Equality of opportunity:
Definitions and testable conditions, with an application to
income in France

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July 17, 2008

Abstract
We offer a model of equality of opportunity that encompasses different conceptions expressed in the public and philosophical debates. In addition to circumstances whose effect on outcome should be compensated and effort which represents a legitimate source of inequality, we introduce a third factor, luck, that captures the non-responsibility factors whose impact on outcome should be even-handed for equality of opportunity to be satisfied. Then, we analyse how the various definitions of equality of opportunity can be empirically identified, given data limitations and provide testable conditions. Definitions and conditions resort to standard stochastic dominance tools. Lastly, we develop an empirical analysis of equality of opportunity for income acquisition in France over the period 1979-2000 which reveals that the degree of inequality of opportunity tends to decrease and that the degree of risk of income distributions, conditional on social origin, appears very similar across all groups of social origins.

JEL Codes: D63, J62, C14
Keywords: Equality of opportunity, Income distribution, Luck, Stochastic dominance.

*This paper is part of a research program supported by the French Commissariat au Plan. We are grateful to Gérard Forgeot from the French national statistical Agency (INSEE) for the access to the data BdF 2000. For useful comments, we wish to thank François Bourguignon, Christine Chwaszcza, Russell Davidson, Jean-Yves Duclos, Marc Fleurbaey, Dirk Van de Gaer, Nicolas Gravel, John Roemer and participants in seminars at the French ministry of the economy (Fourgeaud seminar), the University of Essex (ISER), the University of Oxford, the University of Ghent (Public economics seminar), the University of Verona (Canazei Winter School) and the European University Institute (RSCAS).
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1 Introduction

Most economic analysis of inequality, theoretical and empirical, rely on the assumption that equality of individual outcomes (e.g. welfare, income, health) is *per se* a desirable social objective. This is sometimes criticized for standing at odd with both public perceptions of inequalities and some developments in modern theories of justice. According to this criticism, a distinction must be drawn between morally or socially justified and unjustified inequalities. This has led egalitarian philosophers such as Rawls (1971), Dworkin (1981a, 1981b), Sen (1985), Cohen (1989) or Arneson (1989, 1990) to claim that distributive justice does not entail the equality of individual outcomes but only requires that individuals face equal opportunities for outcome. Despite the growing political audience of this view, few economic analysis have tried to assess the extent to which equality of opportunity is empirically satisfied. Two major issues are likely to account for this state of affairs. First, how should equality of opportunity be characterized? In fact, no consensus has been reached, neither in the philosophical nor in the public debates, regarding how opportunities should be defined and in what sense they should be considered equal.

In this paper we offer a model of equality of opportunity that encompasses several conceptions expressed in these debates. Second, how can equality of opportunity be empirically assessed? This requires that the determinants of individual outcomes be taken into account. However, these determinants are never fully observable. Hence, we analyze how the various conceptions can be empirically identified, given data limitations, and provide testable conditions for equality of opportunity. Lastly, we develop an empirical implementation of these conditions and examine the extent to which equality of opportunity is achieved in the distribution of income in France.

One important implication of the equal-opportunity view is that judgements about equality must take into account the determinants of individual outcomes. At least two sets of factors must be distinguished: on the one hand, factors that reflect individual responsible choices and are considered a legitimate source of inequality; on the other hand, factors beyond the realm of individual responsibility and that do not appear as socially or morally acceptable.

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1Roemer et al. (2003), O’Neill et al. (2000), Checchi et al. (1999), Benabou & Ok (2001), Bourguignon et al. (2007), Goux and Maurin (2003), Alesina and La Ferrara (2005) and Checchi & Peragine (2005) who analyze equality of opportunity for income and Schuetz et al. (2005) who examine educational opportunities are some of the exceptions.
sources of inequality. Following the terminology introduced in Roemer (1998), we refer to the former determinants as *effort* and to the latter as *circumstances*. As most authors would agree, the principle of equality of opportunity essentially requires, that, given individual effort, circumstances do not affect individual prospects for outcome, or to paraphrase Rawls (1971, p.63), that individual with similar effort face “*the same prospects of success regardless of their initial place in the social system*”. However, there remains considerable discussion regarding what factors should count as effort or circumstances.

A prominent view in these debates is the one expressed by John Roemer in a series of contributions. It claims that the definition of circumstances is a matter of political choice. Furthermore, once circumstances have been defined “by society” remaining differences in individual outcomes should be considered the result of effort. Hence, the distinction between circumstances and effort turns into a dichotomic partitioning of the determinants of outcome. As a consequence, requiring that, for a given level of effort, circumstances do not affect individual prospects for outcome, implies that individuals with similar effort should have *equal* outcomes.

This dichotomic approach lies at the heart of most economic analysis of equality of opportunity. However, it does not fully account for the diversity of the determinants of outcome and leads to a specific conception of equality of opportunity. Assuming that society has agreed on a given set of circumstances does not imply that the remaining determinants will reflect individual responsible choice and should be treated as effort. In this respect, international attitudes surveys, such as the one summarized in Figure, reveal two noteworthy differences across countries. Consider that “social injustice” captures inequalities arising from circumstances, as defined “by society”. First, countries differ in their propensity to consider that bad economic outcomes reflect social injustice, which indicates that the definition of circumstances may vary across societies. Second, and more importantly, the figure also suggests that countries differ in their belief in the role of effort in shaping individual outcomes, over and beyond the influence of circumstances. The assumption that the determinants of outcomes excluded from socially defined circumstances relate to individual effort provides a good approximation of US average beliefs. It does not however correspond to the social perception in many European countries, which emphasizes the role of luck in shaping individual success.

