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The Gini coefficient and personal inequality measurement

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# The Gini Coefficient and Personal Inequality Measurement

by

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# The Gini Coefficient and Personal Inequality Measurement

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May 13, 2016

#### **Abstract**

The Gini coefficient is the most popular inequality index. It is based on the sum of pairwise absolute income differences, which can be viewed as taking a separate sum for each individual of the differences between his/her income and others', and then adding up those separate sums. The differences vis-à-vis people with lower income can be used to construct an individual's *advantage*, while the differences with respect to people with higher incomes generate the individual's *deprivation*. Deprivation and advantage can be weighted differently, producing a whole family of "Gini admissible" personal inequality indexes. The population average of any one of the latter equals the Gini coefficient. The properties of the personal inequality indexes explain the sensitivity of the Gini coefficient to transfers in different ranges of the income distribution. They also throw light on individual views of secular changes in income distribution interesting for their own sake. For example, throughout the change from a traditional to a modern economy that gives rise to the Kuznets curve, those in the traditional sector believe that inequality is constantly increasing while those in the modern sector believe the opposite. Personal views about polarization and rising inequality, as seen in most high income countries in recent decades, are also illuminated.

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#### I. Introduction\*

The Gini coefficient has a natural interpretation as the mean of personal inequality assessments. While that fact is fairly obvious, it was not emphasized in the original work by Gini (1914) and has not been highlighted since. This paper shows that this straightforward interpretation throws important light on the properties of the Gini coefficient. It also allows us to better understand individual reactions, as well as that of the Gini coefficient, to secular changes in income distribution. The latter include the transition from a traditional to a modern economy analyzed by Kuznets (1955), and the polarization and rising inequality seen in recent decades in the U.S. and many other countries. Personal assessments of even the direction of change in inequality may differ between people at different income levels. These results suggest that our understanding of inequality measurement can be enriched by studying what it may mean at the personal level.

The Gini coefficient can be defined or interpreted in many ways (Yitzhaki, 1998). For our purposes the most useful is that it equals one half the mean difference divided by the mean. For a finite population, the Gini coefficient can be found by taking the sum of all all pairwise absolute income differences, S, converting to an average and normalizing by the mean. S can be written as the sum across individuals i =1, ..., n of their individual sums of pairwise differences with all other individuals,  $S_i$ . The latter can be used as the basis for a personal inequality index whose average across the population is the Gini coefficient. For each individual,  $S_i$  is composed of the sum of differences with higher incomes plus the sum of differences with lower incomes. Following Yitzhaki (1979) the sum of differences with higher incomes may be used to define the individual's deprivation. That concept is complemented by the individual's advantage, derived from the sum of differences with respect to lower incomes. Summing deprivation or advantage across the whole population produces the same total (Yitzhaki, 1979). An implication is that a weighted average of deprivation and advantage, as well as an unweighted average, will generate a personal inequality index that will equal the Gini coefficient when averaged across the population. This means that there is a whole family of "Gini admissible" personal inequality indexes or GAPIIs. If societies choose to base overall inequality measurement on an average of individual assessments they may all use the same inequality index, that is the Gini coefficient, at the aggregate level even if they differ in the weight their members place on advantage vs. deprivation. <sup>2</sup>

The personal inequality indexes discussed here may be regarded from a "top down" or "bottom up" viewpoint. A GAPII could be interpreted as showing how a social planner would measure inequality at the personal level. This is a "top down" view. An alternative, "bottom up", view is that *individuals*, for whatever reason, assess inequality using a GAPII. Why might individuals do so? One possibility is that

<sup>\*</sup>Thanks are due to Michael Hoy, Stephen Jenkins and Shlomo Yitzhaki for helpful comments on an earlier draft of this paper. Responsibility for any errors or omissions is of course my own.

<sup>&</sup>lt;sup>1</sup> Yitzhaki (1979) used the term "relative deprivation", which was introduced by Runciman (1966) to refer to any case in which some members of a reference group felt deprived compared to other members of their group. "Deprivation" is used here simply because it is shorter. Fehr and Schmidt (1999) referred to the same concept as "disadvantageous inequality", but the term deprivation still dominates in the literature. Yitzhaki (1979) used the term "satisfaction" rather than "advantage". "Advantage" is used here as a more neutral term.

<sup>&</sup>lt;sup>2</sup> It may seem too strong to assume that all individuals in a society would place the same weight on advantage vs. deprivation. With continuous income distributions this assumption could be relaxed to allow weights to differ across individuals as long as those differences were independent of income.

they could have interdependent utility functions, such as that proposed by Fehr and Schmidt (1999), which suggest the use of a GAPII. But one may also appeal to bounded rationality. The difference between incomes is an unobjectionable indicator of inequality between two people (especially if considered in the light of mean income). Although we know there are many alternatives, people may simply think, by extension, that the average of such differences provides the natural basis for measuring inequality when there are more than two people. That conclusion could be reinforced by information and computing constraints. As shown in this paper, in order to compute the value of a GAPII the individual only needs to know the fraction of the population with income above him and the average incomes of those above and below him. While it is not reasonable to suppose that each individual knows everyone's income, he/she might be able to make a serviceable guess at these three quantities.

This paper is related to the large literature on individual attitudes toward inequality. A portion of the literature attempts to measure attitudes within narrow reference groups, e.g. co-workers or members of the same occupation. In that context people tend to be averse to deprivation but to like advantage. As Clark and D'Ambrosio (2015) point out, in the income distribution literature the usual reference group is broader. In that context, following Yitzhaki (1979, 1982) and Fehr and Schmidt (1999) the general expectation has been that people will be averse to both deprivation and advantage. There are now a few empirical and experimental studies that have estimated aversion to deprivation and/or advantage with broader reference groups. Using the German SOEP survey data, D'Ambrosio and Frick (2007) find strong aversion to deprivation (but do not report on attitudes to advantage). Cojocaru (2014) finds significant aversion to both advantage and deprivation using a survey of 27 transition countries. In experiments with subjects who played a sequential public goods game, Teyssier (2012) found that 40% were averse to both advantage and deprivation while 18% were averse to neither. While these studies do not indicate a difference in aversion to deprivation vs. advantage, neural studies find that brain activity reacts more strongly to deprivation and some authors presume that aversion to advantage is likely weaker than aversion to deprivation (Clark and D'Ambrosio, 2015).

The remainder of the paper proceeds as follows. For expositional simplicity we start by working with the case in which advantage and deprivation are equally weighted. Section II defines the personal inequality index and derives some of its basic properties. In Section III we then explore how the behavior of this index helps to explain the sensitivity of the Gini coefficient to income changes in different ranges of a distribution. The analysis is extended to allow unequal weighting of deprivation and advantage in Section IV, which shows that the main insights of the previous two sections survive this generalization. How the personal assessments of inequality vary with income is discussed in Section V and the behavior of those assessments during period of secular change in income distribution is examined in Section VI. Section VII concludes.

#### II. Gini-admissible Personal Inequality Indexes: Base Case

In this section we see how the Gini coefficient can be defined as the average value across individuals of a particular personal inequality index (PII), and begin to examine the properties of the latter. We do not seek a basis for the PII in individuals' personal or social preferences. Our interest is confined to investigating the implications for personal inequality assessments if individuals use a Gini admissible PII,

or GAPII. A PII will be termed Gini admissible if the Gini coefficient can be found by taking a simple average of the values of that PII across individuals.

The Gini coefficient for an income distribution equals one half the mean difference divided by the mean, as we see in:

(1) 
$$G = \frac{1}{2n^2\bar{y}} \sum_{i=1}^n \sum_{j=1}^n |y_i - y_j| = \frac{S}{2n^2\bar{y}}$$

where  $y_i$  is the income of individual  $i, \bar{y}$  is mean income,  $n > 1, y_1 \le y_2 \le \cdots \le y_n$ , S is the sum of differences, and  $S/n^2$  is the mean difference.<sup>3</sup>

A natural but previously overlooked interpretation is that G is the mean value across individuals of a particular GAPII,  $G_i$ :

$$(2) \qquad G = \frac{1}{n} \sum_{i=1}^{n} G_i$$

where

(3) 
$$G_i = \frac{1}{2n\bar{y}} \sum_{j=1}^{n} |y_i - y_j| = \frac{S_i}{2n\bar{y}}$$

and  $S_i$  is the sum of differences for individual i. Equation (3) can be rewritten:

$$(4) \qquad G_i = \frac{1}{2n\bar{y}} \left[ n_i^l \left( y_i - \bar{y}_i^l \right) + n_i^h \left( \bar{y}_i^h - y_i \right) \right]$$

where  $n_i^l$  is the number of individuals with income less than or equal to  $y_i$ , excluding individual i, and  $n_i^h$  is the number with income strictly greater than  $y_i$ , so that  $n_i^l + n_i^h = n - 1$ .<sup>4</sup>  $\bar{y}_i^l$  and  $\bar{y}_i^h$  are mean income among those with income less than or equal to  $y_i$ , excluding i, and strictly greater than  $y_i$  respectively.

