA estimates by taking the average of the two series thus obtained.

#### *Top 1%: combining survey and tax data*

As a growing literature has shown, survey data tend to undercover top incomes. Comparison of survey and tax data has revealed that this is the case in Australia too (Burkhauser et al. 2016) and that it mostly affects the top 1%. We follow the cell-mean imputation method we developed for the



UK in Burkhauser et al. (2018), using tax data (ALife) to impute incomes of the top 1% in the survey data.<sup>4</sup>

To implement this method, we first rank individuals in the ALife unit record data by their 'tax gross income', which is total income subject to taxation prior to any allowable deductions or rebates. This is the closest variable to 'pre-tax income' available in the tax records data. Second, we select individuals in the top 1%, using the ABS estimate of the total adult population shown for the relevant year. Next, we allocate top 1% individuals to income groups, with the size of each group equal to  $1/100,000^{\text{th}}$  of the total adult population, meaning we split the top 1% into 1,000 income groups. Third, we calculate the average income for each income group. Next, we repeat the first and second steps with the SIH data for the same year using our derived measure of individual gross income. We then duplicate each record according to its sample weight. Finally, for each of the 1,000 SIH income groups within the top 1%, we replace the individual-level SIH incomes with the mean income of the corresponding group in ALife.

In addition to imputing gross income from tax data for the top 1%, we also use the labour/capital income-source composition as is obtained from the tax data. An alternative assumption would be to use the income composition as determined by the survey data, but this tends to underestimate the importance of capital income for the top 1%. However, the tax data offer less detail and thus less flexibility in then adjusting incomes to match National Accounts totals (e.g., mixed-income is not directly observable in ALife). We address this issue by maintaining the assumption that the income-source compositions of capital and labour incomes are as obtained from the survey data.

Our procedure ensures that total 'tax gross income' for the top 1% – and for each of the 1,000 groups within the top 1% – is the same in the (adjusted) SIH and ALife data.

### *2.1.2. Labour income*

Grossing up of labour incomes is required because of potential under-reporting in SIH as well as the failure of the SIH to capture (all of) salary sacrificed employment income, fringe benefits and fringe benefits tax, and 'employer social contributions' (i.e., employers' superannuation contributions and workers' compensation premiums). Employee incomes are grossed up by a constant factor so that total employee income in the SIH equals total employment income in the National Accounts.

Mixed income is grossed up separately, also by a constant factor. ABS National Accounts data do not report *net* mixed income. We therefore estimate net mixed income based on *gross* mixed income,

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<sup>4</sup> A few preliminary adjustments to ALife data are required: see Appendix A.1.

which is reported in the National Accounts data, by applying net-to-gross ratios for mixed income sourced from the WID for Australia.<sup>5</sup>

All grossing-up factors are provided in Appendix Table A.2. Total employee incomes have to be increased by between 10% and 26% to ensure consistency with National Accounts. The required increase is much larger and more volatile from year to year for mixed income, ranging from 25% to 226%, depending on the year.

### *2.1.3. Capital income*

Capital income is estimated based on reported business and investment income and imputed rent. A ‘grossing up’ adjustment is done separately for each of superannuation, imputed rent and other capital income. The principle is that superannuation income is imputed based on observed or estimated superannuation balances. Net operating surplus of households and non-profit institutions serving households (NOSHN) is distributed based on imputed rent. The remaining (i.e., non-pension non-imputed-rent) capital incomes not captured by the SIH are distributed according to reported non-pension non-imputed-rent capital incomes (hereafter called ‘other capital income’).

From the total capital stock (“National net wealth”) as measured in the National Accounts, we compute the share of the capital stock in superannuation funds (“Pension funds & life insurance”) and then use that share to allocate the appropriate proportion of total private capital income (other than NOSHN) accruing to superannuation funds. The implicit assumption is that returns on superannuation are the same as the overall return on the national private capital stock. Total private capital income is obtained here from the National Accounts by adding “total net property income of households and non-profit institutions serving households” and “total net primary income of corporations”.

Superannuation income, NOSHN and other capital incomes are thus allocated to each individual separately.

#### *Superannuation income*

We impute superannuation income proportionally to each individual’s superannuation balance. We use superannuation balances from the SIH for all years for which they are available (2003/04, 2005/06, 2009/10, 2011/12, 2013/14, 2015/16 and 2017/2018). For the years not covered by the

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<sup>5</sup> Alvaredo et al. (2020) discuss the disaggregation of total depreciation (consumption of fixed capital, CFC) where its components are missing in the official National Accounts statistics: A share of the total CFC corresponds to each component of gross operating surplus in the gross domestic product (viz. gross corporate operating surplus, gross mixed income, gross household operating surplus, and gross government operating surplus). Refer to <https://wid.world/> for data and methodology.

SIH, we estimate superannuation balances separately for those aged 60 and over and those aged under 60.

For those aged under 60, we estimate a regression model of superannuation balances on age, labour income and sex (as well as interactions). For those aged 60 and over, the model is enriched by including superannuation income. The coefficient estimates (see Appendix A.3) are then used to impute superannuation balances in the SIH data for years with no information, by using the set of estimated coefficients from the closest year available. This means that superannuation balances from 1991 to 2002 are all estimated based on the 2003 model. This approach is likely to generate some prediction errors. However, we note that superannuation wealth was limited in the 1990s, since compulsory contributions only commenced in 1992, initially at only 3% of gross earnings and gradually increased up to 9% as of 1 July 2002.<sup>6</sup> Moreover, it is the relative distribution of superannuation balances that matters for imputation and not the absolute values, and relativities by labour income, age and sex are likely to have remained relatively stable between 1991 and 2003.

*Net operating surplus of households and non-profit institutions serving households (NOSHN)*

ABS National Accounts data report only gross and not net operating surplus of households and non-profit institutions serving households. We use the share of the consumption of fixed capital attributable to operating surplus in NOSHN from the WID Australian National Accounts data (see footnote 5 above) to derive net operating surplus from the ABS National Accounts data on gross operating surplus.

We then impute NOSHN proportionally to each household's net imputed rent. Where a household comprises more than one adult, the income is equally split. Gross and net imputed rents are directly provided in the SIH from 2005 onwards.<sup>7</sup> For earlier years, we predict gross and net imputed rents. Using 2005 values, we estimate a model to predict gross imputed rents based on reported tenure type, state of residence, area of residence, number of bedrooms, household gross income decile and landlord type. The approach draws heavily on the approach developed by the ABS (ABS 2008a). For net imputed rent, all covariates listed above are interacted with (predicted) gross imputed rent and we add mortgage repayments and predicted gross imputed rent to the list of covariates. Coefficient estimates are reported in Appendix A.4. All models are estimated with and without tenure type as this variable was not available before 1995 in the SIH and thus cannot be used for imputation before

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<sup>6</sup> The minimum contribution rate is now 10% and is scheduled to gradually increase up to 12% by 1 July 2025.

<sup>7</sup> According to the ABS (2008b, p.3), 'Gross imputed rent is the market value of the rental equivalent, and has been estimated using hedonic regression. Net imputed rent for owner occupiers has been derived by subtracting the housing costs normally paid by landlords (i.e., council rates, mortgage interest, building insurance premiums, repairs and maintenance) from gross imputed rent.'

that year. These models fit the data well with the adjusted R-square 0.97 for gross imputed rent and 0.69 for net imputed rent.

#### *Other capital income*

Other capital income has two components: that captured by SIH and that not captured by SIH, the latter of which is a residual equal to total capital income<sup>8</sup> from the National Accounts minus superannuation income from the National Accounts minus non-pension capital income as measured in SIH. This non-captured capital income will primarily comprise corporate retained earnings. We distribute it assuming it has the same distribution as observed other (non-superannuation non-imputed rent) capital income. We take the same approach for adding foreign income received from tax havens and reinvested earnings on foreign portfolio investment. The latter captures retained earnings in foreign firms accruing to Australians whose shares comprise less than the 10% foreign direct investment threshold required to appear in the National Accounts. We use WID estimates of foreign income received from tax havens and reinvested earnings on foreign portfolio investment (see Zucman 2013).

Grossing-up factors reported in Appendix Table A.2 indicate that this captured capital income has to be multiplied by a factor of between 2 and 4.3 to match National Accounts totals.

#### *2.1.4. Taxes on production*

As Alvaredo et al. (2020) show, a pre-tax income measure not only requires that income taxes are not deducted from capital and labour income, but that taxes on production (and taxes on wealth, if they exist) need to be *added* to incomes to ensure all of national income is distributed to individuals. As per the guidelines, taxes on production are assumed to have the same distribution as total factor income. This is somewhat arbitrary, but means pre-tax income distributions among those with factor incomes are unaffected by these taxes other than via a scaling up factor applied to all incomes. Inequality measured over the total population increases, however, because people with zero factor incomes become relatively poorer.

### **2.2. Pre-tax national income**

Following the approach adopted for the French DINA by Garbinti et al. (2018) and US DINA by Piketty et al. (2018), as well as the DINA Guidelines, we include the Age Pension as income to produce pre-tax national income. This presents no major difficulty as the Age Pension is reported in the SIH.<sup>9</sup> We

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<sup>8</sup> Total capital income is defined here as ‘total net property income of households and non-profit institutions serving households’ plus ‘total net primary income of corporations’.

<sup>9</sup> Age Pension income is not directly reported in ALife, but we can combine information on receipt of government pensions and age to infer it. In addition, we use ALife only for the top 1%, a group almost certain not to receive the Age Pension given that it is subject to both an income test and an asset test.

distribute the total cost of Age Pension payments as a flat percentage of income tax liabilities. That is, we assume each individual's contribution to the funding of the Age Pension is in proportion of their income tax liabilities.

### 2.3. Post-tax disposable income

To move from pre-tax national income to post-tax disposable income requires deducting all taxes and adding all government cash transfers to individuals' pre-tax incomes. Deducting income taxes and adding cash transfers is straightforward since both are recorded in the SIH and ALife data. However, both income taxes and cash transfers need to be scaled up to match National Accounts totals.

As noted in the DINA guidelines (Alvaredo et al., 2020, p. 53), the aim is to "to describe post-tax, post transfer inequality for the population's actual perceived budget constraints, while excluding in-kind transfers such as health and education and other public spending (as these may impact purchasing power and disposable income only indirectly). For this reason, aggregate post-tax disposable income can be substantially less than aggregate national income."

Government pensions and allowances, as well as income taxes, are distributed according to the survey (and tax) data. For taxes on production (indirect taxes), which were distributed proportionally to factor income in pretax series, the DINA Guidelines advocate they are removed in proportion to *consumption*, proxied by disposable income (before the deduction of taxes on production) minus saving (where savings rates are based on external sources). In the absence of data on savings rates by level of income, we simply remove production taxes proportionally to household disposable income (as defined in the SIH).<sup>10</sup> Corporate taxes are imputed proportionally to capital incomes after excluding imputed rent.

### 2.4. Post-tax national income

Moving from post-tax disposable income to post-tax national income requires distributing government expenditure to individuals. This corresponds to total expenditure of the government adjusted for the surplus or deficit of the government (Alvaredo et al., 2020, p. 64). The DINA Guidelines' definition of government surplus or deficit differs from the usual definition "due to the exclusion of other current transfers and capital transfers" (Alvaredo et al., 2020, p51). Thus the government surplus or deficit is defined as net saving plus net other current transfers.

Three alternative approaches to distributing government expenditure are recommended by the DINA Guidelines: (1) assume health expenditures benefit all adults equally but that the benefits of

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<sup>10</sup> A potential refinement for future work is to estimate expenditure regression models using the ABS Household Expenditure Survey data (collected in 1993, 1998, 2003, 2009 and 2015) to impute household expenditure as a function of income (and perhaps other factors) and use this to distribute taxes on production.

other expenditures are proportional to disposable income; (2) assume everyone benefits equally from all government expenditure; and (3) assume the benefits of government spending are distributed in the same way as disposable income. The third approach means government spending can effectively be ignored since it doesn't affect the distribution other than to scale up everyone's income by the same fraction. Interestingly, the Guidelines do not allow for a scenario where government spending is redistributive.

In Australia, the biggest expenditure items—health and education—are somewhat redistributive to lower-income individuals (ABS, 2018, Table 1.1). Consequently, of the approaches the guidelines recommend, the most appropriate approach for the Australian context is Approach (2). This means average government expenditure per adult is added to disposable income. This acts to lower measured inequality compared with post-tax disposable income, but nonetheless is likely to overstate benefits to high-income earners and understate benefits to low income earners and thus not reduce measured inequality as much as it should.

### 3. Inequality in Australia 1991-2018

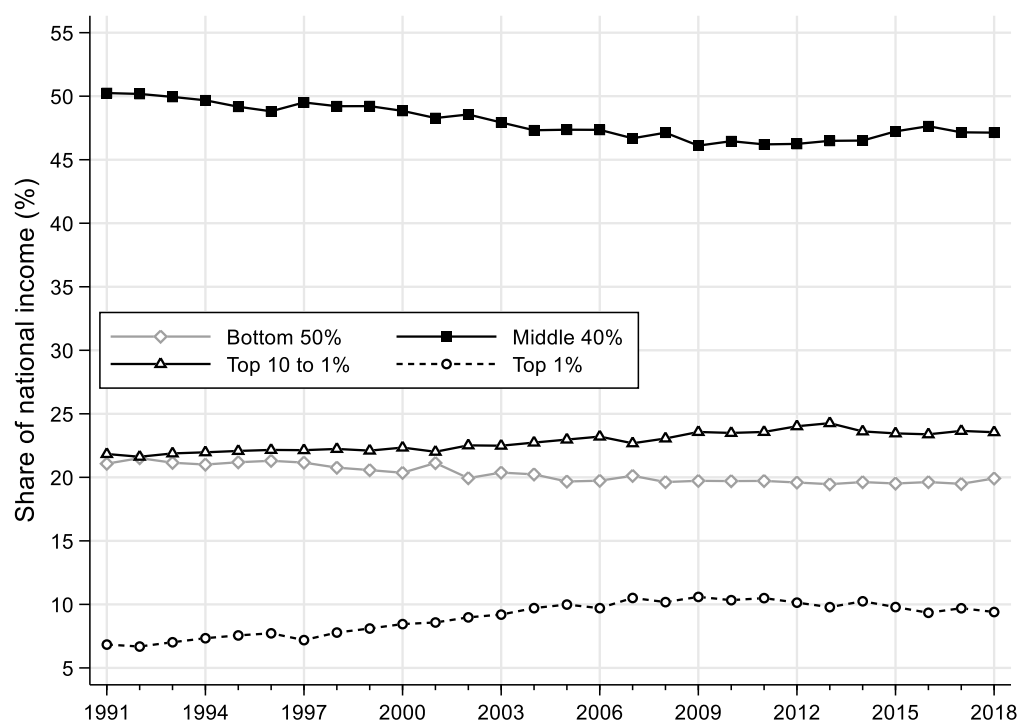
#### 3.1. Pre-tax national income

Figure 3.1 presents estimated shares of pre-tax national income over the 1991 to 2018 period of the bottom 50%, top 50% excluding the top 10% (referred to as the 'middle 40%'), top 10% excluding the top 1%, and the top 1%. As noted, this provides information on how a 'market income' concept of income is distributed across individuals. The share of the bottom 50% remained relatively steady, at approximately 20%, but the middle 40% group experienced a decline from over 50% to 47.5%, with the decline occurring between 1991 and 2008, since when there has been no net change. The income share of the top 10% to 1% rose from 22% to 23.4%, while the top 1% income share rose from 7% to 9.4%, with all the increase occurring between 1995 and 2008 (and indeed there is a small decline evident after 2008).<sup>11</sup>

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<sup>11</sup> Data Appendix F (to be available online) contains series (in Stata files) that describe thresholds, averages and shares for each of the 127 'generalised percentiles' (or g-percentiles) for each income concept, as recommended by the DINA Guidelines.