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2For a theoretical discussion, see Roemer (1993, 1998) and for empirical applications Betts & Roemer (2006), Roemer et al. (2003) and Dardanoni et al. (2005).


4Roemer (1993, p.149)

Our purpose is to build a model of equality of opportunity flexible enough to encompass this diversity of perceptions. This requires to distinguish three generic determinants of individual outcomes: circumstances, effort and luck. As described in the philosophical literature, justice does not necessarily command that the impact of every kind of luck be nullified. In some cases, luck may appear as a fair source of inequality provided that it is even-handed. Consequently, it is too strong a requirement to define equality of opportunity as a situation where individuals with similar effort reach equal outcomes. What equality of opportunity requires is that, given effort, no one faces more favorable outcome prospects, as a result of luck, for reasons related to differential circumstances.

The first contribution of this paper is to offer a characterization of equality of opportunity consistent with this view. Given effort, the outcome prospects of an individual are summarized by the outcome distribution conditional on her circumstances. Our characterization rests on the idea that equality of opportunity prevails when the conditional outcome distribution attached to all possible circumstances cannot unanimously be ranked in terms of well-being, using the tools of stochastic dominance (first and second order). The introduction of luck as a determinant of outcome fully legitimates the use of stochastic dominance instruments, and specifically of second order. Because choosing among the outcome prospects falls down into decision-making problems under risk. If a decision maker is unable to choose among the different prospects, then EOP prevails. This leads us to distinguish several definitions of EOP.
The empirical implementation of these definitions of equality of opportunity would be straightforward if circumstances and effort were observable. However, in practice, this condition may not be easily met. In most data sets, not all the relevant aspects of individual effort can be measured and only a subset of the relevant circumstances can be observable. We discuss the consequences of these limitations for the evaluation of equality of opportunity. The second contribution of the paper is to show that, conditional on further distributional assumptions, it is still possible in some cases to provide testable conditions for equality of opportunity when effort and circumstances are not fully observed.

We then develop an empirical analysis of equality of opportunity for income acquisition in France, using household surveys over the period 1979-2000. In this application, we assume that circumstances are defined by individual social background, measured by father’s occupation and we compare income distributions conditional on social origin. Our analysis of these income distributions relies on non-parametric tests of stochastic dominance developed by Davidson and Duclos (2000).

The rest of the paper is organized as follows. Section 2 discusses our characterization of equality of opportunity. We first review the various conceptions of equality of opportunity that have been discussed in recent philosophical debates and then develop a comprehensive model that accommodates these various conceptions. In section 3, we discuss the identification of equality of opportunity when the relevant determinants of outcome are only partially observable. In section 4 we develop an empirical analysis of equality of opportunity for income in France.

2 Equality of opportunity : Definitions

2.1 Luck and equality of opportunity : a brief review

In the philosophical debates on equality of opportunity, the concept of luck refers to situations where individual control, choice or moral responsibility bears no relationship to the occurrence of outcomes. This broad concept includes the notions of circumstances and luck that we previously referred to. The general idea, shared by many authors, is that inequality related to luck should be compensated: as noted by Fleurbaey (2008) “Luck-egalitarianism has even become the catchword for responsibility-sensitive egalitarianism in a part of the literature”.

7In welfare economics, the analysis of social situations in the presence of risk has revealed a conflict between the ex ante and ex post approaches. See Diamond (1967), Hammond (1981), Broome (1991), Ben Porath et al. (1997), Gajdos and Maurin (2004) and Fleurbaey (2006). This perspective is however essentially absent from philosophical debates on equality of opportunity.
However, according to some authors, this egalitarian requirement may conflict with other values that should receive priority. This leads to distinguish several varieties of luck.

2.1.1 Varieties of luck

The debates about equality of opportunity have singled out at least four ideal-type notions of luck that can be illustrated by simple empirical examples.

First, consider two equally talented and motivated individuals whose outcome differ only because of differences in their family’s social connections. In this situation individual actions and their results are pre-determined by antecedent factors (family and social origin). This illustrates the idea of social lottery developed by Rawls. It is most probably the first candidate to be considered as a circumstance. We propose to call it social background luck.

Second, consider two fraternal twins whose outcome differ only because one of them genetically inherited a special talent. As in the previous example, the determinant of differential success, talent, lies beyond the realm of individual choice or control. One important difference with the previous form of luck is that a specific individual talent can be seen as constitutive of the individual, in the sense that it defines what person she is. This second example illustrates the notion of constitutive luck, or Rawls’ idea of a natural lottery. It includes genetically inherited factors and we therefore propose to call it genetic luck.

Third, consider two individuals with similar talent and social background. Their outcomes differ as a result of a lottery they could not escape. For instance, as a result of the Vietnam draft lottery, one of them is inducted into the Army and subsequently enjoys poor outcomes, but not the other. This is a special form of Dworkin’s notion of brute luck, which represents a situation where the individual cannot reasonably impact the probability of an event taking place. This kind of luck can occur at any time over a life course. Vallentyne (2002) distinguishes two types of brute luck. Initial brute luck is defined as the set of factors that influence lifetime prospects up to the moment when individuals can be considered responsible for their choices and decisions. This roughly corresponds to Arneson (1990)’s idea of a “canonical moment” where individuals become responsible for their choices and preferences. By contrast, later brute luck denotes the luck factors that affect individual outcomes after the canonical moment. Our example illustrates later brute luck.