Let  $H_i$  be the set of all j such that  $y_j > y_i$ , and  $L_i$  be the set of all j excluding i such that  $y_j \le y_i$ . Equation (4) can be expressed as:

$$(4') G_i = \frac{1}{2\bar{y}}(A_i + D_i)$$

where:

<sup>&</sup>lt;sup>3</sup> As mentioned earlier, the Gini coefficient can be expressed in many different ways (Yitzhaki, 1998). This is one of the two principal forms in which it was originally set out in Gini (1914), and is the most convenient for our discussion

<sup>&</sup>lt;sup>4</sup> The choice to include individuals who have the same income as *i* in the lower group rather than in the higher group is arbitrary but does not affect the results in any significant way.

(5*i*) 
$$A_i = \frac{n_i^l}{n} (y_i - \bar{y}_i^l) = \frac{1}{n} \sum_{j \in L_i} (y_i - y_j)$$

(5*ii*) 
$$D_i = \frac{n_i^h}{n} (\bar{y}_i^h - y_i) = \frac{1}{n} \sum_{j \in H_i} (y_j - y_i)$$

 $D_i$  is the discrete analogue of the measure of relative deprivation for an individual, which we will refer to simply as deprivation, proposed by Yitzhaki (1979) for a continuous distribution. It equals the average shortfall of i's income below the income of those who are better off, weighted by the fraction of the population in the latter group. Equation (4') shows that  $G_i$  is the simple average of  $D_i$  and a complementary measure,  $A_i$ , normalized by the mean. We will say that  $A_i$  represents individual i's advantage compared to people with lower income. Thus from the individual perspective inequality consists of both deprivation with respect to the better off and advantage over the worse off.

While  $G_i$  is a natural personal inequality index to associate with the Gini coefficient, it is not the only GAPII. As mentioned earlier, and as shown in Section IV, one can define a more general class of GAPIIs that are based on a weighted average of  $A_i$  and  $D_i$ .  $G_i$  is a special case in which the weights on  $A_i$  and  $D_i$  are equal.

From (4) we have:

**Proposition 1:**  $G_i$  is insensitive to a transfer of income within  $H_i$  or within  $L_i$  if the composition of neither group changes as a result of the transfer.

The proposition follows from the fact that transfers of income confined either to  $H_i$  or  $L_i$  do not alter  $n_i^l$ ,  $\bar{y}_i^l$ ,  $n_i^h$ , or  $\bar{y}_i^h$  or any other term on the right-hand side of (4). In terms of (4'), as noted by Yitzhaki (1979) these transfers have no effect on advantage,  $A_i$ , or on deprivation,  $D_i$ . The insensitivity of  $G_i$  to such transfers means that it does not respect the Pigou-Dalton principle of transfers, which is a cornerstone of the theory of aggregate inequality measurement. That an aggregate index that respects the Pigou-Dalton principle can be built on the basis of personal indexes that violate the principle is striking.

Sensitivity of  $G_i$  to a transfer of income between  $H_i$  and  $L_i$ 

What determines how sensitive  $G_i$  is to a transfer of income between  $H_i$  and  $L_i$ ? Consider the transfer of a total amount R from  $H_i$  to  $L_i$ . Note that such a transfer reduces both  $A_i$  and  $D_i$  by R/n, as can be seen from (5) where  $n_i^l(y_i - \bar{y}_i^l)$  and  $n_i^h(\bar{y}_i^h - y_i)$  both fall by R. We will allow R to be negative, so this also handles the case of transfers from  $L_i$  to  $H_i$ , which increase  $A_i$  and  $D_i$  by equal amounts. Using

$$\frac{\partial A_i}{\partial R} = \frac{\partial D_i}{\partial R} = \frac{-1}{n}$$

and from (4') we have:

$$(6) \quad \frac{\partial G_i}{\partial R} = -\frac{1}{n\bar{y}}$$

which allows us to state:

**Proposition 2:** When income is transferred from a person with income strictly above  $y_i$  to someone with income strictly below  $y_i$ ,  $G_i$  falls, while if income is transferred from a person with income strictly below  $y_i$  to someone with income strictly above  $y_i$ ,  $G_i$  rises. In both cases the change in  $G_i$  is proportional to the amount transferred and independent of  $y_i$ .

Note that this proposition implies that any given individual is equally sensitive to a transfer from the group above him to the group below, or vice versa. In that sense, individuals are equally sensitive to redistribution that does not alter their own income.

Sensitivity of  $G_i$  to a transfer affecting  $y_i$ 

We also need to analyze those cases where distributional changes affect individual i's own income. There are two situations to consider. One is that of a transfer from i to another person j. The other is that of a transfer from j to i. We will consider them in turn. In this analysis, and in the remainder of the paper unless indicated otherwise, we will assume  $y_1 < y_2 < \cdots < y_n$ . This assumption will simplify the analysis since, for example, it implies that when n is odd there is a unique individual with median income,  $y^{med}$ , and half the remaining population has  $y_i < y^{med}$  while the other half have  $y_i > y^{med}$ . If n is even there is no individual with  $y_i = y^{med}$ , but  $y^{med}$ , which is defined as the midpoint between  $y_{n/2}$  and  $y_{n/2+1}$ , again divides the population into two sub-populations of equal size with incomes above and below the median.

<u>Transfer from i to j</u>: Let  $y_i^o$  and  $y_j^o$  be initial incomes and consider the effect on  $G_i$  of the transfer of a small amount r from individual i to individual j. From (4) we obtain:

**Proposition 3a:** The effect on  $G_i$  of a small transfer in the amount of r from individual i to an individual j is given by:

(7i) 
$$\Delta G_i = \frac{1}{2n\bar{y}} \left[ \left( n_i^h - n_i^l \right) - 1 \right] \mathbf{r}, \quad i > j$$

(7*ii*) 
$$\Delta G_i = \frac{1}{2n\bar{y}} \left[ \left( n_i^h - n_i^l \right) + 1 \right] \mathbf{r}, \quad i < j$$

If we could ignore the -1 and +1 in the square brackets on the right-hand side, (7) would say that irrespective of whether i was greater or less than j, a transfer from i to anyone else would increase  $G_i$  if i was below the median and reduce  $G_i$  if i was above the median. This reflects the fact that the main impact of the transfer on  $G_i$  is to reduce  $A_i$  and increase  $D_i$ . If  $n_i^h > n_i^l$ , individual i is below the median and from (5) we see that the increase in  $D_i$  will exceed the drop in  $A_i$ , since those changes are proportional to  $n_i^h$  and  $n_i^l$  respectively. If  $n_i^h < n_i^l$ , individual i is above the median and we have the opposite case. The -1 in (7i) means that the rank at which  $\Delta G_i$  switches from being positive to negative as we go up the income scale in the i > j case is one position higher than it would otherwise be, since the transfer is going to a person with income lower than the "donor" i, which reduces  $\bar{y}_i^l$  and  $A_i$  a little. And

<sup>&</sup>lt;sup>5</sup> If we assume only  $y_1 \le y_2 \le \cdots \le y_n$  then there could be multiple individuals with median income and the groups with income strictly below the median and strictly above the median need not contain an equal number of members. Consider for example a population with the set of incomes (1, 1, 2, 2, 2, 3).

the +1 in (7ii) means that when i < j,  $\Delta G_i$  switches from positive to negative one position *lower* than would otherwise be the case since the transfer goes to a higher income person, raising  $\bar{y}_i^h$  and  $D_i$  a little.

<u>Transfer from j to i:</u> Here incomes after a transfer are  $y_i^o + r$  and  $y_i^o - r$ . and we have:

**Proposition 3b:** The effect on  $G_i$  of a small transfer in the amount of r from an individual j to individual i is given by:

(8i) 
$$\Delta G_i = \frac{1}{2n\overline{y}} \left[ \left( n_i^l - n_i^h \right) + 1 \right] \mathbf{r}, \quad i > j$$

(8*ii*) 
$$\Delta G_i = \frac{1}{2n\overline{y}} [(n_i^l - n_i^h) - 1] \mathbf{r}, \quad i < j$$

Now the main effect of the transfer is to raise  $y_i$  and therefore to increase  $A_i$  and reduce  $D_i$ , which is equalizing if  $y_i$  is below the median and disequalizing if  $y_i$  is above the median. Again the point at which  $\Delta G_i$  switches sign as i rises is offset one position by the small impact of the change in  $y_j$  on  $A_i$  when i > j and on  $D_i$  when i < j.