**Figure 3.1: Pre-tax national income shares 1991-2018**

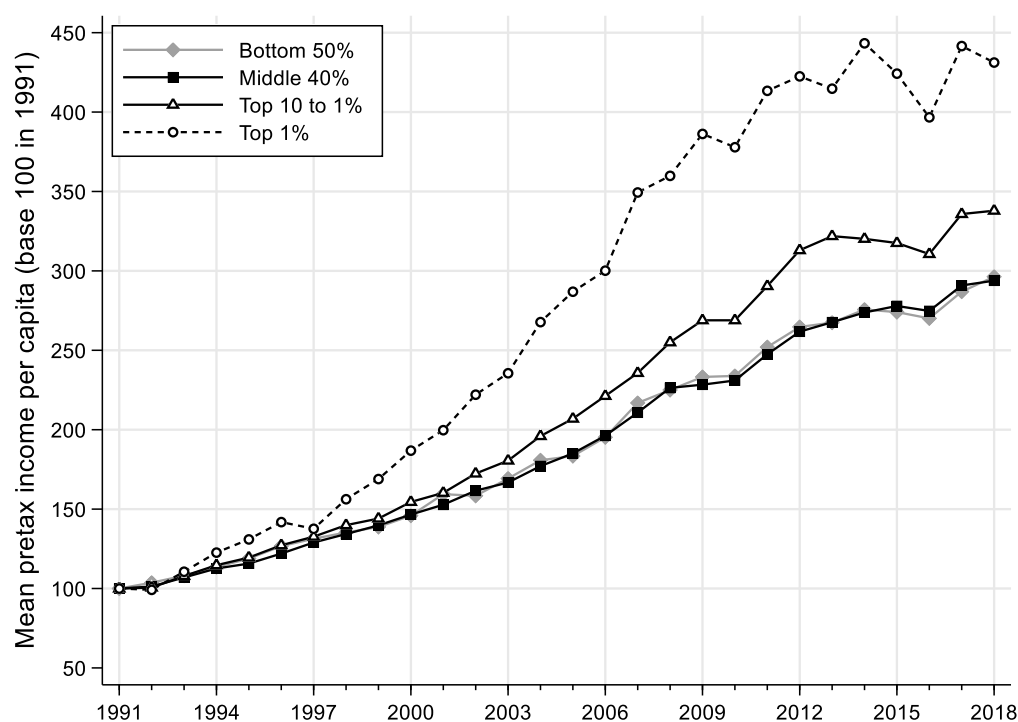


Notes: Distribution of pretax national income (before all taxes and transfers, except age pensions) among adults. Broad equal-split adults series (household income equally split among adults).

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Figure 3.2 compares the changes in mean income per adult of each of the four income groups examined in Figure 3.1. Since 1991, the mean income of the top 1% has increased by a factor of more than four. This compares with nearly 3.5 for the top 10% to 1% and approximately 2.9 to 3 for the two groups comprising the bottom 90%.

**Figure 3.2: Mean adult pre-tax national income by income group 1991-2018**



Notes: Distribution of pretax national income (before all taxes and transfers, except age pensions) among adults. Broad equal-split adults series (household income equally split among adults). Index based on mean incomes in current dollars.

*Stata graph pretax\_mean\_split2\_index\_a*

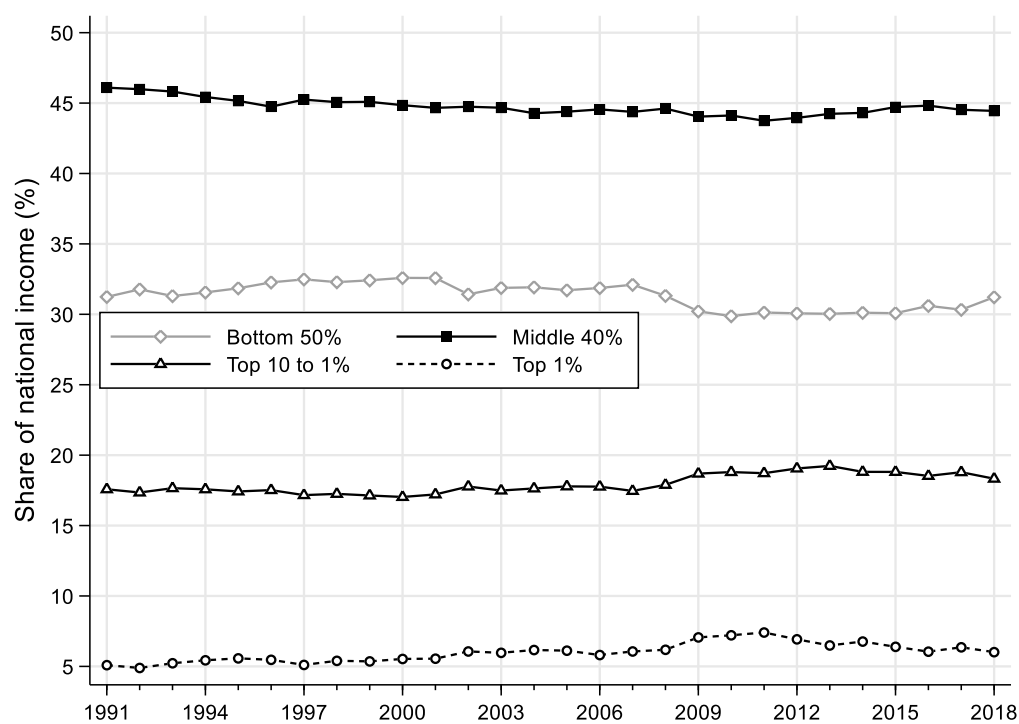
### 3.2. Post-tax national income

Figures 3.3 and 3.4 present the same information as Figures 3.1 and 3.2, but for post-tax national income. This provides information on the distribution across individuals of 'beneficial receipt' of total income in the National Accounts. The relative rise in top income shares is less pronounced for this income measure, but notable is that the income share of the bottom 50%, after rising slightly between 1991 and 2007, subsequently fell to 2010, and has largely not recovered.

Consistent with the findings of Figure 3.3, Figure 3.4 shows differences in income growth across the four income groups are more subdued for post-tax national income than for pre-tax national income.



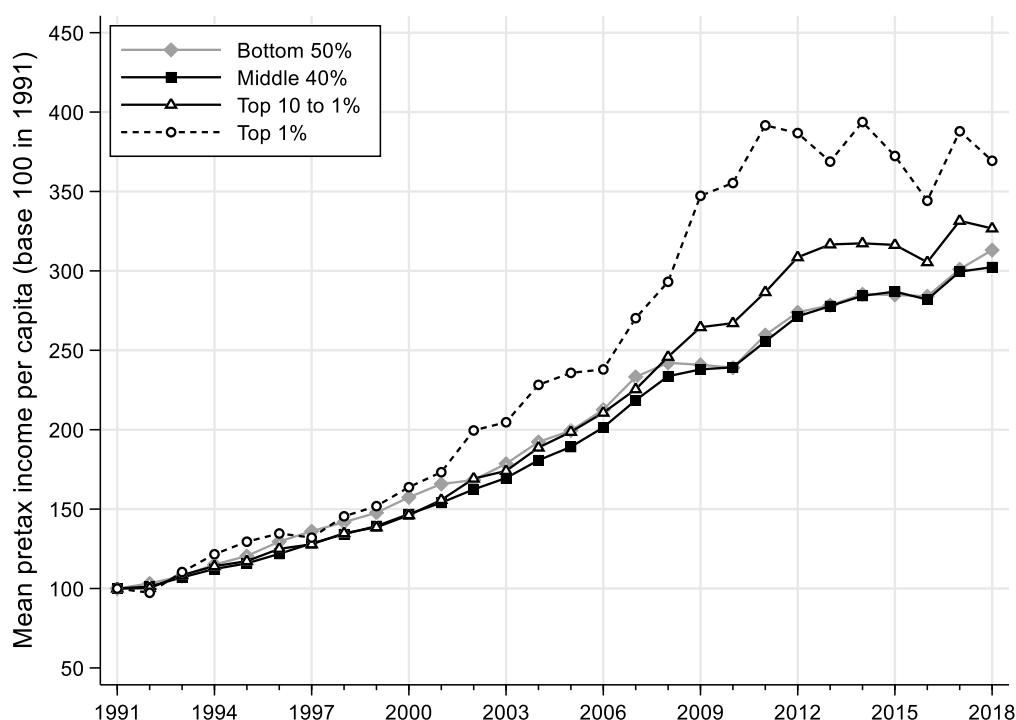
**Figure 3.3: Post-tax national income shares 1991-2018**



Notes: Distribution of post-tax national income (after all taxes and transfers) among adults. Broad equal-split adults series (household income equally split among adults).

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**Figure 3.4: Mean adult post-tax national income by income group 1991-2018**



Notes: Distribution of post-tax national income (after all taxes and transfers) among adults. Broad equal-split adults series (household income equally split among adults). Index based on mean incomes in current dollars.

*Stata graph ptninc\_split2\_mean\_index\_a*

## 4. International comparisons: Australia, US and France

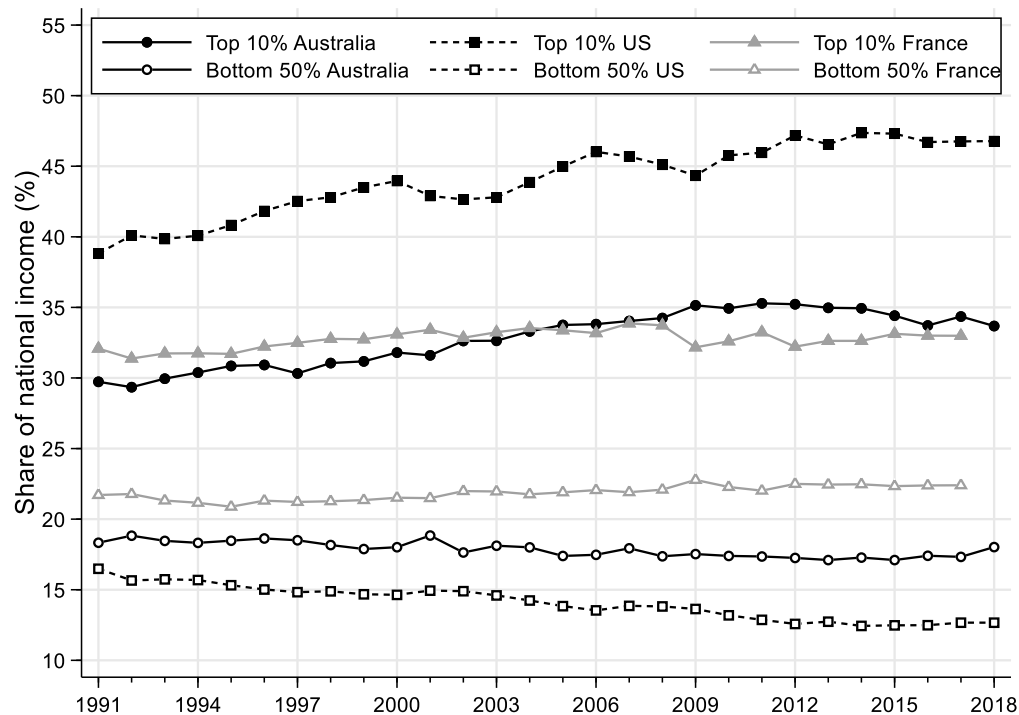
### 4.1. Pre-tax national income

In this section we compare US, French and Australian income shares of four income groups: top 1%, top 10%, top 50% to 10% ('middle 40%') and bottom 50%.<sup>12</sup> Figures 4.1 and 4.2 examine pre-tax national income, the first figure examining the top 10% and bottom 50% and the second figure the top 1% and middle 40%. The top 10% income share is considerably higher in the US than in Australia and France, which have similar top 10% income shares. The income share of the top 10% has also risen considerably in the US. It has also risen in Australia, albeit to a smaller degree, while it has remained relatively stable in France, such that the top 10% income share has gone from being somewhat higher in France than in Australia in the early 1990s to slightly lower in the late 2010s. Similar patterns are evident for the top 1% in Figure 4.2, although the income share of the top 1% in France remains slightly above that of the top 1% in Australia throughout the 1991 to 2018 period.

<sup>12</sup> Results for Australia differ slightly from those presented in the previous section because we use the 'narrow equal-split' series in the comparison with the US and France to ensure comparability with these countries' estimates (see Section 2.1).

For the bottom 50%, France and Australia are again relatively similar and somewhat different to the US. However, there is a slight but steady rise in the income share of the bottom 50% in France from the mid 1990s, compared with a slight decline in Australia. Across the entire period, the 'middle 40%' (top 50% to 10%) has had the highest income share in Australia and lowest income share in the US. In all three countries, this income groups has experienced a decline in income share, with the drop greatest in the US and smallest in France.