Fourth, consider two individuals who both have to choose among two lotteries. The outcome of the first lottery is certain. The outcome of the second is random. Assume that individuals make different choices and end up with different outcomes. The occurrence of outcomes partly

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8See for instance the discussion in Dardanoni, Fields, Roemer & Sanchez Puerta (2005).
escapes individual control, although by making different choices, one can influence the occurrence of outcomes. This corresponds to Dworkin’s notion of option luck. This notion implies that risk is taken deliberately, is calculated, isolated, anticipated and avoidable. We assume it is the case in our example and refer to it as informed option luck.

2.1.2 The requisites of equality of opportunity

Whether (and how), from an egalitarian perspective, these different varieties of luck ought to be compensated has been the subject of numerous papers. Their main (unconsensual) conclusion is that not all types of luck singled out in the previous paragraphs call for full compensation.

Almost all authors would agree that social background luck should be fully compensated, resorting to the ‘starting gate position’ argument: some deep inequalities of life prospects related to economic and social circumstances of birth cannot be justified by appeal to merit and desert (Rawls 1971). By full compensation, we mean that justice requires that outcomes be equal regardless of social background luck, other things being equal.

A similar argument applies to the effects of genetic luck on individual outcomes. However, given the constitutive nature of genetic luck, compensation of its impact may conflict with other ethical values. Hence, it has been claimed that genetic luck should not be compensated, owing to the libertarian principle of self-ownership which states that agents are entitled to the full benefit of their natural personal endowments (e.g. intelligence, beauty, strength) (Nozick 1977, p.225). For some authors, this requirement should receive priority over other principles.

From a moral point of view, compensation for all forms of luck has also been contested on efficiency grounds. The cost of this compensation can obviously be quite high. Such compensation requires considerable (and costly) information on individual situations as well as strong redistribution which may lead to large distortions, in the case of option luck. If these costs are large enough, compensating for all forms of luck may diminish the overall well-being. This has led some authors to formulate a restricted requirement of justice, which only calls for the compensation of initial brute luck and avoids part of the cost of redistribution. According to Vallentyne (2002), justice only requires that the initial value of lifetime prospects be equal across individuals, where the initial value is computed at the onset of adulthood. This re-
quires compensating for *initial brute luck*. Of course, to the extent that *later brute luck* is related to *initial brute luck*, compensation for the latter implies (at least partial) compensation for the former. However, equalizing the value of *initial* lifetime prospects does not erase all the impact of *later brute luck* on individual outcomes and individual can still end up, *ex post*, with different outcomes as a result of brute luck. It simply makes sure that *later brute luck* is *ex ante* even-handed.

Lastly, three distinct views are held regarding the compensation for *informed option luck*. To the extent that the risky outcomes of option luck are avoidable and result from individual choice, some authors have claimed that inequalities resulting from option luck should not be compensated, owing to the principle of natural reward which states that the consequences of individual choice should be maintained. Dworkin supports that idea. A second view, expressed for instance in Fleurbaey (1995), recommends full compensation of the outcomes of option luck. Two distinct arguments are given in favor of this proposal. First, the fact that pure option luck is an extremely restrictive notion of luck that is both very rarely met in practice and very difficult to assess empirically. Second, and more importantly, the fact that not compensating for the effect of option luck implies that small errors of choices involve disproportionate, and thus unfair, penalties for some individuals. Lastly, some authors take an intermediate stance and argue in favor of partial compensation of the effects of option luck. Vallentyne (2002), states that equity authorizes taxation of the results of good option luck to partly compensate individuals who suffered bad option luck. Fleurbaey (2008) proposes to decompose the outcome of option luck into two components: on the one hand, the individual choice of a specific lottery, which belongs to responsibility factors; on the other hand, the intrinsic randomness of outcome attached to any lottery, which is akin to luck. According to Fleurbaey, the former component should not be compensated while the latter should to some extent be compensated. This view is close to the one of Le Grand (1991) who claims that individuals who make similar gambling decisions should enjoy equal well-being and therefore be fully insured and that individuals who make different decisions should only bear the consequences of their gambling over their expected well-being.

(1989) who suggests that equality of opportunity should be defined by the equality of “*preference satisfaction expectations*”.
2.2 A model of equality of opportunity: circumstances, effort and luck

Our purpose is to build an economic model of equality of opportunity flexible enough to accommodate the diversity of positions held in ethical debates. The previous section reveals the lack of agreement regarding how non-responsibility factors should be accounted for in the definition of equality of opportunity. However, three main conclusions emerge from this analysis. First, the idea that social background luck should be included in the set of circumstances seems beyond dispute. Second, some non-responsibility factors could be excluded from the set of circumstances. Third, those non-responsibility factors excluded from the set of circumstances differ from individual effort to the extent that they do not necessarily relate to individual responsibility.

It seems clear from the above discussion that this model should incorporate three types of factors: circumstances denoted by a vector $c$ capture the non-responsibility factors that are not considered a legitimate source of inequality; effort summarized by a scalar $e$ includes the determinants of outcome that pertain to individual responsibility and that are consequently seen as a legitimate source of inequality; luck is captured by a scalar $l$ and this concept, from now on, only comprises the non-responsibility factors that are seen as a legitimate source of inequality as long as they affect individual outcomes in a neutral way, given circumstances and effort. In this paper, we take a neutral stance on the question of what factors should count as circumstances, effort or luck, which, in our view, pertains to moral or political debates.

Define $y$ the individual outcome and $F()$ its cumulative distribution function, which is assumed to be continuous. A type defines the set of individuals with similar circumstances. A variety denotes the set of individuals with similar circumstances and effort.