Summing up, we can say, somewhat loosely, that an individual perceives a small transfer from himself to someone else as equalizing if his income is above the median, and as disequalizing if his income is below the median. If he is the *recipient* he finds a small transfer equalizing if he is below the median and disequalizing if he is above the median. Thus the situation in Gini-admissible personal inequality measurement is quite different from that in the familiar aggregate inequality measurement. In the latter, the impact of a small transfer on inequality is deemed equalizing if the donor's income exceeds the recipient's and disequalizing if the opposite holds. In the case of Gini-admissible personal inequality measurement, in contrast, whether the transfer is considered equalizing or disequalizing depends almost solely on the income of the person making the assessment. Low income people find making a transfer disequalizing and receiving a transfer equalizing. High income people find the opposite.

# III. Explaining the sensitivity of the Gini coefficient to changes in different ranges of the income distribution

From (1) one may derive:

(9) 
$$G = \frac{2}{n^2 \bar{y}} [y_1 + 2y_2 + 3y_3 + \dots + ny_n] - \frac{n+1}{n}$$

(see e.g. Cowell, 2011, p. 114). This provides insight into the sensitivity of the Gini coefficient to changes in different ranges of the income distribution. Consider a small transfer, r, from individual j to individual i where i < j. This is an example of what would be called an "equalizing transfer" in discussions of aggregate inequality. From (9), this transfer will produce a change in the Gini coefficient given by:

(10) 
$$\Delta G = \frac{-2r(j-i)}{n^2 \bar{y}}$$

which also tells us the impact of a transfer from i to j, in which case r < 0. We see that the impact on the Gini coefficient does not depend on  $y_i$  or  $y_j$ , but varies only with r and the difference in income ranks between i and j.

The fact that the sensitivity of the Gini coefficient to transfers is independent of the incomes of the transferor and transferee, but depends on the number of people between them in the distribution is one of the most interesting properties of the Gini coefficient. This property follows directly from those of the personal inequality index  $G_i$  captured in Propositions 1, 2 and 3 above. Proposition 1 implies:

(11*i*) 
$$\Delta G_k = 0$$
.  $k < i, k > j$ .

From Proposition 2 we have:

$$(11ii) \qquad \Delta G_k = \frac{-r}{n\bar{\nu}}. \qquad i < k < j.$$

And from Proposition 3

(12) 
$$\Delta G_i = \frac{(n_i^l - n_i^h - 1)r}{2n\bar{y}}.$$
  $\Delta G_j = \frac{(n_j^h - n_j^l - 1)r}{2n\bar{y}}.$ 

Now, from (1) and (11i), the change in G resulting from a transfer from j to i is given by:

(13) 
$$\Delta G = \frac{1}{n} (\Delta G_i + \Delta G_j + \sum_{k=i+1}^{j-1} \Delta G_k)$$

Note first that

(14) 
$$\sum_{k=i+1}^{j-1} \Delta G_k = -(j-i-1) \frac{r}{n\bar{y}}$$

which is proportional to the number of people between i and j, that is the number of people the transfer from j to i "passes over".

Next, to complete the analysis of  $\Delta G$ , note from (12) that:

$$\Delta G_i + \Delta G_j = \frac{(n_i^l - n_i^h - 1)r}{2n\bar{y}} + \frac{(n_j^h - n_j^l - 1)r}{2n\bar{y}}$$
$$= \frac{-r}{2n\bar{y}} [(n_j^l - n_i^l) + (n_i^h - n_j^h) + 2]$$

Since  $n_j^l - n_i^l$  and  $n_i^h - n_j^h$  both equal j - i we have:

(15) 
$$\Delta G_i + \Delta G_j = \frac{-r}{n\bar{\nu}}(j-i+1)$$

Hence, like  $\Delta G_k$ ,  $\Delta G_i + \Delta G_j$  is proportional to the size of the transfer and rises linearly with the number of people between i and j. In this case the reason for dependence on the number of people between i and j is that the effects of the transfer cancel out for  $A_i$  and  $A_j$  on the one hand, and for  $D_i$  and  $D_j$  on the other, where the sums they are based on overlap. The range of overlap includes all k < i for  $A_i$  and  $A_j$ , and all k > j for  $D_i$  and  $D_j$ . The range where effects that do not cancel out has j - i + 1 people in it.

Summing up, substituting (14) and (15) into (13) we have:

(16) 
$$\Delta G = \frac{-r}{n^2 \bar{v}} [(j-i+1) + (j-i-1)] = \frac{-2r(j-i)}{n^2 \bar{v}}$$

So we have shown that the mean of the effects on the personal inequality indexes resulting from the transfer equals the change in G that one would expect from aggregate inequality analysis.

The purpose of this exercise has been to show that the effects of a transfer on personal inequality explain the impact on G. That the reaction of G is governed by the number of people between transferor j and transferee i is due to two things: (i) aside from i and j themselves, the only people who care about the transfer are the individuals between them in the distribution, and (ii) the effects of the transfer on  $G_i$  and  $G_j$  cancel out except for those based on changes in income gaps between i or j and individuals in the range (i+1, j-1).

### IV. Unequal Weighting of Deprivation and Advantage

Yitzhaki (1979) defined relative deprivation for a society as a whole, D, as the average of individual deprivation indexes  $D_i$ . He worked with continuous distributions. The corresponding relationship with a discrete income distribution is:

(17) 
$$D = \frac{1}{n} \sum_{i=1}^{n} D_i$$

We can define overall advantage in a parallel way as:

(18) 
$$A = \frac{1}{n} \sum_{i=1}^{n} A_i$$

Yitzhaki shows that D is related to the Gini coefficient according to:

$$(19) G = \frac{D}{\bar{y}}$$

This result might appear puzzling, given that, from (4'),  $D_i$  represents only part of an individual's contribution to  $G_i$  and therefore to G. The explanation is as follows. The Gini coefficient is proportional

<sup>&</sup>lt;sup>6</sup> Note that the right-hand-side of (15) is not *proportional* to the number of people between i and j, which is j - i - 1

to the sum of differences, S. We can arrange the pairwise differences  $|y_i - y_j|$  making up S in a matrix M with i indexing rows and j indexing columns. D is the mean of the above-diagonal elements of M while A is the mean of the below-diagonal elements. Now, the above-diagonal elements have the same mean as the below-diagonal elements in M, since e.g.  $|y_2 - y_1| = |y_1 - y_2|$ . Hence A = D. To get from D to S we must therefore double D and multiply by  $n^2$  (to go from an average to a sum). The same procedure could be used to generate S from A. Thus we have  $S = 2n^2D = 2n^2A$  or:

$$(20) \qquad A = D = \frac{S}{2n^2}$$

Substituting the expression for D from (20) into (19) we obtain  $G = S/(2n^2\bar{y})$ , that is equation (1).

While Yitzhaki's approach and ours are closely related, his  $D_i$  and our  $G_i$  are distinct.  $G_i$  depends not just on deprivation,  $D_i$ , but also on advantage,  $A_i$ . While, overall, A = D, at the individual level there is no such relationship.  $A_i$  rises and  $D_i$  falls as we move up through the income distribution from  $y_1$  to  $y_n$ , and they do so at rates that rise or fall depending on the shape of the particular income distribution being examined.

The fact that A = D has important consequences for our personal inequality indexes. Using (19) and A = D, G may be found by taking a weighted average of A and D, as in:

(21) 
$$G = \frac{\lambda A + (1 - \lambda)D}{\overline{y}} \qquad 0 \le \lambda \le 1$$

where we require the weights to be positive. This in turn reveals that there is a family of Gini admissible personal inequality indexes or GAPIIs of the form:

(22) 
$$G_i^{\lambda} = \frac{\lambda A_i + (1 - \lambda) D_i}{\bar{v}} \qquad 0 \le \lambda \le 1$$

Hence, while  $\lambda$  may differ across societies, they can nevertheless agree on using G as an aggregate measure of inequality. <sup>7</sup> In the continuous case this result could be generalized to allow  $\lambda$  to differ across individuals, as long as the distribution of  $\lambda$  was independent of individual income.