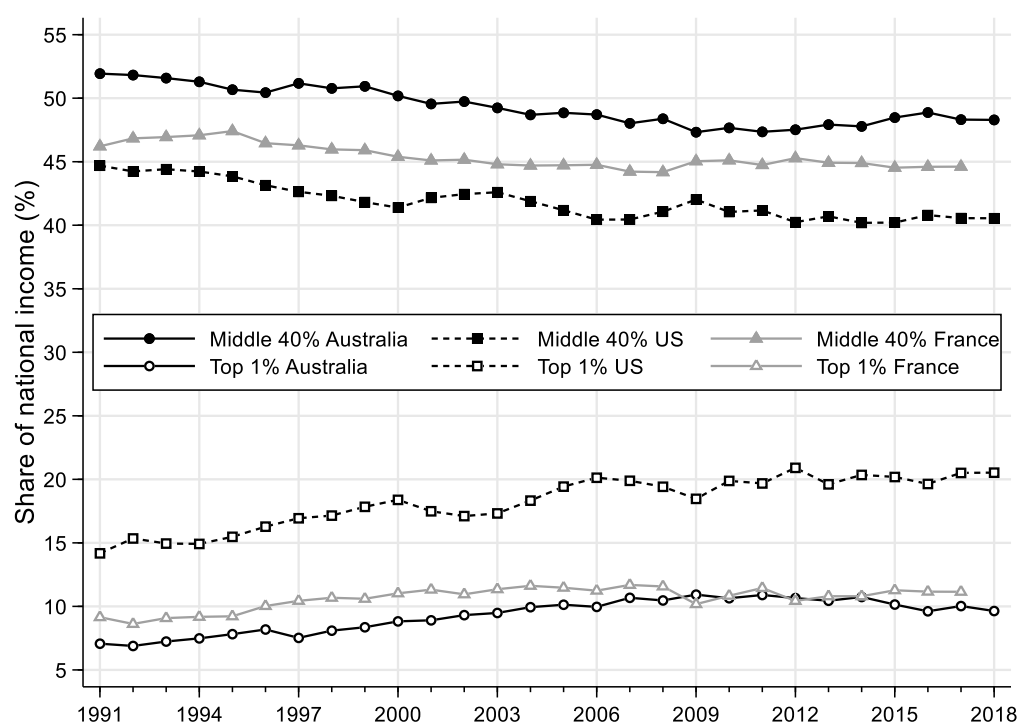
**Figure 4.1: Top 10% and bottom 50% income shares: Australia, US and France 1991-2018**



Notes: Distribution of pretax national income (before all taxes and transfers, except age pensions) among adults. Narrow-equal split-adults series (income of married couples divided by two).

*Stata graph intcomp\_pretax\_sh1*

**Figure 4.2: Top 1% and middle 40% income shares: Australia, US and France 1991-2018**



Notes: Distribution of pretax national income (before all taxes and transfers, except age pensions) among adults. Narrow-equal split adults series (income of married couples divided by two).

*Stata graph intcomp\_pretax\_sh2*

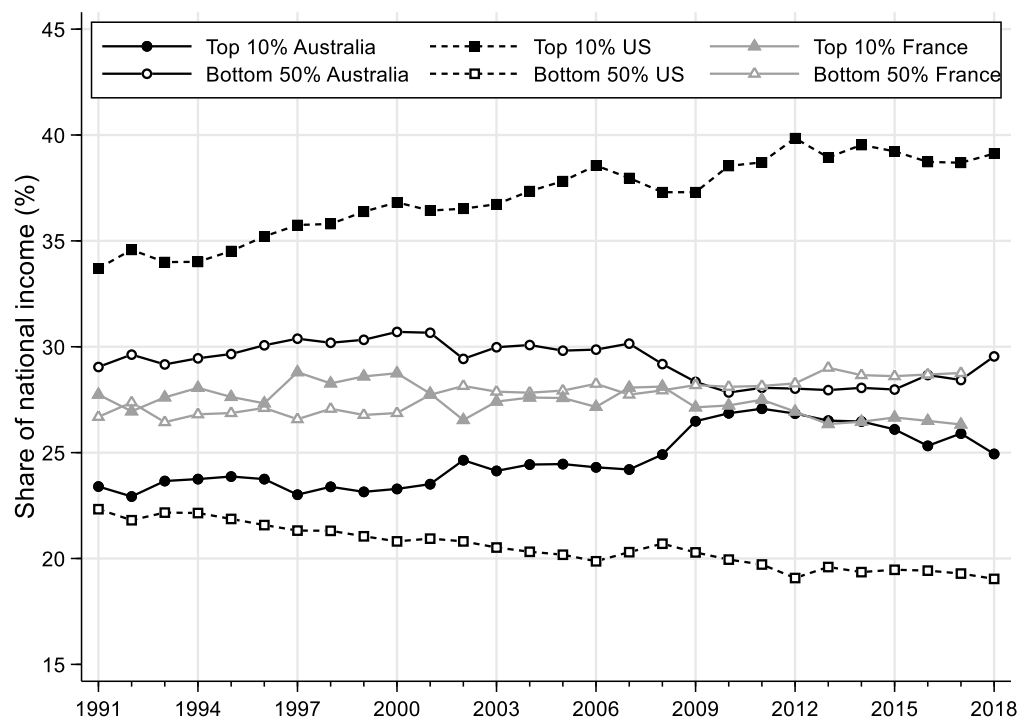
## 4.2. Post-tax national income

Comparisons across the US, France and Australia in the distribution of post-tax national income are presented in Figures 4.3 and 4.4. Differences across the three countries are stark. The top 10% in the US received nearly 34% of income in 1991, and this had risen to nearly 39% in 2018. In France, the top 10% received approximately 27% of income in 1991 and this share fell slightly to approximately 26% in 2018. In Australia, the top 10% income share was approximately 23% between 1991 and 2001, but then increased to nearly 26% in 2010 and subsequently declined only slightly. For the top 1% (Figure 4.4), the US again has a much higher income share and greater growth in the income share than France and Australia. The top 1% share is higher in France than in Australia, with the gap being approximately 2 percentage-points in 1991 as well as in the most recent years.

The income share of the bottom 50% is highest in Australia and lowest in the US. There is little net change evident over the full period for France and Australia, but a considerable decline for the US. At the end of the period, the income share of the bottom 50% was 33% in Australia, 29% in France and 19% in the US. For the middle 40%, income shares are very similar across the three countries,

although across the entire period, France has the highest income share and the US the lowest, and the gap widened slightly between 1991 and 2018.

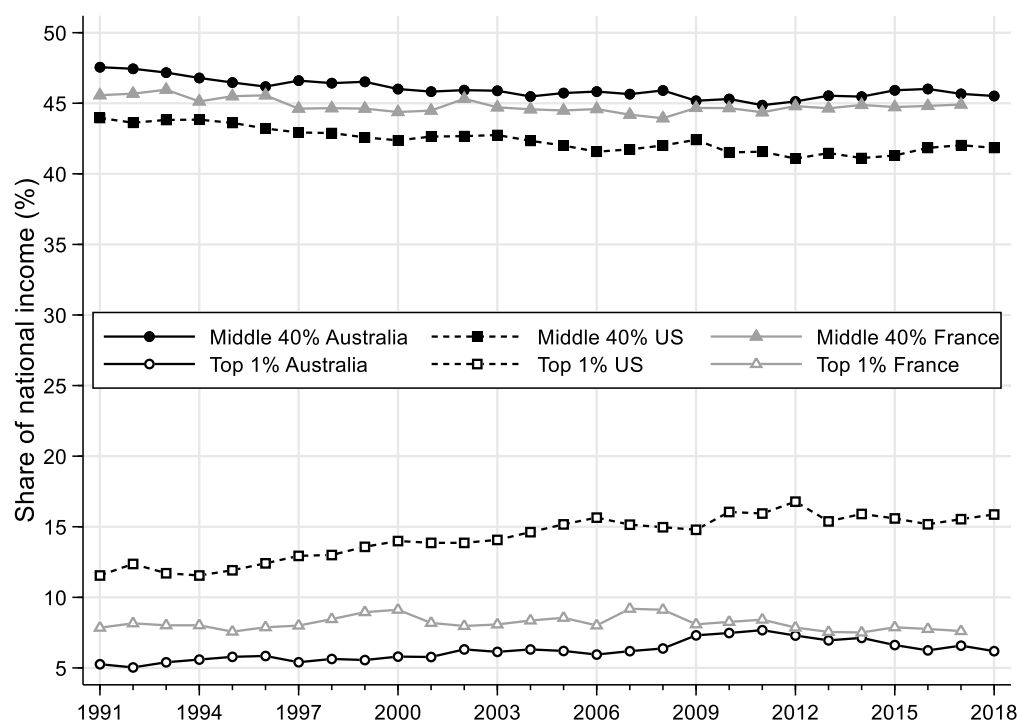
**Figure 4.3: Top 10% and bottom 50% income shares: Australia, US and France 1991-2018**



Notes: Distribution of post-tax national income (after all taxes and transfers) among adults. Narrow-equal split adults series (income of married couples divided by two).

*Stata graph intcomp\_ptninc\_sh1*

**Figure 4.4: Top 1% and middle 40% income shares: Australia, US and France 1991-2018**



Notes: Distribution of post-tax national income (after all taxes and transfers) among adults. Narrow-equal split adults series (income of married couples divided by two).

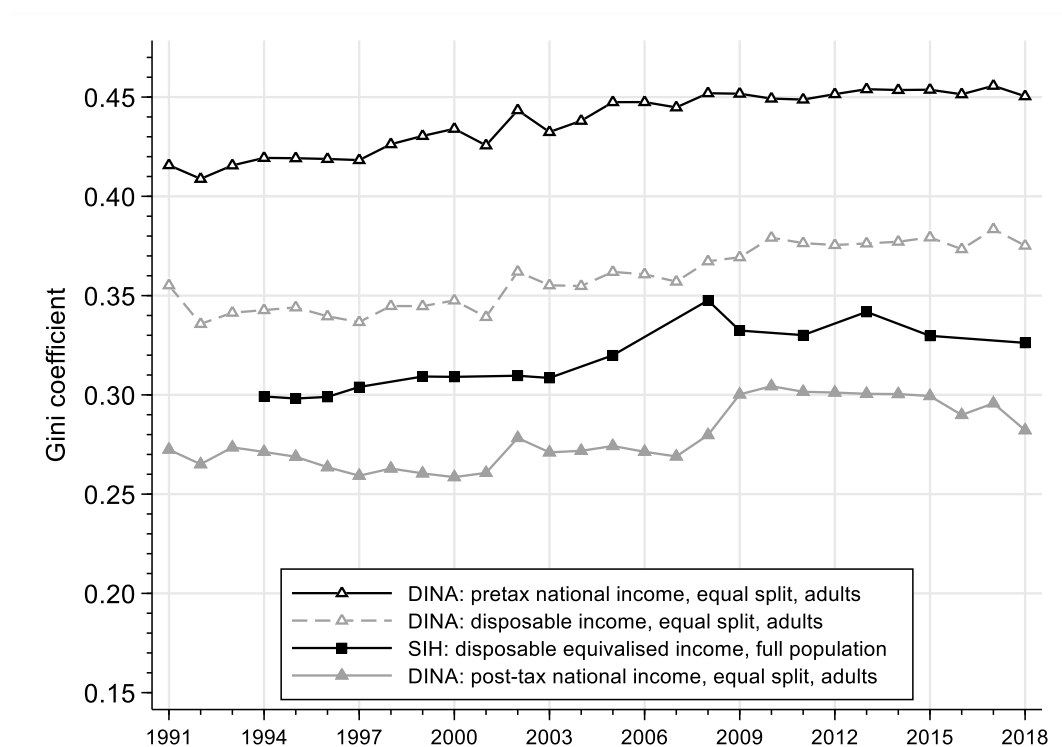
*Stata graph intcomp\_ptninc\_sh2*

## 5. Comparisons of DINA estimates with household survey estimates of inequality

Of considerable interest is how inferences on levels and trends in inequality are affected by moving from traditional household-survey based estimates for household equivalised disposable income to DINA estimates of inequality. Figure 5.1 compares the Gini coefficient for three of the DINA income concepts with the Gini coefficient for equivalised disposable income captured in the SIH (where the modified OECD scale is used to equalise income; see Hagenaars et al. 1994).

As would be expected, comparing across the DINA income concepts, moving from pre-tax national income to post-tax disposable income and then to post-tax national income is associated with decreases in the Gini coefficient. Notably, the Gini coefficient for post-tax national income is consistently below the Gini coefficient for equivalised disposable (cash) income. Between 1994 and 2018, Gini coefficients for both post-tax national income and equivalised disposable income increased, although more so for equivalised disposable income.

**Figure 5.1: Gini coefficients for DINA series and household equivalised disposable income, 1991-2018**



Notes: Broad equal-split adults DINA series (household income equally split among adults).

*Stata graph Gini1a*

## 6. Income levels and inequality disaggregated by demographic characteristics

A valuable feature of the DINA series for Australia is that it is primarily based on survey data which contains demographic information not typically available in administrative data sources such as tax records. Here, we exploit this extra richness of survey data to shed further light on the distribution of national income across and within various population groups.

In what follows we consider sex, age, education, immigrant status and area of residence (i.e., cities versus regional areas). We focus on the post-tax national income series based on the equal splitting of household income between all adult members (i.e., the ‘broad equal-split series’). However, we also bring in insights from the ‘individualistic series’ where it is most relevant, that is for sex and education, because equal-split series can mute differences across these groups.

For each demographic characteristic we consider, we first present and discuss differences in mean incomes before turning to the income shares of national income within each subgroup. Mean incomes are useful to show differences in levels across groups, regardless of the size of each group. Mean income is preferred over income shares because the latter are a reflection of both mean

incomes and population shares. The income shares within each subgroup (e.g., among men or among university graduates) are then presented to shed light on the levels and trends of inequality within each subgroup.

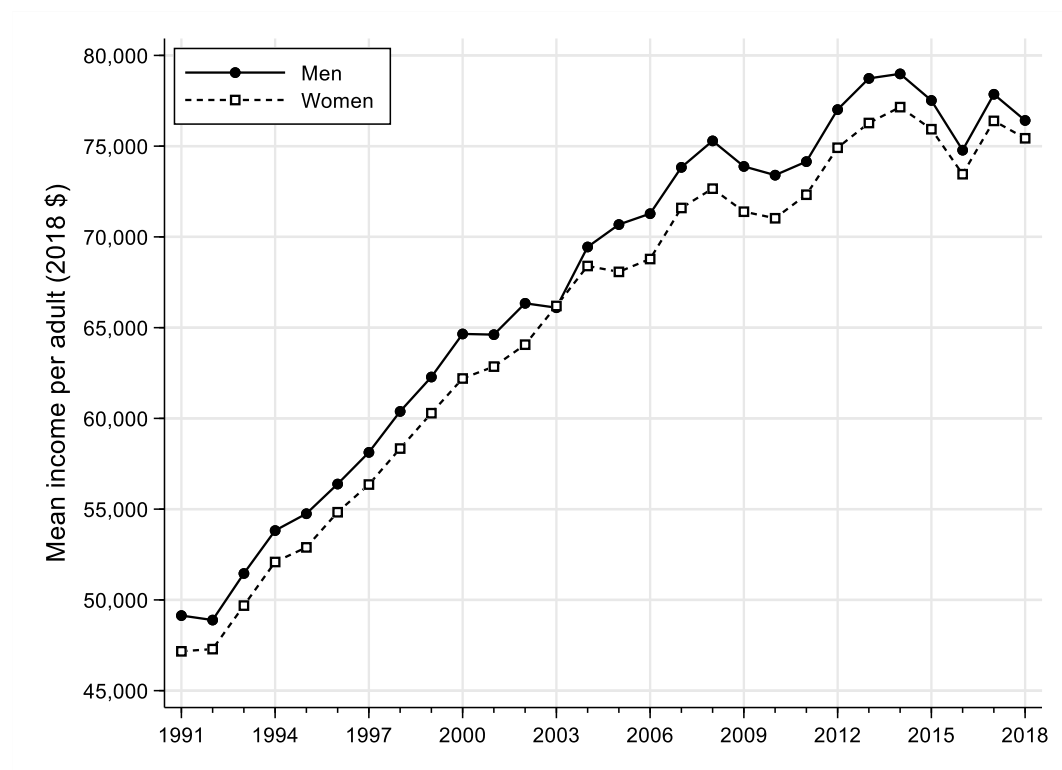
### 6.1. DINA by sex

Figure 6.1 shows that mean post-tax national income does not differ much by sex. While men have slightly higher incomes, the difference is limited and has remained stable. However, gender differences are muted in these DINA series by the use of equal-split incomes, meaning that all incomes are equally split among adult household members. The implication is that any remaining gender difference is driven by differences between single men and single women. Appendix Figure E.1 shows exacerbated gender differences if we use 'individualistic' income series, with women's mean incomes falling below men's mean incomes by about \$25,000 throughout the period.