At this point, an important remark is in order. Among the four parameters, income, circumstances, effort and luck, we only have three “degrees of freedom”. For instance, once the distribution of outcome, circumstances and effort have been identified, we can retrieve luck. More precisely, in this setting, the overall impact of luck can be measured by the level of outcome that an individual reaches for a given level of effort and circumstances, for by definition, lucky individuals are the one who enjoy higher outcomes. The distribution of outcome conditional on circumstances and effort, $F(y|c,e)$, measures how luck affects the outcomes of individuals of a given variety. It gives the odds of all possible outcomes that may occur for an individual of this variety, as a result of the influence of luck. Alternatively, without loss

\footnote{See below, § 3.4 the discussion about the ex ante and ex post approach.}
of generality, luck may be summarized by a scalar index $l$. In this case, let $Y(c,e,l)$ denote the outcome function. Again, by the very definition of luck, this function must be strictly increasing in $l$. An example of such an index $l$ may be defined by the “rank” where the individual sits in the distribution of outcome conditional on her variety: $l = F(Y|c,e)$. So arbitrarily defined, $l$ measures the relative degree of luck, within a given variety. By construction, $l$ is identically distributed across varieties, which does not imply that a given degree of luck is associated with similar outcomes regardless of circumstances and effort.

2.2.1 A strong criterion of EOP

In this context, what does equality of opportunity require? Since inequalities related to effort are morally acceptable, equality of opportunity should only apply among individuals with similar effort. Precisely, individuals with similar effort should face similar prospects for outcome, regardless of their circumstances. This is equivalent to say that given effort the distribution of outcome should not depend on circumstances. In turn, this requires that luck be even-handed, in the sense that the distribution of (the results of) luck is independent of circumstances. This criterion can be formalized by the following definition.

**Definition (EOP-S)**

Equality of opportunity is satisfied iff: \( \forall (c,c') \forall e, F(.|c,e) = F(.|c',e). \)

One should emphasize that EOP-S only provides a formal definition of equality of opportunity. This formal definition is compatible with a variety of substantive conceptions of equality of opportunity, depending on one’s view of how all the concrete determinants of outcome should be classified as circumstances, effort and luck. As already mentioned, the criterion of individual responsibility, put forward by Cohen (1989), Arneson (1989) and Roemer (1993), offers a moral principle that can serve to define effort. This perspective is, however, in no way essential to our analysis. As suggested by the discussion in section 2.1, alternative principles, such as the principle of self-ownership, may serve to define our generic notion of effort. What matters to our analysis is that inequalities originating in differential effort are seen as legitimate and do not call for compensation. To give another illustration, if we consider that effort includes talent and ability, definition EOP-S leads to the Rawlsian conception of “fair equality of opportunity”, defined as a situation where “those who are at the same level of talent and ability, and have the same willingness to use them, should have the same prospects of success regardless of their initial place in the social system” (Rawls 1971, p.63).

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14 A similar definition is adopted in Fleurbaey (2008).
The present model also allows to analyze equality of opportunity in the presence of option luck. In this case, the effort variable should comprise the individual preference towards risk that condition her choice of a particular lottery. Under this interpretation, the condition in EOP-S implies that individuals with similar preferences should face similar lotteries independently of their circumstances. This is consistent with the view held by Dworkin. The compensation for option luck suggested by Le Grand (1991) would amount to place further restrictions on the effect of luck, conditional on effort, \( i.e. \) it would require that \( F(|c,e) \) be a mass point distribution, at the expected value of well-being for a given choice of lottery.

Finally, two particular cases of this general model can be mentioned. First, a model where outcomes are only influenced by circumstances and luck as in the analysis of schooling outcomes undertaken in Jencks et al. (1972). In this case, EOP-S degenerates to the requirement that the distribution of outcome conditional on circumstances, \( F(y|c) \) should be independent of \( c \). Luck can lead to differences in individual outcomes as long as it remains neutral with respect to circumstances.

The second model corresponds to the world described by Roemer where luck is not considered per se. As the author admits (Roemer et al., 2003, p541), in his model, “luck [...] unfortunately [...] appears as effort, because effort [is measured] as a residual after circumstances are accounted for”. In that case, \( F(.|c,e) \) degenerates to a point-mass at \( y(c,e) \). EOP-S requires that the income \( y(c,e) \) be independent of circumstances.

Now we explore another possible justification for EOP-S, in a welfarist setting inspired by Arneson (1989) and Vallentyne (2002), and in which welfare should be equal across individuals, only to the extent that they exercise the same degree of responsibility.

2.2.2 A welfarist foundation

Arneson (1989) and Vallentyne (2002) propose to use the expected value of future prospects as the relevant metric for evaluating opportunities. This is coherent with the idea that, from an \textit{ex ante} perspective, given individual circumstances, luck, and consequently outcomes, may be seen as random processes. In this context, equality of opportunity can be defined by the equality of the expected value of future prospects across individuals. To perform this, one can use a specific Von Neumann-Morgenstern utility function \( u \) and compute the expected utility of the opportunities for outcome offered to a given type. In this case, equality of opportunity is defined by the following proposition:
Definition

Equality of opportunity is satisfied iff: \( \forall (c, c'), \int_y u(y)dF(y|c, e) = \int_y u(y)dF(y|c', e). \)

However, the question of what utility function to choose remains problematic. Ideally, one would like the characterization of equality of opportunity to hold for a sufficiently broad class of utility functions. In the case where there is a natural ordering for the outcome under consideration, as is the case for income, it is reasonable to focus on monotone increasing and concave utility functions. In this context, it is obvious that the expected value of future prospects attached to different circumstances \( c \) will be equal, for all possible increasing concave utility functions if and only if the income distributions for these circumstances are equal. Hence, we get a welfarist foundation to EOP-S.