We may ask which of the results derived above for the  $\lambda = \frac{1}{2}$  case survive once  $\lambda \neq \frac{1}{2}$  is allowed. Proposition 1, which says that the  $G_i$  are insensitive to transfers entirely within the  $H_i$  or  $L_i$  comparator groups, survives. The principle is not affected by re-weighting income differences with the  $H_i$  and  $L_i$  groups via  $\lambda \neq \frac{1}{2}$ . Proposition 2, which says that when income is transferred from those with income above (below)  $y_i$  to those with income below (above)  $y_i$  the fall (rise) in  $G_i$  is proportional to the total amount transferred, R, and is independent of  $y_i$  is also unaltered because we still have:

$$\frac{\partial A_i}{\partial R} = \frac{\partial D_i}{\partial R} = \frac{-1}{n}$$

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<sup>&</sup>lt;sup>7</sup> Note that we are not allowing a negative weight on relative advantage, despite the fact that, as discussed previously, a few studies of attitudes toward inequality find disaversion to relative advantage. Our assumption is in the tradition of Yitzhaki (1979, 1982) and Fehr and Schmidt (1999) and is consistent with significant recent experimental and survey evidence (Teyssier, 2012; Cojocaru, 2014).

and (6) survives unchanged because in the more general formulation, using (22) we have:

(6') 
$$\frac{\partial G_i^{\lambda}}{\partial R} = \frac{1}{\bar{y}} \left[ \lambda \frac{\partial A_i}{\partial R} + (1 - \lambda) \frac{\partial D_i}{\partial R} \right] = -\frac{1}{n\bar{y}}$$

Proposition 3 described the impact on  $G_i$  of making a small transfer from another person to individual i. Assuming  $y_1 < y_2 < \cdots < y_n$ , the conclusion in the  $\lambda = \frac{1}{2}$  case was that, except for a very small region around the median, a transfer from a higher income person would reduce  $G_i$  if  $y_i$  was below the median, and increase  $G_i$  if  $y_i$  was above the median. Converse results held if the transfer came from a lower income person. The critical role of the median arose because with  $\lambda = \frac{1}{2}$ , advantage,  $A_i$ , and deprivation,  $D_i$ , are equally weighted. In general, the critical percentile is given by 1- $\lambda$ . Thus, for example if one placed half as much weight on  $A_i$  as on  $D_i$ , i.e.  $\lambda = 1/3$ , the critical percentile would be 2/3. That means that a small transfer from someone with higher income would be regarded as equalizing by almost everyone in the bottom two thirds of the population, but as disequalizing by most of those in the top third. This occurs because putting a higher weight on  $D_i$  increases the equalizing impact on  $G_i^{\lambda}$  from the fall in  $D_i$  caused by such a transfer.

#### V. Personal Inequality Assessments at Different Income Levels

In this section we examine how  $G_i^{\lambda}$  varies as  $y_i$  rises from  $y_1$  to  $y_n$ . We provide results for the general case where  $\lambda$  can take on any value in the interval [0,1], but note specific conclusions for the case where  $\lambda = \frac{1}{2}$ .

How does  $G_i^{\lambda}$  change as we move up through the distribution of income? We continue to assume  $y_1 < y_2 < \dots < y_n$ . As we go from individual i to i+1, the absolute income gaps in (3) or implicitly in (22) increase in value by  $y_{i+1} - y_i$  for all j such that  $y_j < y_i$ , and the corresponding gaps for all j > i fall by the same amount. Hence we should expect that  $G_i^{\lambda}$  will initially decline as i rises from 1, since at the start there are more people with j > i than with  $j \le i$ , until some critical point is reached, beyond which  $G_i$  should begin to increase. Formally we have:

**Proposition 4:** If  $y_1 < y_2 < \dots < y_n$ ,

$$G_{i+1}^{\lambda} = G_i^{\lambda}$$
 as  $\frac{i}{n} = 1 - \lambda$ .

**Proof:** See Appendix.

Proposition 4 indicates that  $G_i^{\lambda}$  falls up to the  $(1 - \lambda)100$ th percentile of the distribution and increases above that. As indicated above, this U-shaped pattern is based on the fact that moving from income  $y_i$  to income  $y_{i+1}$  increases the income gaps with lower income people and reduces those with higher income people by the same absolute amount. The relative impact of changes in the upper gaps compared with that of changes in the lower gaps is  $(1-\lambda)/\lambda$ . This means that  $G_i^{\lambda}$  will fall more rapidly starting from i=1 if

 $\lambda < \frac{1}{2}$ , compared with the  $\lambda = \frac{1}{2}$  case, and less rapidly if  $\lambda > \frac{1}{2}$ . Note that if  $\frac{1}{2}$ ,  $G_i^{\lambda} = G_i$  falls up to the 50<sup>th</sup> percentile, that is up to the median, and rises thereafter.

We can also readily identify the value of  $G_i^{\lambda}$  at the bottom and top of the distribution (i = 1 and i = n), as well as the value of  $G_i^{\lambda}$  for the median individual,  $G_{med}^{\lambda}$ , if *n* is odd. We have:

**Proposition 5:** If  $y_1 < y_2 < \cdots < y_n$ ,

(i) 
$$G_1^{\lambda} = (1 - \lambda)(1 - \frac{y_1}{\bar{y}})$$

(ii) if n is odd,  $G_{med}^{\lambda} = \frac{n-1}{2n\overline{\nu}}[(1-\lambda)\overline{y}_{med}^{h} - \lambda \overline{y}_{med}^{l}]$ ; if n is even,  $G_{med}^{\lambda}$  is not defined,

(iii) 
$$G_n^{\lambda} = \lambda(\frac{y_n}{\bar{v}} - 1)$$

**Proof:** See Appendix.

Proposition 5 allows us to put upper bounds on  $G_1^{\lambda}$  and  $G_n^{\lambda}$ . If  $y_1$  is non-negative, the highest possible value of  $G_1^{\lambda}$  is  $1 - \lambda$ , which occurs when  $y_1 = 0$ . When individuals weight deprivation and advantage equally, that is when  $\lambda = \frac{1}{2}$ , the maximum value is  $\frac{1}{2}$ . But the maximum value of  $G_1^{\lambda}$  ranges from 0, when  $\lambda = 1$  and people care only about advantage, to 1 when  $\lambda = 0$  and people only care about deprivation. In view of Proposition 4, these maxima also apply to all  $G_i^{\lambda}$  up to the  $(1-\lambda)100$ th percentile.<sup>8</sup> The upper bound on  $G_n^{\lambda}$  occurs when one individual has all the income and  $y_n = n\bar{y}$ . In that case  $G_n^{\lambda} = \lambda(n-1)$ , which is also an upper bound for all  $G_i^{\lambda}$ 's above the  $(1 - \lambda)100$ th percentile.

Part (ii) of the proposition is also interesting, in throwing light on the value of the personal inequality index for the "average person", that is on the value of  $G_{med}^{\lambda}$ . The latter is based on a weighted average of  $\bar{y}_{med}^h$  and  $\bar{y}_{med}^l$ , with the weight on  $\bar{y}_{med}^h$  falling with  $\lambda$ . In the focal case with = 1/2, we have:

$$G_{med} = \frac{(n-1)}{4n\overline{\nu}} (\overline{y}_{med}^h - \overline{y}_{med}^l)$$

Since in any real-world example  $(n-1)/n \approx 1$ , this says:

$$G_{med} pprox rac{ar{y}_{med}^h - ar{y}_{med}^l}{4ar{y}}$$

In the U.S. today, for household income before tax,  $\bar{y}_{med}^h \approx \frac{8}{5}\bar{y}$  and  $\bar{y}_{med}^l \approx \frac{2}{5}$ , which yields  $G_{med} \approx 0.3$ , less than the value of the Gini coefficient, which was 0.476 in 2013. We may also note values of  $G_{med}$ under some familiar continuous distributions.  $G_{med}$  would equal  $\frac{1}{4}$  for a uniform distribution, and if  $y_i \sim$  $N(\mu, \sigma)$ , it would equal  $\frac{2\sigma}{5\mu}$ , that is two-fifths of the coefficient of variation.

<sup>&</sup>lt;sup>8</sup> Note that with  $\lambda = 1$ , the  $(1 - \lambda)100$ th percentile = 0, so that  $G_i^{\lambda}$  has no falling range.

<sup>9</sup> With the help of quintile share and other data from U.S. Census Bureau (2015) it can be estimated that  $\bar{y}_{med}^h = 0$  $1.64\overline{y}$  and  $\overline{y}_{med}^l = 0.36\overline{y}$ .