Figure 6.2 shows that the income shares among men and among women are largely similar. The income shares of the top 10% to 1% of men is slightly larger than the share of this group among women, whereas the bottom 50% of men have a smaller income share. Again, relaxing the equal-split assumption to consider individualistic series leads to larger differences (Appendix Figure E.2). Notably, the individualistic series reveal that top income shares (for the top 10% to 1% and the top 1%) were higher among women than among men in the early 1990s, and the bottom 50% income shares were smaller. This thus shows more inequality among women than among men. This is likely to reflect the lower labour market participation of partnered women, with a significant proportion not employed and therefore having low or no personal income. However, consistent with the rise in female employment participation over the period since 1991, the following decades saw a reversal. By 2018 the income share of the top 10% to 1% was larger among men than among women, while the bottom 50% of women had a larger income share than the bottom 50% of men. Hence, there appears to be more income disparities among men than among women in recent years.



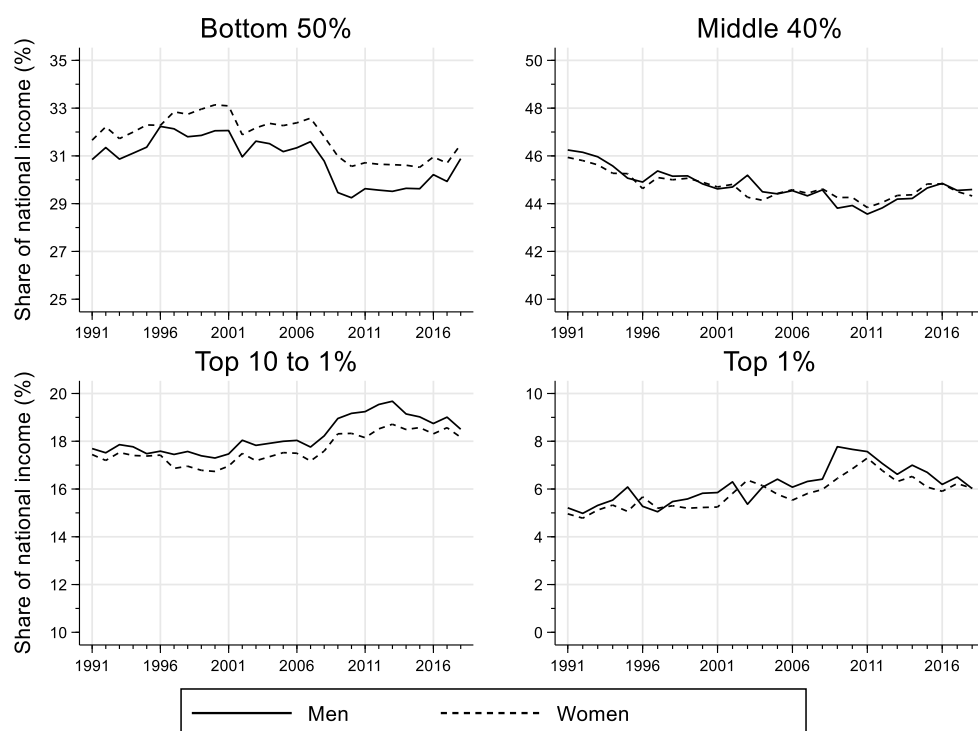
**Figure 6.1: Mean real post-tax national income per adult by sex (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph ptninc\_mean\_split2\_sex\_b*

**Figure 6.2: Post-tax national income shares among men and women (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

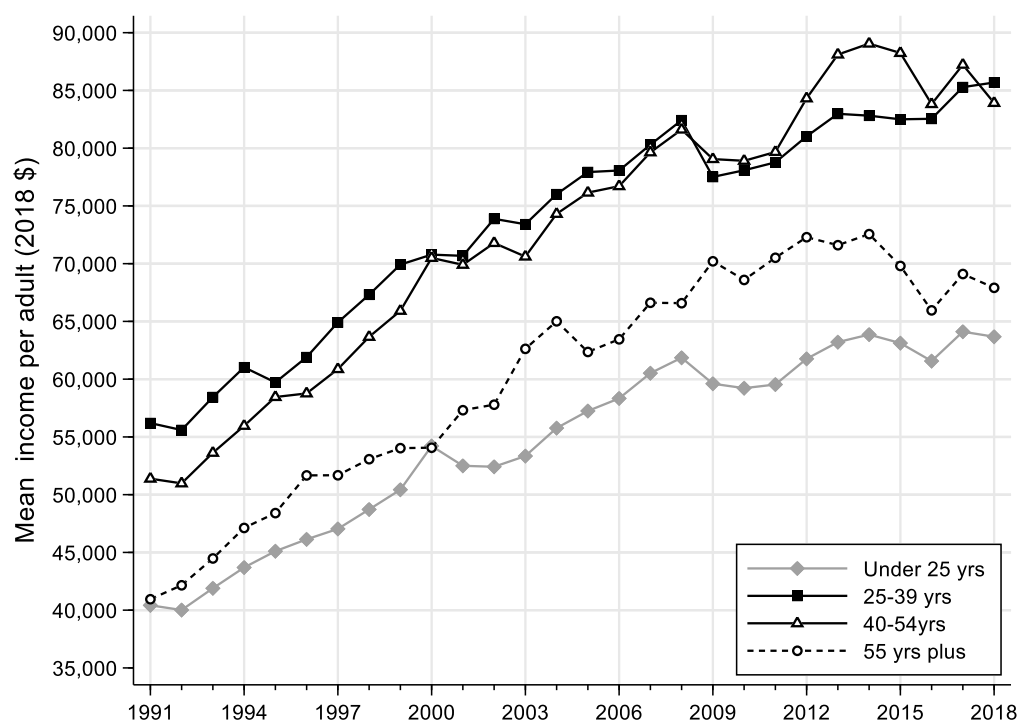
*Stata graph pshares\_split2\_sexp\_ptninc\_b*

## 6.2. DINA by age group

Mean post-tax national income is the highest among prime working-age adults—that is, those aged 25 to 39—and those aged 40 to 54, followed by those aged 55 and above. It is lowest for those under 25. Although, these differences in levels have been true since 1991, Figure 6.3 shows that the mean incomes of both the youngest and oldest age groups have been falling further behind that of the two prime working-age groups, particularly in the 2010s.

Figure 6.4 shows income shares within each age group. The share of the bottom 50% is the highest among those under 25 and the lowest among those 40 to 54 years of age. The share of the top 10% to 1% is higher among the two older age groups—that is, those above 40—and the lowest in the youngest two age groups. The same is largely true for the top 1% income shares, although there are more fluctuations, and the differences across age groups tend to be smaller.

**Figure 6.3: Mean post-tax national income per adult by age group (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

Stata graph ptninc\_mean\_split2\_age\_b

**Figure 6.4: Post-tax national income shares within age groups (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

Stata graph pshares\_split2\_agecat2\_ptninc\_b

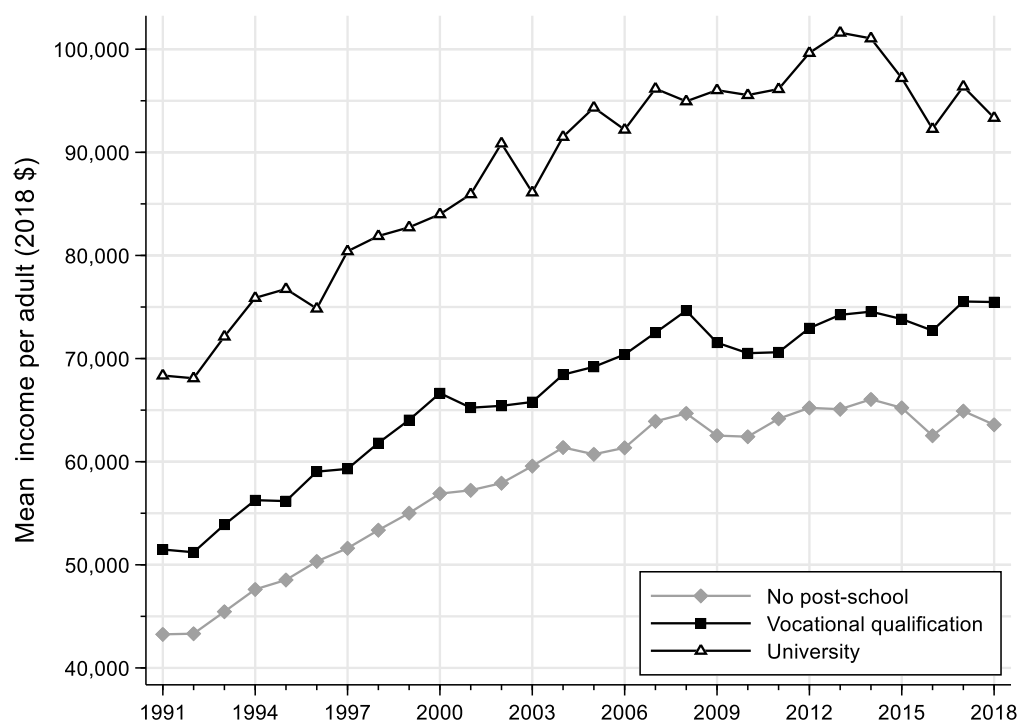
### 6.3. DINA by educational attainment

Figure 6.5, comparing across three levels of educational attainment, shows that mean post-tax national incomes are, unsurprisingly, ordered by educational attainment. Perhaps more interesting is that all three education groups seemed to equally benefit from the trend increase in post-tax national income per adult up until the GFC in 2008. However, since 2008 there has been no net growth in mean incomes for all three education groups.

To the extent that there is not complete assortative mating based on education, this DINA series based on equal-split incomes will show smaller education premia than the individualistic series. Indeed, Appendix Figure E.3 shows larger education premia, in particular for vocational qualifications. The individualistic series also reveals a trend increase in the university education premium, rising from approximately 47% (compared with no post-school qualifications) in 1991 to 57% in 2018.

Figure 6.6 takes a look inside each of these three educational groups to reveal that inequality is the highest among university graduates, as measured by their higher top 10% to 1% and top 1% income shares, and their lower shares for the bottom 50%. Income shares among those with no post-school qualifications and those with vocational qualifications are comparable. Over the period as a whole, there is no clear trend in these income shares. As was the case with sex, inequality within each education group is exacerbated by the use of individualistic income series (Appendix Figure E.4).

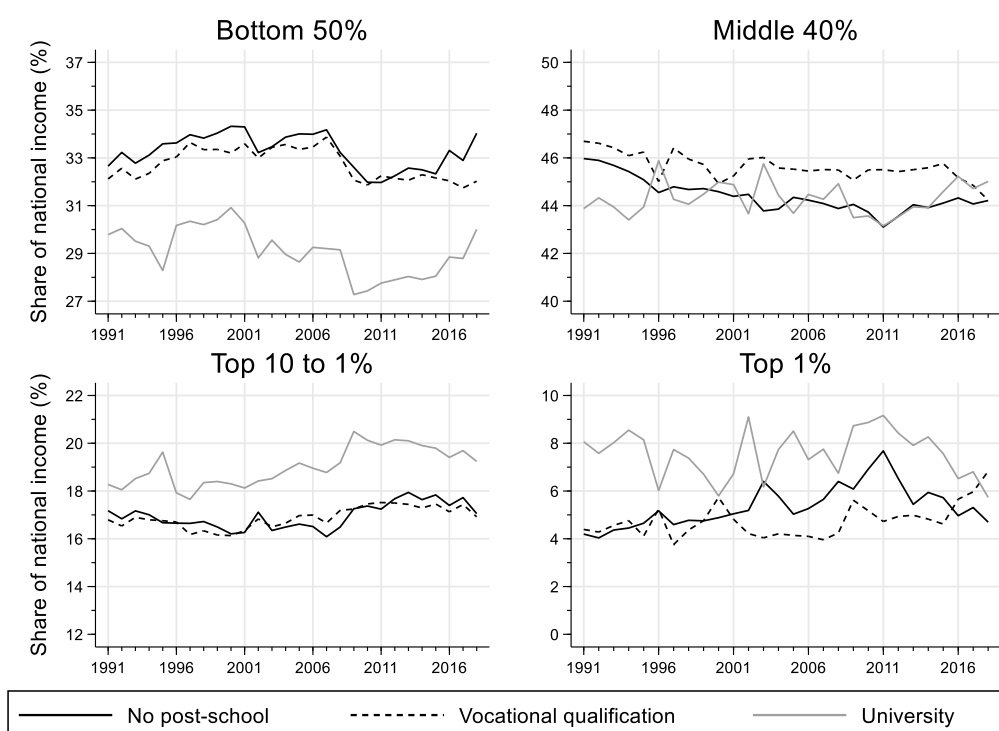
**Figure 6.5: Mean post-tax national income per adult by educational attainment (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph ptninc\_mean\_split2\_edu\_b*

**Figure 6.6: Post-tax national income shares within education groups (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