2.2.3 Two weak criteria of EOP

Even without placing any further restriction on the distribution of income, the situation characterized by EOP-S appears as a situation of strong equality of opportunity. This condition is very stringent and may not easily be satisfied in practice. Consequently one may wonder whether all situations where EOP-S is violated should be considered equivalent from the point of view of equality of opportunity.\(^{15}\)

Assume that EOP-S is not satisfied for two types \( c, c' \) and a given effort level. Two situations can arise. First, for all relative degrees of luck, one type, say \( c \), always gets higher outcome than the other (\( \forall l, Y(c, e, l) \geq Y(c', e, l) \) and the inequality is strict for some levels of \( l \)). Second, one type gets higher outcomes for some degrees of luck while the other type gets higher outcomes for other degrees of luck (for instance, unlucky type-\( c \) do better than unlucky type-\( c' \) but lucky type-\( c \) do worse than lucky type-\( c' \)). Now consider the hypothetical situation of someone who would be given the option to choose between circumstances \( c \) and \( c' \), without knowing her degree of luck. This is a typical case of choice under risk. In the first case, the outcome distribution associated with \( (c, e) \) stochastically dominates the one associated with \( (c', e) \). There is a large agreement among specialists of decision theory (Starmer 2000) to say that in this case, consistent preferences under risk, should lead to choose \( c \) over \( c' \).\(^{16}\)

\(^{15}\)Empirically, this question seems highly relevant. For instance Dardadoni et al. (2005) and O’Neil et al. (2000) both test a condition close to EOP-S and conclude that it is violated. However, they do not offer a formal ranking criterion for situations in which this condition is violated.

\(^{16}\)This consensus reaches well beyond the Expected Utility Theory. There is also empirical support for that view. In some experiments (Birnbaum & Navarette 1998) individual choices may not accord with the first-order stochastic dominance criteria. However, as argued by Starmer (2000) this may occur in situations where stochastic dominance is opaque to the agent.
clear case of inequality of opportunity while the second corresponds to a weak form of equality of opportunity, where no set of circumstances yields an unambiguous advantage over the other.\textsuperscript{17}

Avoidance of first-order stochastic dominance is however a very weak requirement to define equality of opportunity and one may object that this condition is not restrictive enough. For instance, it would consider that equality of opportunity prevails between \( c \) and \( c' \) in the case where all agents of type \( c \) do worse than those of type \( c' \) except for the one with the highest relative degree of luck. One may provide a more restrictive definition of weak equality of opportunity by resorting to the criterion of second order stochastic dominance (SSD). It is indeed commonly assumed that the Von Neumann-Morgenstern utility function exhibits risk-aversion, which corresponds to the case where \( u() \) is concave. It is well known\textsuperscript{18} that the expected value derived from a distribution \( F(y|c,e) \) will be greater than the one derived from \( F(y|c',e) \) for all increasing concave utility functions if and only if \( F(y|c,e) \) stochastically dominates \( F(y|c',e) \) at the second order, which we will denote \( F(\cdot|c,e) \succ_{SSD} F(\cdot|c',e) \).\textsuperscript{19}

Using these notations, we get the following definition of weak equality of opportunity under risk aversion:

**Definition (EOP-W1)**

Weak Equality of opportunity under risk aversion is satisfied iff:

\[
\forall c \neq c' \forall e, F(\cdot|c,e) \not\succ_{SSD} F(\cdot|c',e).
\]

The conception of equality of opportunity embodied in definition EOP-W1 states that, whatever the value of effort, it is never possible to rank the opportunities offered by different types using second order stochastic dominance. Hence this notion of equality of opportunity rules out situations in which, for at least one effort level, one type would provide an unambiguous advantage (in the sense of stochastic dominance) over other types. A special case of such advantage, which, admittedly, represents a strong form of inequality of opportunities, is the situation where one type would provide an advantage over other types for all values of effort. An even weaker form of equality of opportunity than EOP-W1 is to rule out the latter possibility. This leads to the following definition:

\textsuperscript{17}Van de Gaer (1993) and Van de Gaer et al. (2001) already proposed the use of stochastic dominance tools to define equality of opportunity in a different framework.

\textsuperscript{18}The requirement that choices under risk be consistent with the principle of second-order stochastic dominance (SSD) stated below does not require that the Von Neumann - Morgenstern axioms be satisfied. Machina (1982) proved that this property is valid under more general conditions within the context of non-expected utility theories.

\textsuperscript{19}Formally, the definition of second-order stochastic dominance is the following:

\( F(\cdot|c,e) \) strictly stochastically dominates \( F(y|c',e) \) at the second order \( (F(\cdot|c,e) \succ_{SSD} F(\cdot|c',e)) \) iff: \( \forall x \in \mathbb{R}_+ \int_0^x F(y|c,e)dy \leq \int_0^x F(y|c',e)dy \) and \( \exists x \mid \int_0^x F(y|c,e)dy < \int_0^x F(y|c',e)dy \).
Definition (EOP-W2)

Very weak Equality of opportunity is satisfied iff:

\[ \exists (c, c') \text{ such that:} \]

\[ \forall e, \quad F(.|c, e) \geq_{SSD} F(.|c', e) \]

and

\[ \exists e \text{ such that } F(.|c, e) >_{SSD} F(.|c', e). \]

If EOP-W2 is satisfied, two situations can occur: either there exists no effort level for which one type is unambiguously advantaged (which corresponds to EOP-W1), or, if there exists one effort level for which one type is advantaged, than the same type is disadvantaged for at least one other effort level. In both cases, no type can be unambiguously preferred when effort has not been chosen yet. Hence, although it is a weaker requirement, EOP-W2 still represents an appealing notion of equality of opportunity.