We can see that  $G_i^{\lambda}$  will generally not be symmetric around the median. Looking at the  $\lambda=1/2$  case again, for example,  $G_i$  will never be greater than 1/2 at the lowest income level, but can be very high at the top end.  $G_i$  is not bounded on the upper end by 1, unlike the Gini coefficient.  $G_n=1$  is reached when  $\frac{y_n}{\bar{y}}=3$ . That ratio is exceeded in almost all real-world cases. This implies that, in a mathematical sense, the rich perceive that there is more inequality than do the poor when  $\lambda=1/2$ , which is not unintuitive. If you are rich there are relatively few people whose incomes are close to yours, meaning there is a large gulf between your income and most others'.

#### VI. Personal Inequality During Secular Change in Income Distribution

This section asks how  $G_i^{\lambda}$  can be predicted to behave at different income levels during periods of secular change in income distribution. We focus initially in each case on the  $\lambda = 1/2$  case, in which individuals weight deprivation and advantage equally, referring to  $G_i^{1/2}$  simply as  $G_i$ , as above. We start with the Kuznets transformation and go on to the polarization and rising inequality that we have seen in the U.S. and many other high income countries in the last few decades. The principles at work are explored with the help of examples, which are intended merely to be illustrative.

# Kuznets Transformation

Kuznets (1955) studied what happens to income distribution and inequality in a growing economy where the composition of output is shifting from an initially large traditional agricultural sector to a modern sector. The modern sector eventually comprises most if not all of the economy. The consequences for inequality can be illustrated by considering a stylized model in which individual incomes are uniform within each of the sectors, higher in the modern sector, and unchanging during the growth process. <sup>10</sup> In this case the Gini coefficient, G, rises until the fraction of the population in the modern sector, p, hits a critical value, after which it declines. This critical value of p is less than one half. That is because, while the mean difference has a maximum at p = 1/2, the mean, which appears in the denominator of the expression for G, is rising throughout, so G has already started to decline at p = 1/2.

The behavior of the GAPIIs, that is the  $G_i$ s, and G during the Kuznets transformation will be illustrated here using an example whose implications are shown in Figure 1. It is assumed that income of each person in the traditional sector is 11.7% of per capita income in the modern sector. This gap is sufficient for the peak value of G to be 0.49, the value observed in China in 2008 (Li and Sicular, 2014). China is the most prominent recent example of a society going through the kind of transformation that Kuznets described. In the early 1980s its Gini coefficient for family income fluctuated around 0.30 (Sicular,

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<sup>&</sup>lt;sup>10</sup> Kuznets considered a richer range of possibilities. He allowed unequal income distribution within both sectors and believed the leading case was one in which there was greater inequality in the modern sector than in the traditional, or agricultural, sector. He also considered the impacts of changes in the relative income, and of income inequality, in the modern vs. the agricultural sector over time. In most cases he found that as the relative population of the agricultural sector declined over time there was an initial increase in inequality followed by a decline.

2013). This was followed by a rapid rise with later deceleration to the 2008 peak, after which G began to fall slowly. <sup>11</sup> China may now be past a Kuznets curve peak.. <sup>12</sup>

We will refer to the individual inequality measures of people in the low and high income groups as  $G_L$  and  $G_H$  respectively. Since no one is worse off than those in the low income group,  $G_L = \frac{D_L}{\bar{y}}$ , that is it is based entirely on deprivation, while  $G_H = \frac{A_L}{\bar{y}}$  and is based wholly on advantage. As shown in Figure 1, when the modern sector is tiny,  $G_L$  is not far above zero. Almost everyone in the society has the same low income, so that  $n_L^h/n$  and therefore  $D_L$  are very low. The situation in the modern sector is the opposite. Since almost everyone has much lower income than those in the modern sector, the individual inequality measure there,  $G_H$  is very high. Now, as development proceeds,  $G_L$  rises monotonically and  $G_H$  falls monotonically (and dramatically, in the example reflected in Figure 1). It is as if people in the traditional sector become steadily more aware of the inequality between themselves and people in the modern sector as the modern sector grows. On the other hand, from the viewpoint of individuals in the modern sector, inequality is falling because more and more of their fellow citizens are as well off as they are.

How does one resolve the conflict when two population groups have such radically opposed views about the trend in inequality? The Gini coefficient proposes a solution - - take an average of the individual assessments. Thus in the Kuznets curve example, G is a population weighted average of the values of  $G_L$  and  $G_H$ . An alternative would be to take a vote on the question of whether inequality was rising or falling - - a "democratic" approach. Here the democratic approach would say that inequality rises until  $p = \frac{1}{2}$  and falls thereafter. In the example, G says that inequality rises until  $P = \frac{1}{4}$  and falls thereafter. That is because  $G_H$  falls faster than  $G_L$  rises, so that averaging  $G_H$  and  $G_L$ , even using population weights, places greater relative importance on the decline in  $G_H$  than on the rise in  $G_L$ . Thus the Gini procedure of averaging individual inequality assessments does not correspond to the democratic approach in this situation, and places more importance on the views of the wealthy.

The difference in the views of traditional vs. modern sector people about the trend of inequality during development clearly has the potential to create resentment and misunderstanding. Observers sometimes wonder why high income people seem to be unconcerned about what they view as rising inequality in the initial stages of development. The suspicion is perhaps that these people turn a blind eye because they benefit from the process. What we see here is that, from their viewpoint, inequality is actually falling. This is their honest assessment. Hence we have a "perfect storm" - - numerous poor people who think inequality is rising and a growing number of rich people who think the opposite. In the real world such a situation could clearly cause tension.

Our analysis shows that, unfortunately, use of the Gini coefficient could cause confusion about what is happening to inequality during the Kuznets transformation due to its greater sensitivity to the views of the high income group. The Gini begins to fall "too soon". If the behavior of *G* were used as an input into

<sup>11</sup> The National Bureau of Statistics estimates of the national Gini coefficient for family income were 0.491 in 2008 (Li and Sicular, 2014, Appendix A) and 0.469 in 2014 (Qi, 2015).

<sup>&</sup>lt;sup>12</sup>Knight (2014) discusses whether China may be beyond the peak of the Kuznets curve. His conclusion is that this depends in part on public policy but that there are now strong underlying forces pushing in the direction of reducing inequality in China.

policy decisions, this could potentially lead to a relaxation of inequality-reducing measures in a country where the majority of the population had yet to join the modern sector and still felt that inequality was rising.

The above analysis would not be affected significantly by moving from the  $\lambda = \frac{1}{2}$  case to  $\lambda \neq \frac{1}{2}$ . There would of course be no impact on the time path of G. Since those in each sector are only concerned either about deprivation (in the traditional sector) or advantage (in the modern sector) what occurs at the individual level is simply a rescaling of  $G_L$  and  $G_H$  at each point in the Kuznets process. A majority of people still believe inequality is rising until  $p = \frac{1}{2}$  is reached, and above this point the majority think inequality is falling. G still has its peak at the same point as with  $\lambda = \frac{1}{2}$ . In terms of Figure 1, there will be a proportionate shift of the  $G_L$  curve by the factor  $2(1 - \lambda)$  and a shift of the  $G_H$  curve in the opposite direction by the factor  $2\lambda$ . In the case where  $\lambda < \frac{1}{2}$  the  $G_L$  and  $G_H$  curves will move towards each other, while if  $\lambda > \frac{1}{2}$  the result will be the opposite.

#### Polarization

There is much theoretical and empirical literature on polarization (including Foster and Wolfson, 1992; Esteban and Ray, 1994; Acemoglu and Autor, 2011; Autor and Dorn, 2013; Green and Sand, 2015). Polarization can take different forms. Without saying so, we have already been discussing one form in the context of the Kuznets transformation, which has two poles: the traditional society and the modern sector. At the starting point, with everyone in the traditional sector, there is extreme polarization. As population shifts to the modern sector that polarization initially declines, but aggregate inequality rises according to the Gini coefficient, which people in the traditional sector agree with but people in the modern sector do not. Then there is a phase where polarization continues to decline but changes in the Gini coefficient turn from positive to negative. Finally, when the modern sector population becomes a majority, polarization begins to decrease, as does the Gini coefficient, but inequality continues to rise in the view of those in the traditional sector.

The behavior of polarization, the aggregate Gini coefficient, and personal inequality assessments over the course of the Kuznets transformation illustrate two important points about polarization and inequality:

- i) Polarization and individual inequality assessments may move in opposite directions,
- ii) Polarization and aggregate inequality measures may move in opposite directions.

It is clear from the Kuznets case alone that the relationship between polarization and inequality is complex. The relationship is even more complex in the case of the polarization in labor markets that has received attention in the US and other high income countries in recent years. In this case the relative demand for labor shifts away from mid-level occupations to both low-skilled and (especially) high skilled occupations Other things constant this should result in a shift in labor force composition away from the middle earning levels toward both high and low labor incomes. Such a shift has indeed occurred over significant timespans in the U.S., Canada, the UK, Germany and some other European countries (Acemoglu and Autor, 2011; Green and Sand, 2015). In most cases the relative wages of highly skilled workers have increased. In the US it has also been found that the relative wages of certain low skilled occupations have risen (Autor and Dorn, 2013).