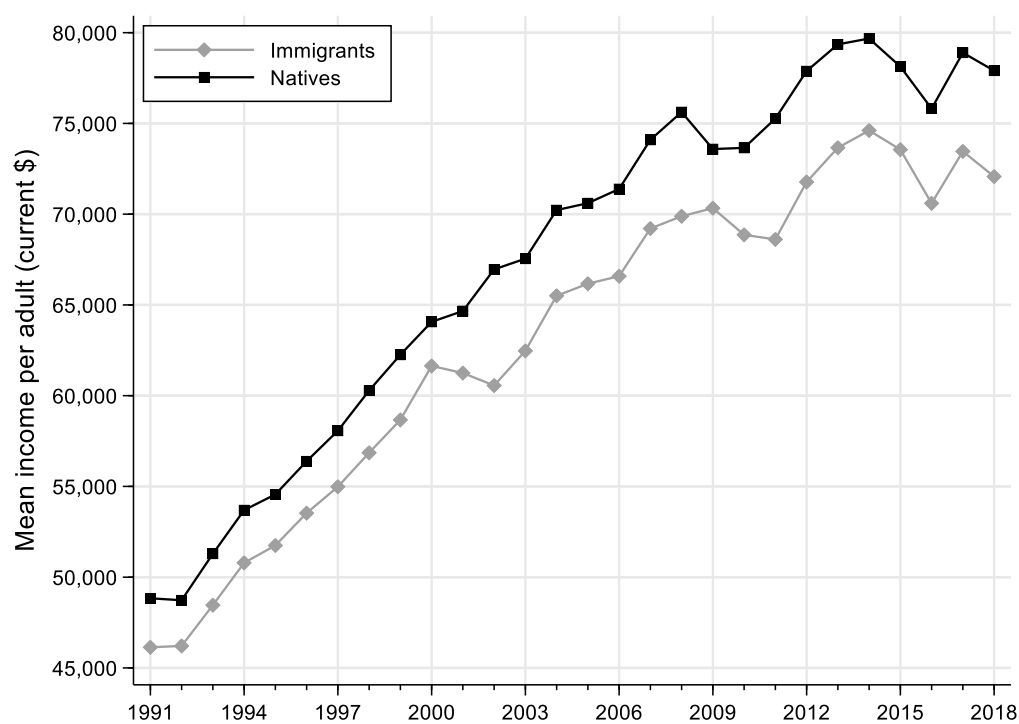
*Stata graph pshares\_split2\_edu\_ptninc\_b*

#### 6.4. DINA by immigrant status

The ABS SIH data allow us to distinguish foreign-born and native-born individuals.<sup>13</sup> Figure 6.7 shows that mean post-tax national income is greater for native-born Australians than for immigrants throughout the 1991-2018 period. The gap increased after the 1990s to reach almost 8% (or about \$6,000) in 2018, compared to 6% (\$3,000) in 1991. Figure 6.8 shows that income shares are, perhaps surprisingly given the mean income differences, similarly distributed among immigrants and among the native-born.

<sup>13</sup> More detailed information on country of birth is available in some survey years, but not consistently across our period of analysis.

**Figure 6.7: Mean post-tax national income per adult for immigrants and natives (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph ptinc\_mean\_split2\_migrant\_b*

**Figure 6.8: Post-tax national income shares among immigrants and natives (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph pshares\_split2\_ausborn\_ptninc\_b*

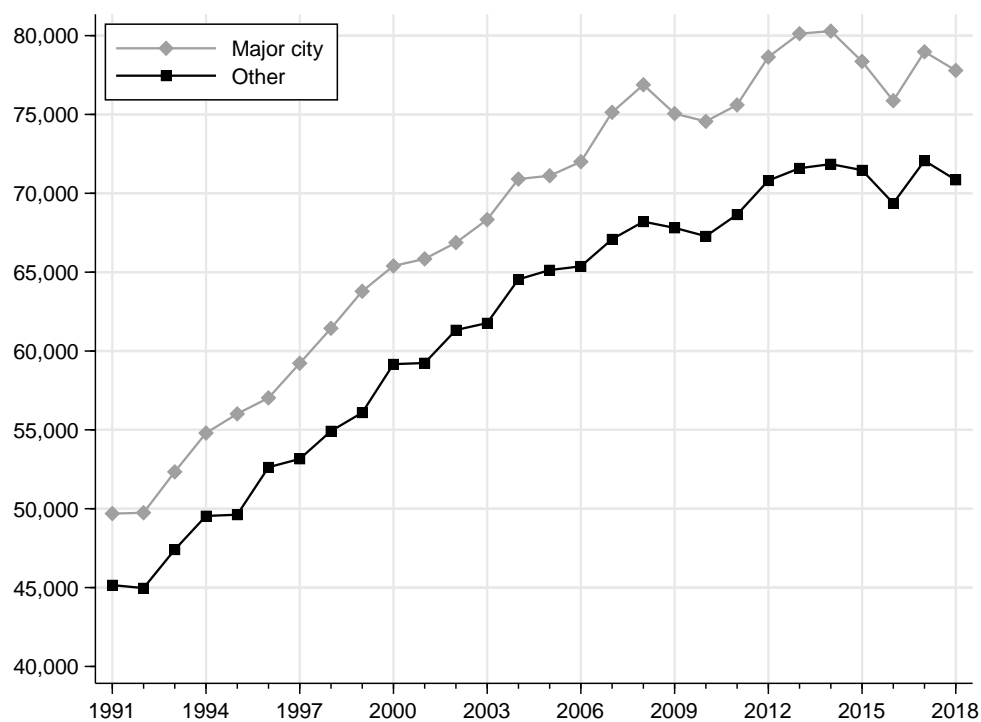


## 6.5. DINA by area of residence

We distinguish major cities and the rest of Australia. Figure 6.9 shows that mean post-tax national income is higher in major cities by about 10% and that this gap has remained stable in relative terms over the 1991-2018 period.

Figure 6.10 reveals some shifts in the distribution of post-tax national income in major cities and in the rest of Australia. Top 1% income shares are larger in major cities throughout the period but, while other income shares were similar across both types of region in the early 1990s, they have since diverged somewhat, driven by an increase in inequality in major cities. The result is that the income shares of the bottom 50% are now lower, and the income shares of the top 10% to 1% are larger, in major cities than in the rest of Australia.

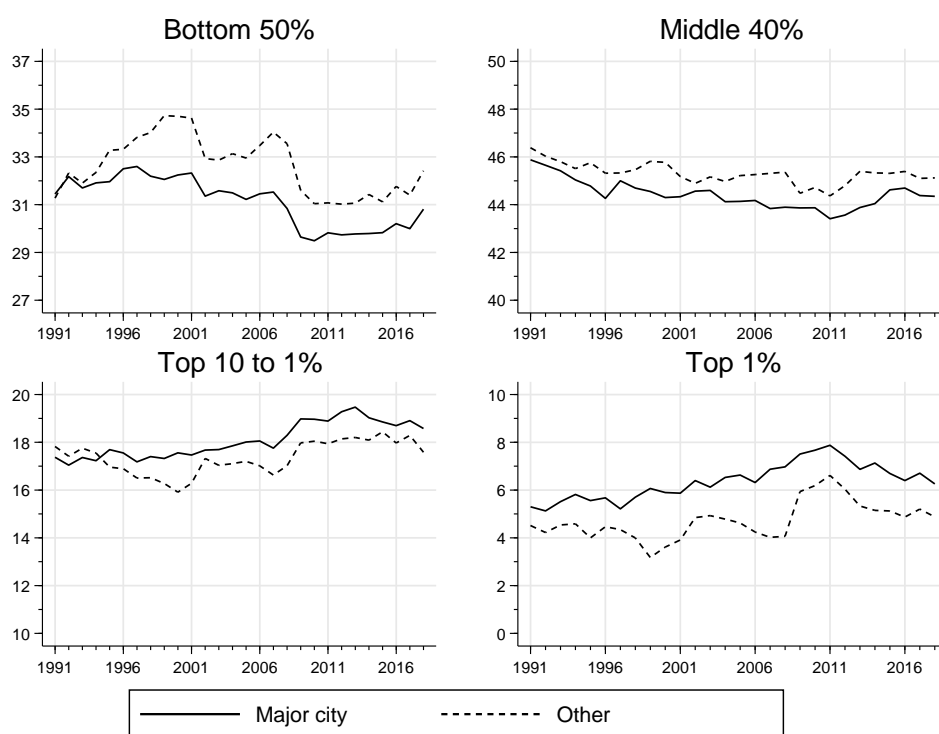
**Figure 6.9: Mean post-tax national income per adult by area of residence (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph ptinc\_mean\_split2\_areahcf\_b*

**Figure 6.10: Post-tax national income shares in major cities and other areas (1991-2018)**



Notes: Broad equal-split adults series (household income equally split among adults).

*Stata graph pshares\_split2\_areahcf\_ptninc\_b*

## 7. Conclusion

We have produced the first DINA estimates for Australia consistent with the DINA Guidelines described in Alvaredo et al. (2020), spanning the period 1991 to 2018. Further refinement is certainly possible and should be a priority for further research. This notwithstanding, our estimates suggest Australia has a somewhat similar distribution to France, with both countries having considerably more equitable distributions than the US. Australia has, however, had greater growth in inequality than France.

Significantly, our DINA estimates for Australia indicate that income inequality is somewhat lower when all income as measured in the National Accounts is distributed to individuals compared with a focus on cash incomes as is conventional in household survey based studies of income inequality. This is despite our assumption that in-kind income from government expenditure is equally distributed across the population, which is almost certainly not true. On balance, government expenditure is likely to be progressive in its effects, as evidenced by the ABS in its periodic ‘fiscal incidence’ studies (ABS, 2018).

In contrast to other DINA studies internationally, our reliance on household survey data to anchor our distributional analysis has allowed us to consider income differences between and within demographic groups. The analysis presented in this paper has only investigated these differences in a cursory fashion, but clearly there is considerable potential to exploit this feature of our series in future research.

The DINA Guidelines recommend production of synthetic microfiles, but confidentiality requirements of ABS and ALife data access preclude this. However, it would be possible to produce a synthetic file which, while an aggregation of the data, would have a sufficient number of ‘observations’ and sufficient information for each observation to allow a variety of distributional analyses. This is a potentially valuable avenue for future research.

A further important future research direction is to attempt to extend the DINA estimates back to earlier years. Unit record income survey data is more sparse prior to the 1990s, and indeed non-existent prior to 1975. Similarly, unit record tax data only extends back to 1991. Methods for producing DINA estimates will therefore need to rely on more aggregated forms of data, such as the tax tables used to produce the original (cash income) top income shares for WID.

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## Appendix A Data appendix

### 1. Appendix A.1 Preliminary adjustments to top 1% in ALife

A few adjustments are performed in ALife before it is combined with survey data for the top 1%.

Incomes and income components are not top coded in ALife, with one exception: in each year, the 24 largest ‘employment termination’ (redundancy) payments in the entire tax filer population are reduced to the level of the 25<sup>th</sup>-largest payment value. Between 1991 and 2017, this represented a total adjustment of between \$8 million and \$57 million in total and (noting that ALife is a 10% sample) affected between 0 and 7 individuals in ALife each year (see Table A.1). We distribute the portion of Employment Termination Payment that was cut due to top-coding: we take 10% of the total shortfall and divide it between all top-coded observations in ALife.

**Table A.1: ATO adjustment of Employment Termination Payments (1991-2017, in current dollars)**

Tax year	Total adjustment	Mean adjustment
1991	-8,241,221	-343,384
1992	-14,613,357	-608,890
1993	-11,291,962	-470,498
1994	-15,332,505	-638,854
1995	-15,336,185	-639,008
1996	-10,649,353	-443,723
1997	-16,235,498	-676,479
1998	-16,477,313	-686,555
1999	-27,519,147	-1,146,631
2000	-42,319,718	-1,763,322
2001	-15,676,018	-653,167
2002	-27,836,179	-1,159,841
2003	-24,939,858	-1,039,161
2004	-18,548,200	-772,842
2005	-20,565,614	-856,901
2006	-21,123,878	-880,162
2007	-32,612,964	-1,358,874
2008	-47,913,907	-1,996,413
2009	-51,932,913	-2,163,871
2010	-51,184,871	-2,132,703
2011	-56,628,558	-2,359,523
2012	-34,672,593	-1,444,691
2013	-20,742,471	-864,270
2014	-36,156,344	-1,506,514
2015	-14,594,126	-608,089
2016	-23,385,390	-974,391
2017	-19,181,113	-799,213

*Notes:* Employment termination payment is a lump sum payment made as a result of the termination of a person’s employment.

*Source:* ATO (private communication).

As ALife is a 10% random sample of tax filers, it is subject to sampling error. We address this issue by reconciling top income outliers (defined here the top 100 individuals in terms of taxable gross income in Australia each year) with the ATO full population for which the ATO has provided us mean

income values (separately for the top 100 to 50 and the top 50 individuals). We average income for top 100 to 50 and top 50 individuals in Australia to adjust incomes of the top 10 and top 5 individuals in ALife. All income components are scaled up by a constant factor. This approach fixes the top 0.001% but sampling error may still affect income groups below the top 0.001%, with the issue likely to be more important the smaller the income group. In practice, top income shares for groups smaller than 0.1% of the population and above the top 0.1% may not be reliable.

## 8.1. Appendix A.2 Grossing-up factors

**Table A.2.1: Survey to National Account grossing-up factors (1991-2018)**

Year	Employee income	Mixed income	Non-pension capital income	Personal income tax	Cash benefits
1991	1.11	1.74	2.93	1.14	0.98
1992	1.10	1.40	3.18	1.05	1.12
1993	1.14	1.58	3.51	1.06	1.20
1994	1.17	1.54	3.76	1.12	1.28
1995	1.18	1.24	3.87	1.14	1.25
1996	1.18	1.40	3.24	1.18	1.25
1997	1.19	1.32	3.51	1.19	1.26
1998	1.17	1.56	3.24	1.16	1.25
1999	1.18	1.41	2.69	1.17	1.28
2000	1.16	1.56	3.15	1.16	1.29
2001	1.16	1.02	3.34	1.02	1.38
2002	1.16	1.83	3.43	1.12	1.39
2003	1.17	1.49	3.40	1.12	1.33
2004	1.17	1.64	3.25	1.12	1.45
2005	1.18	1.86	2.70	1.12	1.44
2006	1.19	1.85	2.68	1.19	1.52
2007	1.23	1.47	2.60	1.15	1.53
2008	1.27	2.00	2.43	1.23	1.65
2009	1.16	2.02	3.36	1.22	1.46
2010	1.16	2.17	3.14	1.16	1.36
2011	1.17	2.24	3.35	1.16	1.44
2012	1.19	2.12	3.07	1.17	1.46
2013	1.19	2.09	2.40	1.14	1.47
2014	1.19	2.09	2.48	1.11	1.46
2015	1.17	2.80	2.34	1.12	1.39
2016	1.14	2.56	1.88	1.09	1.39
2017	1.14	2.90	2.19	1.11	1.34
2018	1.12	2.50	2.01	1.10	1.30

Notes: Constant factors by which each income component has to be multiplied in the survey data (complemented by tax data for the top 1%) to restore consistency with National Account. For instance, a factor of 2 means that incomes have to be doubled.