3 Equality of opportunity: Empirical identification

If circumstances and effort are observable, it is straightforward to examine whether the requirements of EOP-S, EOP-W1 or EOP-W2 are empirically satisfied. However, in many cases, some of that information may be missing. In this section we discuss to what extent equality of opportunity can be assessed when effort or circumstances are not fully observable to the empirical analyst and derive empirically tractable implementation criteria (IC) for equality of opportunity. We first consider the case where only outcome and circumstances are fully observable.

3.1 Unobservability of effort

We only observe the distribution of outcome conditional on circumstances \( F(y|c) \) while in a complete information setting the conditions for equality of opportunity are stated in terms of the distribution of outcome conditional on circumstances and effort, \( F(y|c, e) \). Letting \( G() \) denote the cumulative distribution of \( e \), the relationship between the two distributions is given by:

\[ F(y|c) = \int_{e} F(y|c, e)dG(e|c). \]  (1)

Here, one may think of assessing equality of opportunity on the sole basis of observable conditional distributions. One obvious candidate for an implementation criterion of EOP is the
equality of these conditional distributions.

**Definition (IC1)**

*Implementation criterion IC1 is satisfied iff:* \( \forall (c, c') \), \( F(c) = F(|c|) \).

Unfortunately, without additional conditions, IC1 does not provide an implementation condition for EOP-S. If effort is not observed, assessing EOP is only possible if effort is distributed independently of circumstances. In turn, this independence condition can be seen as an empirical property or as an ethical requisite. This second option has been argued forcefully by Roemer. We begin by detailing Roemer’s argument in a world where luck plays no role, that is, in the Roemer model.

### 3.1.1 Unobservability of effort in Roemer’s model

Up to now, we have implicitly taken an *absolutist* view on effort, as supported for instance by Barry, which amounts to assume that effort can be defined in itself, independently of its relationship with circumstances. This view is not indisputable. As Roemer argues, if we are to take seriously the idea that individuals are not responsible for their circumstances, the definition of effort needs to be purged of any residual influence of circumstances. This leads to a *relativist* conception according to which effort is, by construction, distributed independently of circumstances: "The choice of a degree of effort (as measured by the percentile of effort levels within a type) as the relevant metric for how hard a person tried, is justified by a view that, if we could somehow disembowel individuals from their circumstances, then the distribution of the propensity to exert effort would be the same in every type."[21] This makes it clear that the question of whether \( e \) is distributed independently of \( c \) is not solely, nor primarily, an empirical matter related to the identification of equality of opportunity under partial observation of the determinants of outcome. It pertains to more fundamental ethical debates.[22] Of course, not everyone would subscribe to the relativist view of effort. It is far from obvious that Roemer’s argument carries over to the case where effort also includes non-responsibility factors such as genetic or option luck.

Roemer’s conception of effort can be formalized by a function of some objectively measurable effort variable \( e \) and circumstances \( c \), \( e^+(e, c) \), such that the distribution of \( e^+(e, c) \) is independent of \( c \). A natural candidate is \( e^+(e, c) = G(e|c) \), the rank in the distribution of effort.

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21Roemer (1998, p.15)
22Along similar lines, see the example developed in Cohen (1989, pp917-921).
Adopting this relativist view makes it very easy to obtain an implementation condition for EOP in the Roemer setting, without luck. Since outcome can be expressed as a function $Y(c,e^\perp)$ that only depends on relative effort and circumstances, EOP-S implies that individuals with similar relative effort should receive equal outcomes regardless of circumstances. Roemer further assumes that the outcome function is strictly increasing in effort. In this case, although effort is not directly observable, relative effort $e^\perp$ can be inferred from the observation of outcome and circumstances. For an individual with circumstances $c$, $e^\perp$ can be assessed by the rank $p$ where she sits in the conditional outcome distribution, $F(y|c)$. Let $Q(p|c)$ denote the quantile function associated to the distribution $F(y|c)$ and defined by $p = F(Q(p|c)|c)$. EOP-S requires that $\forall (c,c'), \forall p \in [0,1], Q(p|c) = Q(p|c')$ which is equivalent to require IC1. Hence we have the following proposition.

**Proposition 1**

Under Roemer’s assumptions : EOP-S $\iff$ IC1.

Consequently, under the assumption that outcome is strictly increasing in effort and a relativist view of effort, IC1 is a necessary and sufficient condition for EOP-S to hold in a model without luck. We now examine whether this proposition remains valid when luck is introduced.

### 3.1.2 Unobservability of effort in the general model

One interesting case arises when the distribution of effort is independent of $c$ or if we stick to the relativist view of effort. Whether effort should also be purged of the influence of luck is an opened question that has not been addressed in the normative literature. Two opposite arguments seem relevant from an ethical perspective. On the one hand, luck is beyond individual responsibility, which suggests that its influence on effort should be nullified. In this case, relative effort would be defined by $e^\perp\perp = G(e | c,l)$. On the other hand, luck differs from circumstances in the sense that we want to compensate circumstances but not luck, as long as it remains neutral. If luckier individuals exercise more effort, their higher outcome may be seen as legitimate. In this case, effort should only be purged of the effect of circumstances. These two points of view lead to different conceptions of equality of opportunity. However it is important to emphasize that they yield similar testable restrictions, in the case where effort is not observable. Consequently,

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23 A technical condition is required for $e^\perp(e,c)$ to be properly defined. The distribution of $e$ conditional on $c$ should not exhibit any mass point.

24 For a complete discussion of the conditions of identification of equality of opportunity in Roemer’s model, see O’Neill, Sweetman & Van De Gaer (2000).