We will analyze the kind of polarization seen over the last few decades in labor markets by first considering the effects of *population shift*, that is a rise in the number of individuals at low and high incomes combined with a reduction in the number at middle income. Subsequently we will look at the effect of changes in relative incomes as well. As in the Kuznets analysis it helps to consider a stylized situation. Assume that there are just three income levels in a society and that they display  $y_L < y_M < y_H$ . Numbers of individuals in the three groups are  $n_L, n_M$ , and  $n_H$ . As in the Kuznets case the GAPIIs of people in the bottom group and top groups are given by  $G_L^{\lambda} = \frac{\lambda D_L}{\bar{\nu}}$  and  $G_H^{\lambda} = \frac{(1-\lambda)A_H}{\bar{\nu}}$ .

Also as in the Kuznets analysis the increase in  $n_H$  will tend to make  $A_H$  and  $G_H^{\lambda}$  increase since (from 5i):

(23) 
$$A_{H} = \frac{(n_{L} + n_{M})}{n} (y_{H} - \bar{y}_{H}^{l})$$

However, there is now an offsetting effect because  $\bar{y}_H^l$  falls due to the population shift from the middle to lower groups, and therefore  $(y_H - \bar{y}_H^l)$  increases. It can readily be shown that:

(24) 
$$\Delta A_H, \Delta G_H = 0 \text{ as } \frac{\Delta n_L}{-\Delta n_M} = \frac{y_H - y_M}{y_H - y_L}$$

Now  $\frac{y_H - y_M}{y_H - y_L} < 1$  and  $\frac{\Delta n_L}{-\Delta N_M} < 1$  as well, so it is not immediately clear which way the inequality will go. However, with a positively skewed distribution of income we would have  $\frac{y_H - y_M}{y_H - y_L} > \frac{1}{2}$ , so that if half or fewer of those leaving the middle income group go to the lower group (which is in line with the experience in the US at least),  $A_H$  and  $G_H$  will decline, as in the Kuznets case.

Turning to the bottom group, from (5ii) we have:

(25) 
$$D_{L} = \frac{(n_{M} + n_{H})}{n} (\bar{y}_{L}^{h} - y_{L})$$

And it can be shown that:

(26) 
$$\Delta A_L, \Delta G_L = 0 \text{ as } \frac{-\Delta n_M}{\Delta n_H} = \frac{y_H - y_L}{y_M - y_L}$$

Now,  $\frac{y_H - y_L}{y_M - y_L} > 1$  and  $\frac{-\Delta n_M}{\Delta n_H} > 1$  as well, so again there is ambiguity. Once more appealing to positive skewness,  $\frac{y_H - y_L}{y_M - y_L} > 2$  is likely. So if half or more of those leaving the middle group go to the top group (which is of course the same as saying that half or fewer go to the bottom group, as above),  $A_L$  and  $G_L$  will fall, which is the opposite of what we found in the Kuznets analysis. This would be the result of the increase in  $\bar{y}_L^h$  having a larger effect on  $A_H$  and  $G_H$  than the decline in  $n_L^h = (n_M + n_H)$ .

The analysis of  $A_L$  and  $A_H$  is sufficiently complex that one may (correctly) anticipate that the analysis of  $A_M$  and  $G_M$  would be tedious. This is not only because  $G_M$  depends on both  $A_M$  and  $D_M$ , immediately doubling the algebra, but also because for a general analysis allowing  $\lambda \neq \frac{1}{2}$  one would need to think about how different weightings of  $A_M$  and  $D_M$  would affect the results. Clearcut results are hard to get.

Suffice it to say that during polarization, population shift alone may increase, decrease, or leave unchanged personal inequality as viewed by the middle group.

What about changes in relative incomes associated with polarization? A robust finding across countries is that the relative income of the highly skilled has risen during observed labor market polarization. With the incomes of the two lower groups assumed unchanged, from (23) we see that this makes it more likely that  $A_H$  and  $G_H$  would rise with polarization, rather than declining as in the Kuznets analysis. This effect would be strengthend if  $\bar{y}_H^l$  declined, which could occur as a result of  $y_M$  falling, which is also consistent with what is generally observed. From (25) we can see that a rise in  $y_H$  could also give  $A_L$  and  $G_L$  more of a tendency to increase, via its effect on  $\bar{y}_L^h$ , although that could be offset by a fall in  $y_M$  which would act to reduce  $\bar{y}_L^h$ .

Given the theoretical ambiguity of the behavior of  $G_L$ ,  $G_M$ , and  $G_H$  it is helpful to consider an example based on real-world observations. Autor and Dorn (2013) set out the changes in employment shares and wage rates for six broad occupational groups in the U.S. from 1980 to 2005. As shown in Table 1 here, the top group, consisting of managers, professionals, technicians, finance and public safety occupations experienced a 29% increase in employment share and a 36% rise in wage rates over those years. The middle group shown in Table 1, which aggregates the middle four occupational groups in Autor and Dorn (2013), had a 22% drop in employment share and only a 9% increase in wages. Finally, the bottom group, consisting of service occupations, had a 30% rise in employment share and a 17% increase in wages. These changes provide a dramatic example of labor market polarization.

Table 1 shows  $G_i$  rising for all three groups, as do  $A_i$  and  $D_i$  where applicable. The wage gap between the top group and the rest of the labor force expands considerably, leading to  $A_H$  more than doubling from 1980 to 2005. The middle group experiences a large increase in deprivation, which is not surprising in view of its poor wage performance and the large employment and wage increases for the top group. But the middle group also sees a rise in its advantage over the bottom group, which is due to the increase in the relative size of the latter group. The 17% wage rise of the bottom group is not large enough to overcome the deprivation-increasing effect for it of the expansion and large wage increase of the top group, so its deprivation increases quite a bit.

The above results are obtained with  $\lambda=1/2$ , of course. But changing  $\lambda$  will not produce a direction of change in  $G_i^\lambda$  different from that in  $G_i$ , since we do not have a case where either advantage or deprivation are falling. Reweighting  $A_i$  and  $D_i$  cannot produce a sum that decreases. This result does, however, depend on how Autor and Dorn's original six broad occupational groups are aggregated into three groups. Autor and Dorn (2013) stress that the only low wage group that sees a rise in employment share is their service occupations group, and that original group has been treated here as the bottom of our three more aggregated groups. However, although the original 1980 group with the second-lowest wage, those in clerical and retail sales occupations, has a small drop in employment share, it, like the service occupations, shows a relative wage *increase*. Thus the clerical and retail sales occupations benefit from wage polarization if not from employment polarization. Again aggregating to three groups, but putting clerical and retail sales occupations in the bottom category along with the service occupations, changes results a little.  $G_L$ ,  $G_M$ , and  $G_H$  all increase, but  $A_M$  falls. Hence, if  $\lambda$  is sufficiently high, more precisely 0.72 or more,  $G_M$  declines between 1980 and 2005. That is, if the middle group is sufficiently concerned

about advantage it will regard polarization as having reduced inequality in this case. While worth noting, this result may not affect one's conclusions much in view of the broad consensus in the literature that it is likely that  $\lambda \le \frac{1}{2}$ .

# Rising Overall Inequality

In the last four decades substantial periods of rising overall income inequality, as measured by the Gini coefficient and other conventional indexes, have been observed in a wide range of high income countries (Roine and Waldenström, 2015). In some cases this reflects polarization, but labelling all increases in inequality as polarization would abuse the latter term. It is probably best to refer to a broadly-based downward movement in the Lorenz curve simply as an increase in inequality.

It is interesting to ask what is likely to happen to personal inequality assessments during a period of rising inequality. Table 2 provides some insight on this question. Using  $\lambda = 1/2$ , it shows  $G_i$  at selected percentiles of lognormal income distributions that have overall G = 0.2, 0.3, 0.4, 0.5 and 0.6. These Gini values span most of the range observed across countries. For reference, the Gini coefficient for household income in the U.S. was 0.397 in 1975 and rose with little interruption to 0.476 in 2013 (U.S. Census Bureau, 2015). In the UK the Gini coefficient for equivalized household income was 0.272 in 1977 and rose to 0.324 in 2013/14 (Office of National Statistics, 2015, Figure 5).