## 8.2. Appendix A.3 Superannuation balance regression estimates

**Table A.3.1: Superannuation balance regression estimates (adults under 60 years of age)**

	2003		2005		2009		2011		2013		2015		2018	
	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.
Age	872***	66.3	992***	86.5	1,241***	80.3	1,496***	101.5	1,635***	109.4	2,266***	106.8	2,385***	122.3
Labour income (in \$1,000s)	-1,092***	69.1	-1,316***	77.1	-1,307***	61.7	-1,306***	68.5	-1,439***	69.8	-1,458***	70.2	-1,147***	67.1
Labour income (in \$1,000s) squared	-1.589***	0.15	-1.026***	0.07	-0.925***	0.04	-0.577***	0.04	-1.219***	0.05	-0.214***	0.01	-0.346***	0.02
Age x labour income (\$1,000s)	52.1***	1.6	58.5***	1.7	55.7***	1.4	57.0***	1.5	63.9***	1.6	56.4***	1.5	52.3***	1.4
Female	19,138***	3,023	16,908***	4,113	15,137***	3,729	20,744***	4,870	16,960***	5,171	23,683***	5,166	9,736*	5,901
Female x age	-755***	77.5	-617***	104.3	-618***	94.6	-770***	122.7	-632***	130.3	-946***	129.1	-463***	148.3
Zero labour income	9,600***	1,426	13,820***	1,854	7,525***	1,635	18,261***	2,215	12,849***	2,327	4,912**	2,209	9,153***	2,680
Constant	-23,828***	2,584	-27,555***	3,439	-29,054***	3,162	-39,385***	4,049	-41,409***	4,336	-46,995***	4,372	-54,929***	4,991
Sample size	17,491		14,800		23,666		20,896		19,783		23,072		18,883	
Adjusted R <sup>2</sup>	0.273		0.287		0.271		0.273		0.322		0.303		0.347	

Notes: Ordinary Least Square estimates. 'x' denotes interaction terms. \* p < 0.10, \*\* p < 0.05, \*\*\* p < 0.01.

**Table A.3.2: Superannuation balance regression estimates (adults over 59 years of age)**

	2003		2005		2009		2011		2013		2015		2018	
	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.
Age	-2,066***	351.3	-2,999***	496.8	-2,677***	391.1	-3,848***	577.8	-5,273***	570.1	-4,190***	492.3	-4,743***	609.2
Superannuation income (in \$1,000s)	30,463***	1793.7	19,297***	2282.3	17,273***	1676.9	11,652***	1765.3	7,598***	1632.4	11,144***	1478.7	24,516***	1431.4
Superannuation income (in \$1,000s) squared	-14.27***	2.66	5.69***	2.09	-4.00	3.95	-31.72***	1.65	-24.73***	1.21	-16.22***	2.16	-17.38***	0.89
Age x superannuation income (\$1,000s)	-358***	25.2	-202***	33.0	-132***	24.6	-7	25.8	26	23.7	-12	21.1	-186***	20.3
Female	-114,578***	30,515	-105,355**	42,529	-26,962	34,548	-93,112*	49,348	-83,279*	49,324	-35,430	43,155	-96,150*	53,620
Female x age	1,467***	434	1,369**	603	273	488	1,159*	705	1,054	704	482	607	1,347*	757
Zero labour income	461	4,461	18,713***	6,927	3,658	4,477	5,796	6,834	-12,831*	6,628	-22,210***	6,474	-17,068**	8,117
Labour income (in \$1000s)	1,804***	82	3,089***	146	2,027***	73	2,227***	93	1,841***	77	1,744***	84	2,035***	82
Sample size	4,792		4,386		10,265		7,307		7,479		10,811		7,952	
Adjusted R <sup>2</sup>	0.314		0.298		0.307		0.357		0.356		0.366		0.468	

Notes: Ordinary Least Square estimates. 'x' denotes interaction terms. \* p < 0.10, \*\* p < 0.05, \*\*\* p < 0.01.

### 8.3. Appendix A.4 Imputed rent regression estimates

**Table A.4.1: Weekly imputed rent regression estimates (2005-06)**

	Gross imputed rent				Net imputed rent			
	(1) Coef.	S.E.	(2) Coef.	S.E.	(1) Coef.	S.E.	(2) Coef.	S.E.
Tenure type (ref. is owner without a mortgage)								
Owner with a mortgage	-0.003	0.010			-0.198***	0.009		
Renter	-1.361***	0.095			0.450***	0.096		
Other	-0.048*	0.028			0.218***	0.023		
Has a mortgage	0.000		0.009	0.010			-0.274***	0.010
State of residence (ref. is NSW)								
VIC	-0.121***	0.012	-0.119***	0.012	-0.001	0.010	-0.003	0.010
QLD	-0.093***	0.012	-0.092***	0.012	0.004	0.010	0.006	0.010
SA	-0.205***	0.013	-0.204***	0.013	-0.046***	0.012	-0.049***	0.012
WA	-0.244***	0.013	-0.242***	0.013	-0.013	0.012	-0.017	0.012
Tas	-0.184***	0.016	-0.179***	0.016	-0.016	0.016	-0.021	0.016
ACT & NT	0.008	0.018	0.008	0.018	0.056***	0.013	0.057***	0.013
Area of residence (ref. is Capital city)								
Balance of State	-0.198***	0.009	-0.198***	0.009	-0.033***	0.008	-0.035***	0.008
Number of bedrooms	0.128***	0.005	0.131***	0.005	0.028***	0.004	0.025***	0.004
Household gross income decile (ref. is 1)								
2	-0.005	0.017	-0.001	0.017	-0.001	0.016	-0.004	0.016
3	-0.043**	0.017	-0.036**	0.017	-0.017	0.016	-0.026	0.016
4	-0.018	0.017	-0.019	0.017	-0.037**	0.016	-0.042***	0.016
5	0.001	0.018	0.000	0.018	-0.044***	0.016	-0.052***	0.016
6	0.004	0.018	0.005	0.018	-0.054***	0.016	-0.057***	0.016
7	-0.007	0.018	-0.011	0.018	-0.073***	0.016	-0.080***	0.016
8	0.015	0.018	0.012	0.019	-0.050***	0.016	-0.055***	0.016
9	0.051***	0.019	0.048**	0.019	-0.058***	0.016	-0.072***	0.016
10	0.109***	0.019	0.109***	0.019	-0.022	0.016	-0.042***	0.016
Landlord type (ref. is real estate agent)								
No landlord	3.962***	0.095	5.313***	0.013	2.311***	0.381	1.862***	0.379
State or territory housing authority	5.140***	0.020	5.167***	0.020	1.596***	0.391	1.590***	0.379
Parent	5.300***	0.034	5.307***	0.035	1.423***	0.392	1.420***	0.380
Other person	0.053***	0.019	0.058***	0.019	1.682***	0.419	1.621***	0.418
Other	3.941***	0.029	3.946***	0.029	1.492***	0.392	1.492***	0.380
Mortgage weekly repayments					-0.439***	0.005	-0.396***	0.006
Gross imputed rent					-1.636***	0.381	-1.170***	0.379
Sample size	9,857		9,857		9,857		9,857	
Adjusted R2	0.969		0.968		0.690		0.692	

Notes: Ordinary Least Square estimates. Model (1) is with tenure type, model (2) is without tenure type. In the net imputed rent models all variables are interacted with gross imputed rent, with the exception of mortgage weekly repayments and gross imputed rent. \* p < 0.10, \*\* p < 0.05, \*\*\* p < 0.01.



## Appendix B Supplementary results

**Table B.1: Mean pre-tax adult income by income group (current dollars, 1991-2018)**

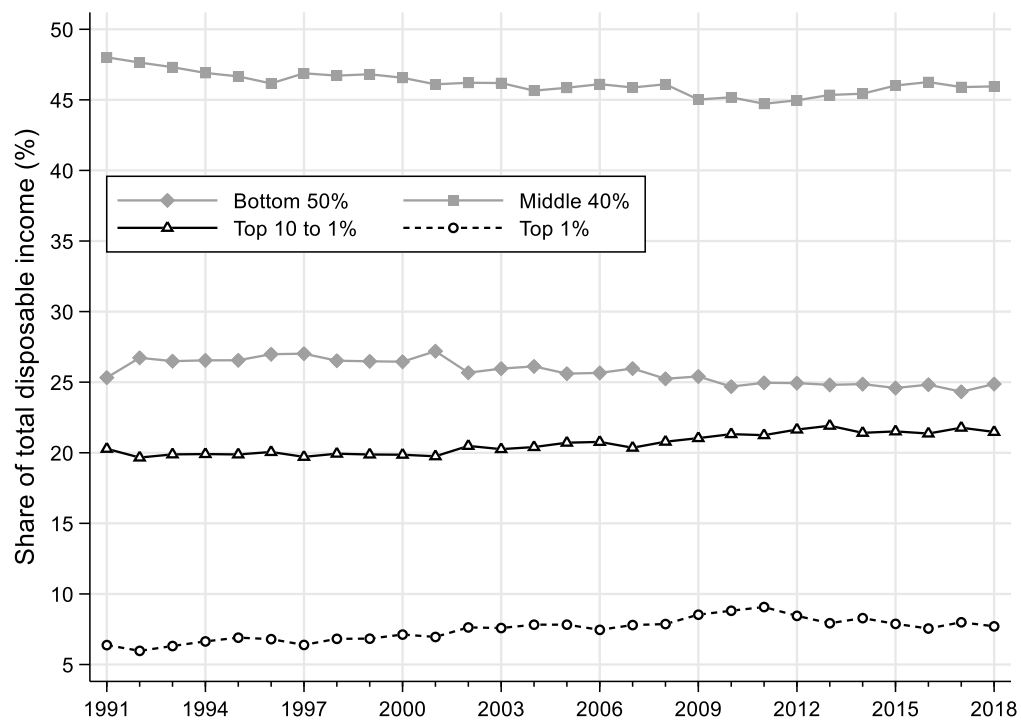
	Income group				All
	Bottom 50%	Middle 40%	Top 10% to 1%	Top 1%	
1991	10,197	30,406	58,550	162,852	24,141
1992	10,823	30,620	58,278	161,773	24,511
1993	11,337	32,337	62,629	182,177	26,037
1994	11,927	33,992	66,201	201,152	27,528
1995	12,532	35,053	69,946	195,753	28,523
1996	13,324	36,809	73,666	225,678	30,266
1997	13,812	38,918	76,736	221,114	31,538
1998	14,223	40,462	80,723	248,906	33,043
1999	14,561	42,077	83,216	269,915	34,283
2000	15,439	44,256	89,952	284,875	36,361
2001	16,834	46,084	92,993	308,709	38,301
2002	16,670	48,744	99,508	349,734	40,277
2003	17,251	50,659	105,894	379,418	42,199
2004	18,409	53,773	114,852	427,802	45,316
2005	18,658	56,123	121,404	455,394	47,248
2006	20,238	59,310	128,463	472,402	50,114
2007	22,497	63,609	137,018	558,410	54,534
2008	23,864	68,142	147,157	579,469	58,208
2009	23,765	69,375	157,164	634,094	60,103
2010	23,821	70,147	156,412	622,404	60,258
2011	25,685	75,252	168,774	689,945	65,029
2012	26,993	79,637	182,466	706,888	68,830
2013	27,259	81,324	187,210	685,060	69,827
2014	28,025	82,976	186,245	709,438	71,037
2015	27,842	84,142	185,418	671,095	70,946
2016	27,451	83,336	181,216	640,997	69,765
2017	29,186	88,220	195,828	712,084	74,601
2018	30,125	89,235	196,108	702,775	75,417

**Table B.2: Mean post-tax national adult income by income group (current dollars, 1991-2018)**

	Income group				All
	Bottom 50%	Middle 40%	Top 10% to 1%	Top 1%	
1991	15,131	27,911	47,320	123,522	24,221
1992	15,621	28,254	47,442	120,227	24,582
1993	16,318	29,867	51,326	136,344	26,079
1994	17,411	31,344	53,972	150,189	27,596
1995	18,227	32,333	55,481	159,939	28,630
1996	19,601	34,009	59,151	166,351	30,386
1997	20,590	35,840	60,493	163,170	31,694
1998	21,451	37,459	63,788	179,679	33,237
1999	22,352	38,890	65,559	187,608	34,487
2000	23,819	40,992	69,169	202,323	36,552
2001	25,090	42,992	73,715	214,010	38,511
2002	25,464	45,333	80,060	246,504	40,526
2003	27,016	47,331	82,365	252,877	42,380
2004	29,096	50,453	89,305	281,936	45,578
2005	30,174	52,799	93,993	291,275	47,576
2006	32,171	56,219	99,703	293,952	50,476
2007	35,292	60,995	106,681	333,825	54,975
2008	36,634	65,215	116,352	362,085	58,488
2009	36,442	66,421	125,183	428,971	60,319
2010	36,171	66,782	126,392	438,939	60,544
2011	39,284	71,313	135,625	483,792	65,201
2012	41,446	75,736	146,016	477,770	68,926
2013	42,121	77,541	149,839	455,603	70,114
2014	43,167	79,345	150,176	486,384	71,673
2015	43,100	80,091	149,701	459,942	71,645
2016	42,973	78,695	144,544	425,162	70,217
2017	45,552	83,598	156,873	479,191	75,107
2018	47,380	84,389	154,612	456,248	75,918

## Appendix C Results for post-tax disposable income

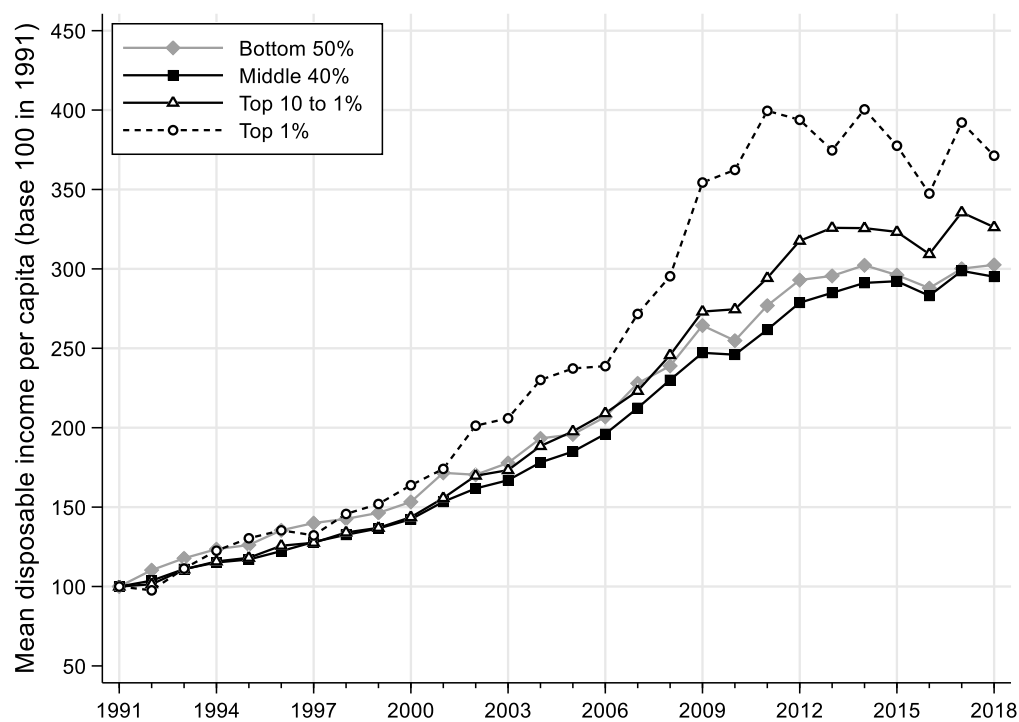
**Figure C.1: Income shares 1991-2018**



Notes: Distribution of post-tax disposable income (after all taxes and transfers) among adults. Broad equal-split adults series (household income equally split among adults).