15
without loss of generality, we will adopt the second view and define relative effort as effort net of the influence of circumstances, i.e. $e^\perp(e,c) = G(e|c)$.

Of course, if conditional on circumstances, outcome is determined by effort and luck, the rank in the distribution of outcome conditional on $c$ can no longer serve to identify effort, even if we assume that outcome is strictly increasing in effort, as the rank reflects the joint impact of luck and effort. This explains why, in the general model, we only have necessary (but not sufficient) conditions for EOP-S.

**Proposition 2**

If either $\forall c, G(e|c) = G(e)$ or if we substitute $e^\perp$ for $e$ in the definition of EOP-S then EOP-S $\implies$ IC1.

**Proof**: If for all $c$, $G(e|c) = G(e)$ and if EOP-S is satisfied, then, from equation 1 we get:

$$F(y|c) = \int e F(y|e) dG(e).$$

Hence $F(.|c)$ is independent of $c$ and IC1 is satisfied.

Hence IC1 is a necessary condition for EOP-S under the relative view of effort. Note however that it is not a sufficient condition. For instance, consider two values of circumstances $c$ and $c'$.

Assume that the opportunities offered to low effort type-$c$ individuals are offered to high effort type-$c'$ individuals and vice versa. In this case, IC1 will be satisfied but not EOP-S.

We now turn to the assessment of weak equality of opportunity, as defined by EOP-W1. Consistent with the relativist view of effort, we assume that effort is independent of circumstances. The consequences of the unobservability of effort are more serious for asserting EOP-W1 than for EOP-S. In a nutshell, weak equality is defined by some inequalities, conditional on effort and circumstances, that do not survive well to integration over effort levels. Hence, in general, it is not possible to assess EOP-W1 without observing effort.

However, when effort is independent of circumstances, we get a sufficient condition for preserving EOP-W2, given by the following implementation criterion.

**Definition (IC2)**

Implementation criterion IC2 is satisfied iff:

$$\forall c \neq c', \ F(.|c) \not\succ_{SSD} F(.|c').$$

This condition rules out situations of strict stochastic dominance across all possible pairs of types. It is sufficient to ensure the weakest conception of EOP.

**Proposition 3**

If $\forall c, G(e|c) = G(e)$ then: IC2 is a sufficient condition for EOP-W2.
Proof: Assume that for all \( c \), \( G(e|c) = G(e) \) and that \( \exists c, c' \) \( F(.|c) \not\succ_{SSD} F(.|c') \). Then, from equation \( \square \) and the definition of second-order stochastic dominance, we get for this pair \( (c, c') \) \( F(.|c) \succ_{SSD} F(.|c') \). Hence EOP-W2 is violated.

3.2 Partial observability of circumstances

We now consider the case where effort and some circumstances are not observable. The vector of observable circumstances is denoted by \( c_1 \) and the vector of unobservable circumstances by \( c_2 \). In this case, it is still possible to provide a necessary condition for EOP-S when effort is independent of circumstances without any other assumption. This is summarized by the following proposition.

**Proposition 4**
If \( \forall c, G(e|c) = G(e) \) then : EOP-S \( \implies \) IC1.

**Proof**: If EOP-S is satisfied, we have : \( \forall (c_1, c'_1), \forall c_2, \forall e, \forall y, F(y|c_1, c_2, e) = F(y|c'_1, c_2, e) \).

Furthermore, if \( e \) is distributed independently of circumstances, integrating over values of \( e \) implies : \( \forall (c_1, c'_1), \forall c_2, F(.|c_1, c_2) = F(.|c'_1, c_2) \)

Integrating over values of \( c_2 \) implies : \( \forall (c_1, c'_1), F(.|c_1) = F(.|c'_1) \).

Roemer has suggested a similar result in his own setting without luck. It is important to stress that our result does not require the independence of unobserved circumstances vis-à-vis observed circumstances is needed. In this setting, as in Bourguignon, Ferreira & Menendez (2007), observed circumstances will also capture the influence of any correlated unobserved circumstances.

The reciprocal is however not true. Loosely speaking, proposition \( \square \) simply states that the values of \( F(.|c_1, c_2) \) and \( F(.|c'_1, c_2) \) are equal on average, where the average is computed over values of \( c_2 \). Of course, this does not imply that the CDF are equal for all values of \( c_2 \).

Again, assessing EOP-W1 is not achievable under partial observability of circumstances. However a special case of EOP-W2 can be assessed, which corresponds to the case where no visible type \( c_1 \) dominates a visible type \( c'_1 \), for all possible efforts, whatever the value of their unobservable circumstances \( c_2 \). That is

\[
\forall (c_1, c'_1) \forall (c_2, c'_2), \forall e, F(.|c_1, c_2, e) \succ_{SSD} F(.|c'_1, c'_2, e) \tag{EOP-W3}
\]

It is still possible to provide a sufficient condition for the avoidance of EOP-W3, which is given in the following proposition.
Proposition 5

If $\forall e, G(e|c) = G(e)$ then IC2 is a sufficient condition for EOP-W3.

Proof: Negating EOP-W3 and integrating the dominance relation over effort and unobservable circumstances implies that $F(.|c_1) \succ_{SSD} F(.|c'_1)$. Hence if effort is independent of circumstances, IC2 is violated.