Table 2 shows, first, that  $G_i$  falls with income up to the median and then rises, as predicted by Proposition 4. The latter increase, from percentile to percentile, rises with income, particularly at the highest levels. We see, for example, with an overall Gini of 0.4 that while  $G_i$  approximately doubles, from 0.272 to 0.586, in going from the median to the 90<sup>th</sup> percentile, it then roughly triples to arrive at 1.670 for the 99<sup>th</sup> percentile. Second, the table shows that sensitivity to rising inequality is greatest at top income levels. This is more clearly illustrated in Table 3 which shows high income-low income  $G_i$  ratios by percentile, given different values of G. The P90:P10  $G_i$  ratio rises from 1.62 when G = 0.4 to 1.80 when G = 0.5, and the P99:P1 ratio rises from 3.85 when G = 0.4 to 5.36 when G = 0.5. The increase of G by 0.1, from 0.4 to 0.5 is roughly similar to the rise seen in the U.S. since 1975, so this is suggestive with respect to real-world changes in inequality assessments by people at the top of the income distribution. Thus these results raise the interesting possibility that high income people could have experienced the largest perceived increases in inequality in recent decades.

An idea of the quantitative impact of allowing  $\lambda \neq 1/2$  is provided in Tables 4 and 5, which repeat the exercises of Table 2 and 3, but with the range of G confined to [0.3,0.5] and alternate values of  $\lambda = 0.25$  and 0.75 considered. Note first that  $G_i^{\lambda}$  initially declines as income rises but hits a minimum at the  $(1-\lambda)100$ th percentile, as predicted by Proposition 4. Next, we can see that raising  $\lambda$  twists the  $G_i^{\lambda}$  profile. For lower incomes,  $G_i^{\lambda}$  falls but for higher incomes  $G_i^{\lambda}$  rises. This means that there is an increase with  $\lambda$  in the acceleration of  $G_i^{\lambda}$  as one goes up the income scale, and a rise in  $G_i^{\lambda}$  ratios for such income percentiles as P90/P10 and P99/P1 (Table 5). The switch from a negative to positive impact of  $\lambda$  on  $G_i^{\lambda}$  occurs at P61 for G = 0.3, at P65 for G = 0.4, and at P69 for G = 0.5.

#### VII. Discussion and Conclusion

We have seen that recognizing the Gini coefficient as the average of personal inequality indexes generates rich results. This is partly because a wide range of personal viewpoints about inequality measurement are "Gini admissible". The Gini admissible personal indexes, or GAPIIs, are a weighted average of an individual's deprivation and advantage. Deprivation is based on the sum of differences between the individual's own income and that of people with higher income, while advantage is based on the sum of differences vis-à-vis people with lower income. Deprivation and advantage are standard concepts in the literature on individual attitudes toward inequality. However, what happens when they are combined to form GAPIIs has not been previously investigated.

One remarkable feature of GAPIIs is that they are completely insensitive to transfers of income that occur only among people who have incomes strictly above those of the reference individual, or among those who have incomes strictly below that individual. This means that GAPIIs do not obey the Pigou-Dalton principle of transfers. But the individual does regard transfers from those in the "above group" to those in the "below group" as equalizing while transfers in the other direction are considered disequalizing, as one would normally expect. These features help to explain the fact that the sensitivity to transfers of the Gini coefficient itself depends critically on the number of people with incomes between those of the donor and recipient. That property can now be seen to result mostly from the fact that the only transfers individuals with a GAPII regard as affecting inequality at all, aside from those that alter their own incomes, are transfers that "pass over" them.

Another important aspect is that the relative weights placed on deprivation and advantage can vary across societies. Thus, in one society, people might care only about deprivation - - they may be said to "resent" the fact that some others have higher incomes. In another society, they might only care about advantage - - either exulting in being better off than some others or showing concern for "those less fortunate than themselves". And, of course, any weighting between these extremes may be allowed. It is tempting to imagine that this feature might have something to do with the apparently universal appeal of the Gini coefficient. If everyone agrees with assessing overall inequality by averaging individuals' inequality assessments and that the latter should be based on sums or averages of absolute income differences, people in all societies can agree that the Gini coefficient is the appropriate measure of overall inequality, even if the weights placed on deprivation and advantage differ across societies.

The pattern of GPAII values as we go up the income scale is also of interest. As we have seen, starting from the lowest income, the personal index values fall up to a point - the median when deprivation and advantage are weighted equally - and then rise. With the positively skewed income distributions seen in the real world, if deprivation is not weighted sufficiently less than advantage, the value of the index will rise to a higher level at top incomes than it displays at low income levels. The paper ended with a discussion of how personal inequality assessments may behave during secular changes in income distribution. We have seen, for example, that in the simplest model people in the traditional sector will regard inequality as rising *throughout* the Kuznets transformation, while those in the modern sector think precisely the opposite. The resulting scope for misunderstanding and conflict seems large. This may throw some light on the tensions that are observed during periods of rapid modernization and rural-urban migration in developing countries. A further insight comes from the fact that the Gini coefficient says that the Kuznets transformation stops being disequalizing well before half the population is in the modern

sector. This signals that the fact that the Gini is the average of the personal inequality values does not imply that it is democratic in its judgements. The reason it is not democratic is that higher income people tend to have numerically larger personal inequality assessments, so that changes in their assessments have more influence on the average than do changes for lower income people.

Less clearcut results than those found for the Kuznets transformation are obtained for polarization. Under polarization, population shifts not only to the top but also to the bottom, with a shrinking middle group. Complex changes in relative incomes can also occur. The result is that there are circumstances under which people in each of the top, middle and bottom income groups may regard inequality as rising, and others in which they may all think it is falling, or may have mixed assessments. Given this ambiguity we turned to the real world for some guidance. In a three-group example set up to parallel the actual polarization seen in the U.S. over the period 1980 – 2005, we saw that personal inequality rose from the viewpoint of all three groups in a base case. However, broadening the bottom group led to the result that the middle group could have regarded inequality as falling if it placed a sufficiently high weight on advantage compared with deprivation.

Finally, we examined the impact of a general spreading of the income distribution by seeing how rising dispersion of a lognormal distribution would affect personal inequality assessments. Such a trend raises personal inequality values at all levels of income irrespective of the relative weights placed on deprivation and advantage. However it does not do so equally. Unless sufficiently more weight is placed on deprivation, the increase in inequality is greatest from the viewpoint of those with the highest incomes. Could this help to explain why rising inequality has begun to get so much attention recently in the global financial media and such quarters as the annual meetings of the World Economic Forum in Davos? It is an intriguing question.

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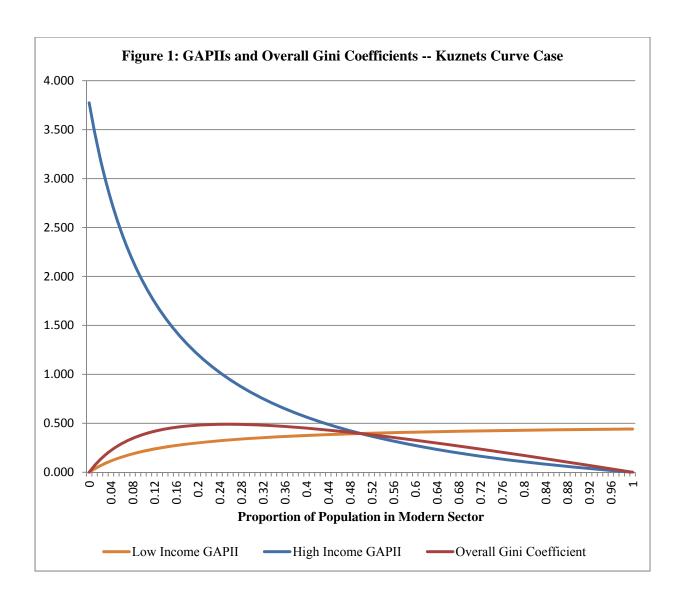
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 $Table\ 1$  Advantage, Deprivation and Personal Inequality Indexes with  $\lambda=1/2,$  by Occupation Group - - Polarization Example Based on U.S. Data, 1980 and 2005

# I. 1980

Occupation Group	Employment Share	Mean Wage (2004 \$s)	$A_i$	$D_i$	$G_i$
1. Top	0.316	17.0	3.42	0	0.126
2. Middle	0.585	12.6	0.44	1.38	0.067
3. Bottom	0.099	8.2	0	5.36	0.198

#### II. 2005

Occupation Group	Employment Share	Mean Wage (2004 \$s)	$A_i$	$D_i$	$G_i$
<b>1. Top</b>	0.409	23.1	6.11	0	0.226
2. Middle	0.462	13.7	0.53	3.86	0.162
3. Bottom	0.129	9.6	0	7.41	0.274

**Notes:** (i) The mean wage is the geometric mean hourly wage, derived from the mean log hourly wage reported by Autor and Dorn (2013).