*Stata graph net\_split2\_incs\_h\_a*

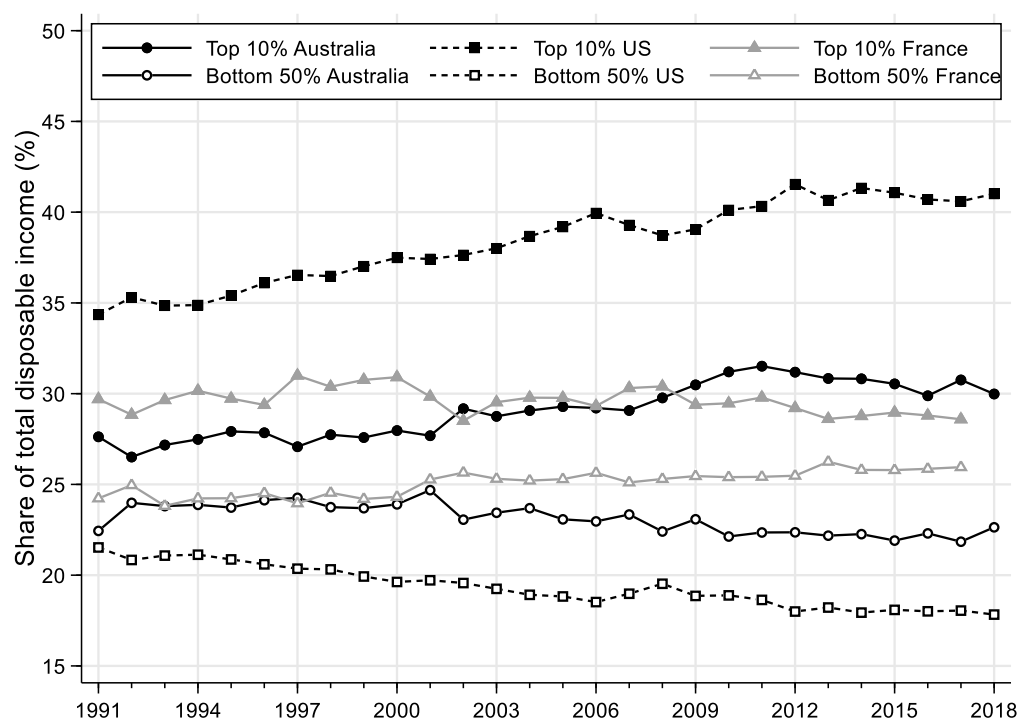
**Figure C.2: Mean adult income by income group 1991-2018**



Notes: Distribution of post-tax disposable income (before all taxes and transfers, except age pensions) among adults. Broad equal-split adults series (household income equally split among adults). Index based on mean incomes in current dollars.

*Stata graph net\_split2\_mean\_index\_a*

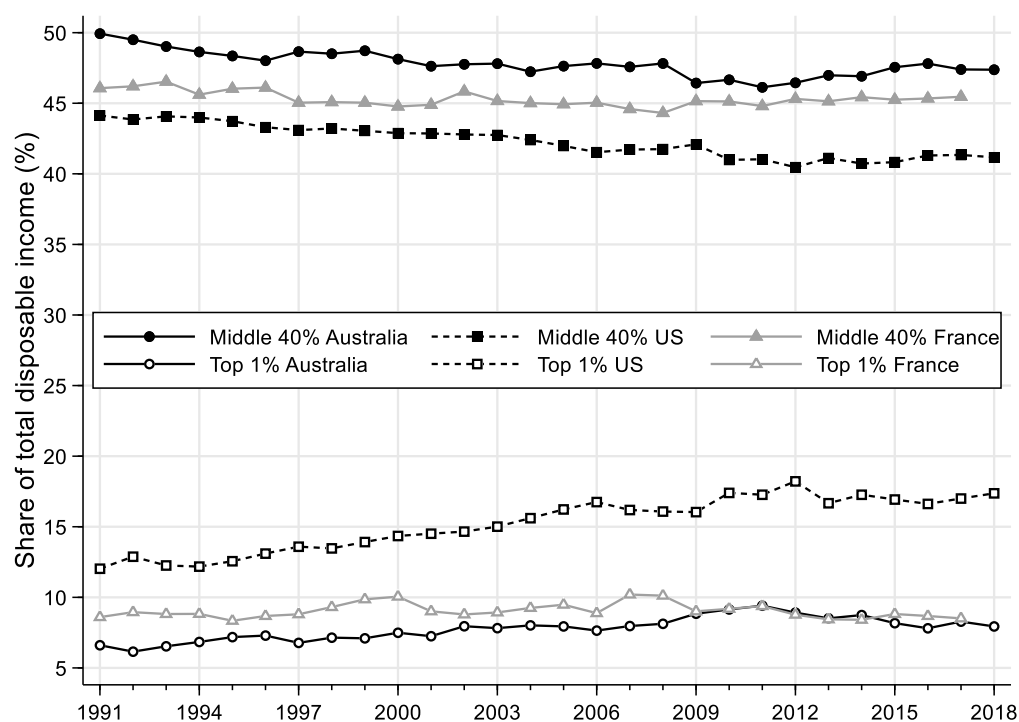
**Figure C.3: Top 10% and bottom 50% income shares: Australia, US and France 1991-2018**



Notes: Distribution of post-tax disposable income (after all taxes and transfers) among adults. Narrow-equal split adults series (income of married couples divided by two).

*Stata graph intcomp\_net\_sh1*

**Figure C.4: Top 1% and middle 40% income shares: Australia, US and France 1991-2018**



Notes: Distribution of post-tax disposable income (after all taxes and transfers) among adults. Narrow-split-adults series (income of married couples divided by two).

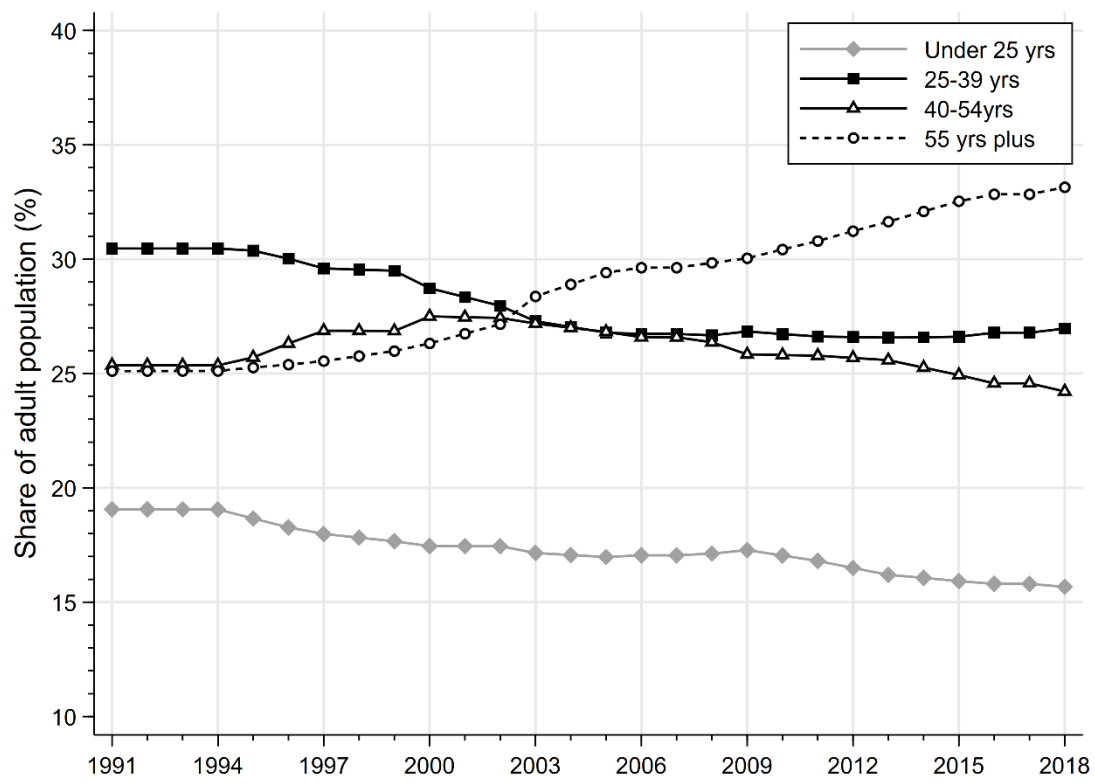
*Stata graph intcomp\_net\_sh2*

**Table C.1: Mean post-tax disposable adult income by income group (current dollars, 1991-2018)**

	Income group				All
	Bottom 50%	Middle 40%	Top 10% to 1%	Top 1%	
1991	9,327	22,107	41,517	117,719	18,417
1992	10,291	22,925	42,113	114,898	19,252
1993	10,986	24,536	45,995	131,012	20,747
1994	11,522	25,454	48,082	144,299	21,706
1995	11,765	25,871	49,019	153,476	22,167
1996	12,630	27,037	52,179	159,379	23,415
1997	13,055	28,304	52,957	155,635	24,159
1998	13,301	29,309	55,637	171,529	25,087
1999	13,652	30,190	56,859	178,908	25,787
2000	14,297	31,470	59,647	192,801	27,030
2001	16,004	33,906	64,628	204,923	29,424
2002	15,890	35,759	70,486	236,930	30,952
2003	16,587	36,902	71,936	242,448	31,951
2004	18,028	39,384	78,236	270,868	34,510
2005	18,264	40,889	82,083	279,365	35,667
2006	19,290	43,338	86,822	281,070	37,594
2007	21,260	46,963	92,650	319,793	40,944
2008	22,282	50,862	102,000	347,733	44,135
2009	24,656	54,635	113,397	417,185	48,533
2010	23,769	54,381	113,990	426,537	48,142
2011	25,825	57,854	122,167	470,334	51,743
2012	27,327	61,617	131,897	463,651	54,806
2013	27,571	62,990	135,288	441,052	55,563
2014	28,194	64,373	135,203	471,410	56,701
2015	27,619	64,610	134,219	444,461	56,163
2016	26,855	62,577	128,426	409,045	54,099
2017	27,995	66,040	139,315	461,633	57,550
2018	28,219	65,227	135,450	437,086	56,756

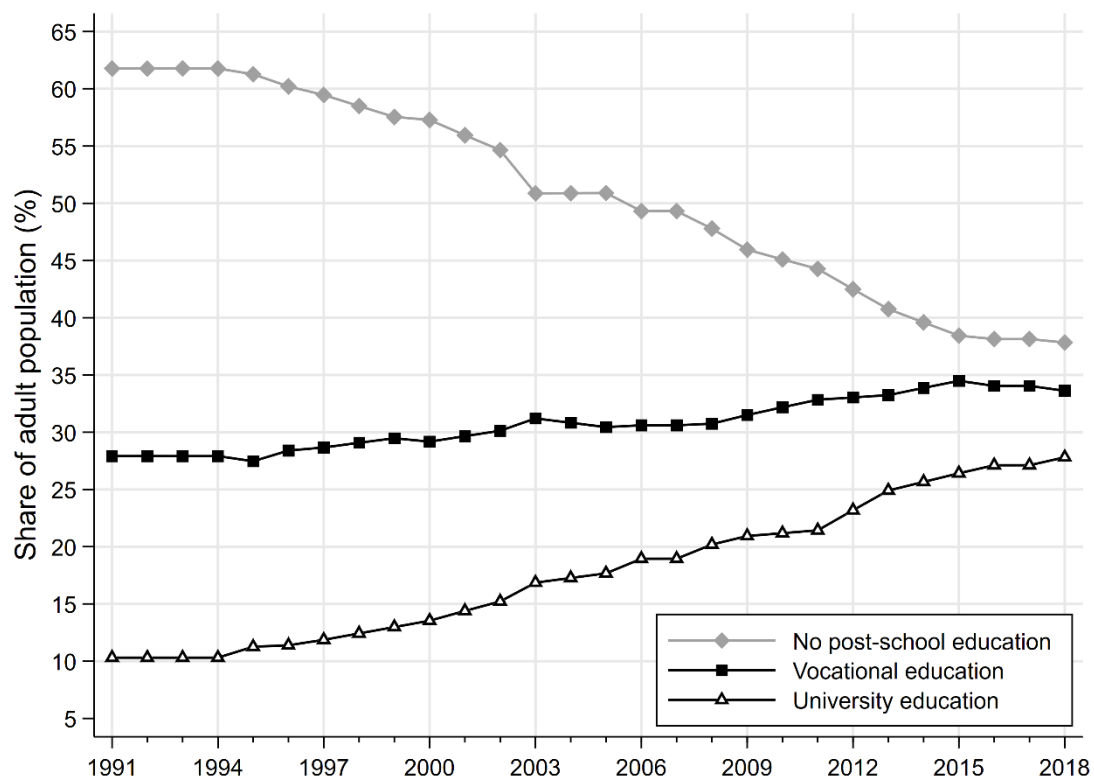
## Appendix D Population shares by demographic group

Figure D.1: Adult population shares by age group (1991-2018)