(ii) The Top occupational group is the first category in Autor and Dorn (2013). It includes managers, professionals, technicians, finance and public safety occupations.

The Middle occupational group consists of the four middle groups in Autor and Dorn (2013): production and craft occupations; transportation, construction, mechanics, mining and farm occupations; machine operators and assemblers; and clerical and retail sales occupations. The Bottom occupational group consists of service occupations.

(iii)  $A_i$ ,  $D_i$ , and  $G_i$  are the personal advantage, deprivation, and inequality indexes.

**Source:** Employment share and mean wage are from Autor and Dorn (2013, Table 1) - - see Note (i). The other columns are from calculations by the author.

Table 2  $Personal\ Inequality\ Indexes\ with\ \lambda=\frac{1}{2'},\ by\ Overall\ Gini\ Coefficient\ and\ Percentile,$   $Lognormal\ Example$ 

Overall Gini:	0.2	0.3	0.4	0.5	0.6
Percentile:					
1	0.297	0.380	0.434	0.467	0.486
5	0.244	0.328	0.392	0.437	0.468
10	0.213	0.295	0.362	0.414	0.452
25	0.165	0.240	0.307	0.366	0.415
50	0.140	0.208	0.272	0.331	0.385
75	0.175	0.261	0.342	0.415	0.476
90	0.273	0.427	0.586	0.743	0.886
95	0.362	0.591	0.850	1.135	1.431
99	0.585	1.048	1.670	2.502	3.608

Table 3  ${\it Ratios of Personal Inequality Indexes for Selected Percentiles with } \lambda = 1/2, Lognormal Example$ 

Overall Gini:	0.2	0.3	0.4	0.5	0.6
P60/P40	1.004	1.006	1.007	1.008	1.009
P70/P30	1.030	1.044	1.056	1.065	1.070
P75/P25	1.059	1.087	1.112	1.132	1.146
P80/P20	1.103	1.155	1.204	1.247	1.277
P90/P10	1.281	1.444	1.618	1.795	1.960
P95/P5	1.483	1.799	2.171	2.596	3.060
P99/P1	1.968	2.757	3.850	5.361	7.430

Table 4  $\label{eq:personal} \mbox{Personal Inequality Indexes by Overall Gini Coefficient and Percentile with Alternative Values of $\lambda$, $$ Lognormal Example$ 

	λ = 0.25			λ = 0.75		
Overall Gini:	0.3	0.4	0.5	0.3	0.4	0.5
Percentile:						
1	0.570	0.650	0.700	0.190	0.217	0.233
5	0.491	0.586	0.655	0.166	0.197	0.219
10	0.439	0.539	0.618	0.152	0.185	0.210
25	0.341	0.443	0.533	0.139	0.172	0.199
50	0.242	0.332	0.423	0.173	0.211	0.240
75	0.199	0.278	0.361	0.322	0.406	0.468
90	0.242	0.342	0.449	0.611	0.830	1.037
95	0.311	0.452	0.613	0.870	1.248	1.656
99	0.528	0.843	1.266	1.567	2.496	3.739

Note:  $\lambda$  is the weight placed on relative advantage,  $A_i$ , in the calculation of the personal inequality index,  $G_i$ . See text.

Table 5  $\label{eq:ratios} \mbox{Ratios of Personal Inequality Indexes for Selected Percentiles with Alternative Values of $\lambda$, } \\ \mbox{Lognormal Example}$ 

	λ = 0.25			λ = 0.75		
Overall Gini:	0.3	0.4	0.5	0.3	0.4	0.5
P60/P40	0.787	0.811	0.837	1.406	1.401	1.382
P70/P30	0.636	0.674	0.717	1.965	1.974	1.949
P75/P25	0.584	0.627	0.677	2.320	2.358	2.353
P80/P20	0.549	0.599	0.656	2.747	2.847	2.897
P90/P10	0.552	0.634	0.727	4.016	4.487	4.941
P95/P5	0.633	0.772	0.936	5.252	6.329	7.548
P99/P1	0.927	1.296	1.808	8.239	11.506	16.015

Note:  $\lambda$  is the weight placed on relative advantage,  $A_i$ , in the calculation of the personal inequality index,  $G_i$ . See text.

### **Appendix**

This appendix provides proofs of propositions 4 and 5.

**Proposition 4:** If  $y_1 < y_2 < \dots < y_n$ ,

$$G_{i+1}^{\lambda} = G_i^{\lambda}$$
 as  $\frac{i}{n} = 1 - \lambda$ .

**Proof:** From (4'), (5) and (22), and using the assumption that  $y_1 < y_2 < \dots < y_n$ ,  $G_{i+1}^{\lambda} = G_i^{\lambda}$  as:

(A1) 
$$\lambda \sum_{j=1}^{i} (y_{i+1} - y_j) + (1 - \lambda) \sum_{j=i+2}^{n} (y_j - y_{i+1}) = \lambda \sum_{j=1}^{i-1} (y_i - y_j) + (1 - \lambda) \sum_{j=i+1}^{n} (y_j - y_i)$$

Now, the left-hand side of this expression can be written:

(A2) 
$$\lambda \left[ \sum_{j=1}^{i-1} (y_i - y_j) + i(y_{i+1} - y_i) \right] + (1 - \lambda) \left[ \sum_{j=i+1}^{n} (y_j - y_i) + (n-i)(y_i - y_{i+1}) \right]$$

Hence, (A1) simplifies to:

$$\lambda i(y_{i+1} - y_i) + (1 - \lambda)(n - i)(y_i - y_{i+1}) = 0$$

which is equivalent to:

$$\lambda i - (1 - \lambda)(n - i) = 0$$

which becomes:

$$\lambda n + i - n = 0$$

from which one readily derives the result that

$$G_{i+1}^{\lambda} = G_i^{\lambda} \quad \text{as} \quad \frac{i}{n} = 1 - \lambda$$

**Proposition 5:** If  $y_1 < y_2 < \cdots < y_n$ ,

(i) 
$$G_1^{\lambda} = (1 - \lambda)(1 - \frac{y_1}{\bar{v}})$$

(ii) if n is odd,  $G_{med}^{\lambda} = \frac{n-1}{2n\bar{y}}[(1-\lambda)\bar{y}_{med}^{h} - \lambda\bar{y}_{med}^{l}]$ ; if n is even,  $G_{med}^{\lambda}$  is not defined,

(iii) 
$$G_n^{\lambda} = \lambda(\frac{y_n}{\bar{y}} - 1)$$

Proof: (i) From (5) and (22), given that  $n_1^l = 0$ ,

$$G_1^{\lambda} = \frac{(1-\lambda)}{n\bar{y}} [n_1^h (\bar{y}_1^h - y_1)]$$

$$= \frac{(1-\lambda)}{n\bar{y}} [\sum_{j=2}^n y_j - (n-1)y_1]$$

$$= \frac{(1-\lambda)}{n\bar{y}} (\sum_{j=1}^n y_j - ny_1)$$

$$= \frac{(1-\lambda)}{n\bar{y}} (n\bar{y} - ny_1)$$

$$= (1-\lambda)(1 - \frac{y_1}{\bar{y}})$$

(ii) From (5) and (22), if n is odd we have:

$$G_{med}^{\lambda} = \frac{\lambda}{n \bar{\nu}} n_{med}^l \left( y_{med} - \bar{y}_{med}^l \right) + \frac{(1-\lambda)}{n \bar{\nu}} n_{med}^h \left( \bar{y}_{med}^h - y_{med} \right)$$

Noting that  $n_{med}^l = n_{med}^h = \frac{n-1}{2}$ ,

$$G_{med}^{\lambda} = \frac{n-1}{2n\bar{y}} [(1-\lambda)\bar{y}_{med}^{h} - \lambda \bar{y}_{med}^{l}]$$

If *n* is even there is no individual with median income since  $y_1 < y_2 < \dots < y_n$  .

(iii) From (5) and (22), given that  $n_n^h = 0$ ,

$$G_n = \frac{\lambda}{n\bar{y}} [n_n^l (y_n - \bar{y}_n^l)]$$

$$= \frac{1}{n\bar{y}}[(n-1)y_n - \sum_{j=1}^{n-1} y_j]$$

$$= \frac{\lambda}{n\bar{y}} [ny_n - \sum_{j=1}^n y_j]$$

$$= \frac{\lambda}{n\bar{y}} [ny_n - n\bar{y}]$$

$$= \lambda [\frac{y_n}{\bar{y}} - 1]$$