**Figure D.2: Adult population shares by education attainment (1991-2018)**



**Figure D.3: Adult population shares for natives and immigrants (1991-2018)**

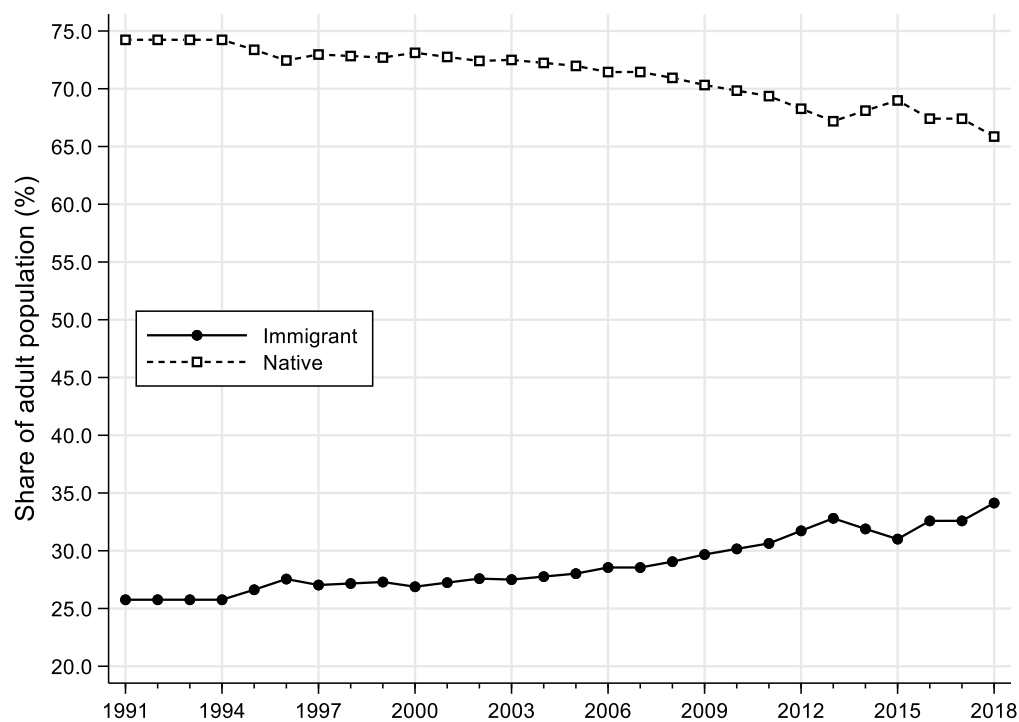
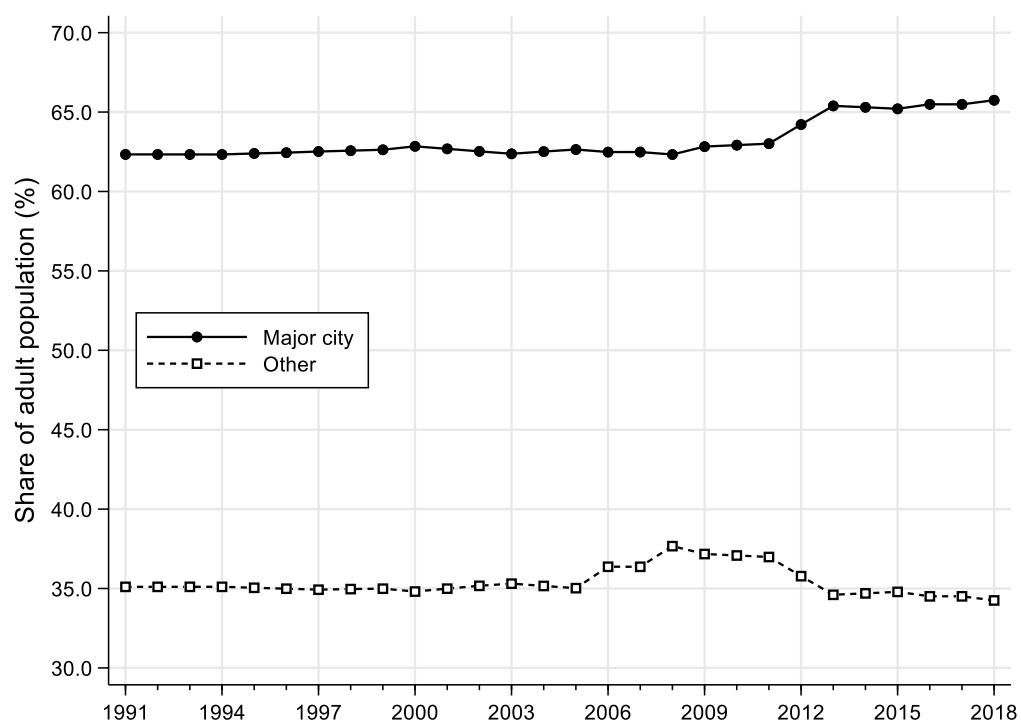
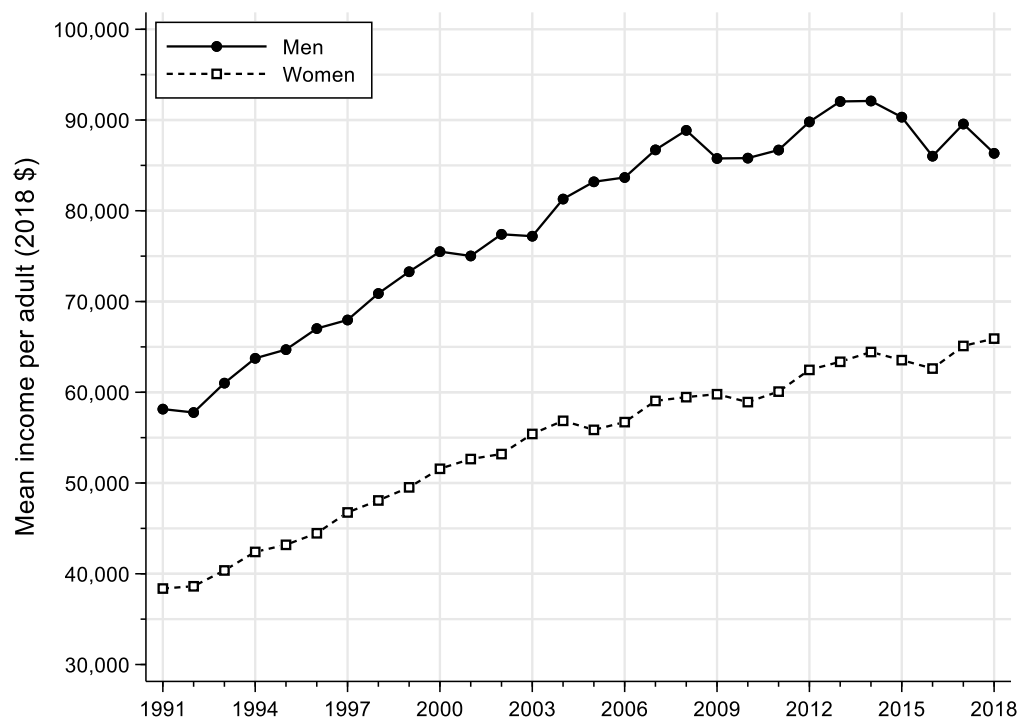


Figure D.4: Adult population shares by area of residence (1991-2018)



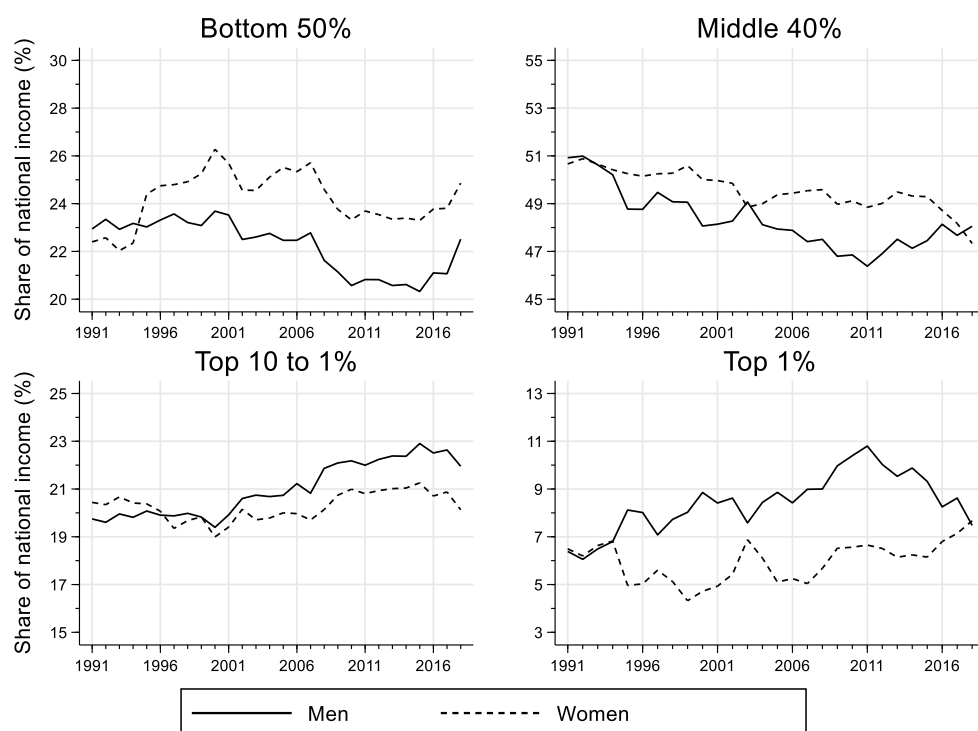
## Appendix E Inequality by demographic group: Individualistic series

Figure E.1: Mean post-tax national income per adult by sex (1991-2018)



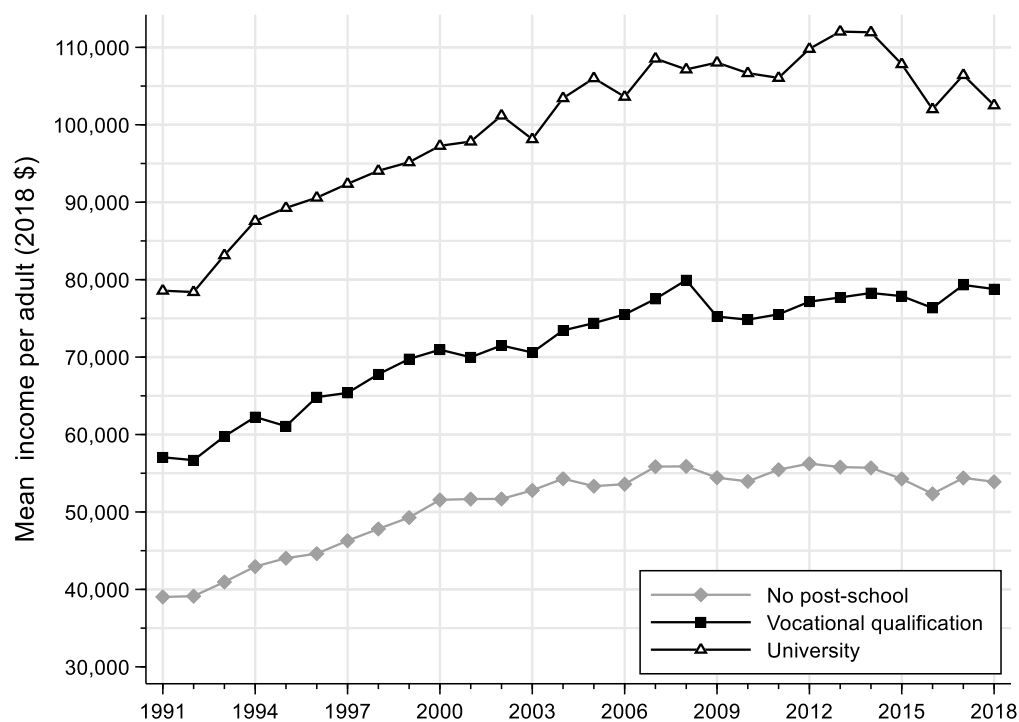
*Stata graph ptninc\_mean\_split0\_sex\_b*

**Figure E.2: Post-tax national income shares among men and women (1991-2018)**



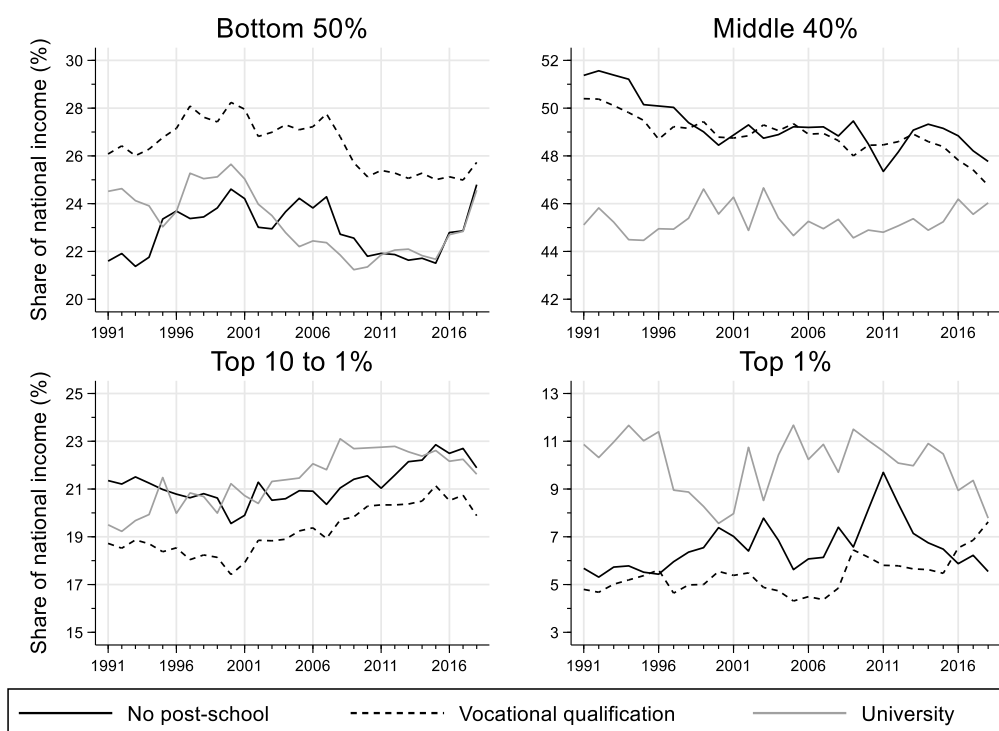
*Stata graph pshares\_split0\_sexp\_ptninc\_b*

**Figure E.3: Mean post-tax national income per adult by educational attainment (1991-2018)**



*Stata graph ptninc\_mean\_split0\_edu\_b*

**Figure E.4: Post-tax national income shares within education groups (1991-2018)**



*Stata graph pshares\_split0\_edu\_ptninc\_b*