3.3 Ex ante versus Ex post

All conceptions of equality of opportunity discussed so far amount to equalize the prospects for outcome that individuals face, given their effort level. These prospects obviously refer to the ex ante distribution of outcomes, before the individuals know their degree of luck. Of course this ex ante distribution is never observable empirically. At best, what we can recover is the ex post distribution of outcome among individuals of a given variety. The empirical characterization of equality of opportunity undertaken in the previous sections rests on the assumption that the ex ante prospects offered to individuals of a given variety can be identified with the ex post distribution of outcomes among individuals of that type. This reduction hypothesis raises some empirical issues and requires specific conditions that, although unavoidable, deserve to be explicitly stated.

The first problem is common in statistical inference. In some cases, the size of the sample may be too small to prevent a noticeable difference between the probability and the frequency. In testing stochastic dominance, we have to keep an eye on the size of the sample.

The second problem is more specific to the reduction hypothesis implicitly made thus far. By construction, all individuals within a variety are offered similar ex ante prospects. But this will only correspond to the ex post distribution of outcome if individuals independently draw from the ex ante distribution. In some cases, however there are reasons to expect outcomes to be correlated within a type. For instance, the children of farmers are more likely to become farmers themselves and thus, are exposed, in any given year, to common weather shocks. Hence, the ex post distribution of yearly income may differ from the ex ante distribution of prospects. In this simple example, one may think that using repeated observations of yearly income may help solve the problem. But in some cases, there may be some path-dependency at the type level. For instance, it can occur that a generation of farmers has been hurt by bad weather conditions at the beginning of their professional life, which forced them to sell their land to banks and to become tenants for their remaining life. As a result, the income level of the whole

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25 We thank an anonymous reviewer for drawing our attention to this problem.
type has been lowered. In this case, although *ex ante*, their life chances of income may have been the same as for people living in cities (assumption), it is no longer not true *ex post*. But in this case, one may wonder how relevant the *ex ante* distribution turns out to be.

In summary, in any empirical implementation, whatever the cautions one can take, we have to admit that it is an *ex post* view that it is tested. Is it annoying? Since *ex ante* EOP may be always true if we go sufficiently back in time, a robustly observed *ex post* view may defended as the right one as well. As Marc Fleurbaey (2008) wrote “taking the *ex-post* point of view does not imply ignoring individuals’ *ex ante* life and sense of agency, it simply consists in looking at their *ex ante* life with the benefit of hindsight about the *ex-post* distribution of luck”. This is exactly what we attempt to do in practical terms.

3.4 Summary

To summarize the main conclusions of this section, equality of opportunity can be empirically assessed using the conditions IC1 and IC2, in the case where effort or some circumstances are not observable. We need to compare the cumulative distributions of income, conditional on observed circumstances. Since IC1 is only a necessary condition for equality of opportunity in the general model, we can only draw firm conclusions in the case where the cumulative distributions are not found equal. This case indicates that equality of opportunity, as defined by EOP-S is violated. The situation where the cumulative distributions are found equal is only indicative of equality of opportunity: we can only conclude to equality of opportunity if we are willing to consider that the determinants of outcome excluded from the circumstances only reflect either luck or effort.

Since it is also well-known that SSD is equivalent to generalized Lorenz dominance, IC2 requires to compare the generalized Lorenz curves associated with observed circumstances. When comparing two generalized Lorenz curves, three situations can occur: (a) the two curves are identical, (b) the two curves intersect, (c) one curve lies above the other. Case (a) is equivalent to the equality of the cumulative distributions, which has already been discussed. Case (b) entails very weak inequality of opportunity, as defined by EOP-W2 or W3. It also implies weak equality of opportunity in the sense of EOP-W1 if we are willing to assume that the determinants of outcome excluded from the circumstances resort to luck or to effort alone. Lastly, case (c) is suggestive of a deviation from equality of opportunity for two reasons: first, EOP-S is not satisfied; second we cannot even rule out the situation of strict dominance between two observed types. Case (c) corresponds to the situation of second-order stochastic dominance.
of one observable conditional distribution over the other. A special case of this situation is the case of first-order stochastic dominance. This case is worth investigating in its own right, since FSD signals a strong deviation from equality of opportunity with a full ranking of the opportunity sets offered to individuals with different circumstances whatever the attitude of the decision maker towards risk.


In this section, we analyze equality of opportunities for income in France. To this end, we examine whether income distributions conditional on social origin are equal or exhibit stochastic dominance patterns. We first present the data and the statistical procedure used in the analysis. We then discuss the results.

4.1 Data

The data come from the French household survey "Budget des Familles" (BdF) conducted by the French national statistical agency (INSEE). Five waves of the survey have been collected (1979, 1984, 1989, 1994 and 2000), each on a sample of about 12,000 households. We use all available waves. For each household, the data provide detailed information on all sources of income and expenditures and enable to identify the household’s social background. Sample summary statistics are presented in appendix table A-1.

4.1.1 Main variables

All waves of the BdF data contain information on the social origin of heads of household and their spouse. Both are asked to report the one-digit occupational group of their two parents. From these four variables, it would be possible to build a detailed classification of social origin. Given the size of our samples, the use of a detailed classification would lead to small sub-samples, and to inaccurate estimations of the income distributions conditional on social origin. For this reason, we only use information on the occupational group of the household head’s to define individual circumstances. This leads to distinguish the following six social origins: farmers, artisans, higher-grade professionals, lower-grade professionals, non-manual workers and manual workers.

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26 We use sample weight to ensure sample-representativeness.
27 For every survey, except 1979, it is the occupational group when the respondent was 16. In 1979, it is the last occupational group of the parents.
28 These six groups correspond to the French INSEE job classification. Children of artisans also include the children of small proprietors. The occupational groups of the 1979 survey have been recoded to account for the
The outcome variable we focus on is household standard of living. The BdF data provide a detailed record of all income sources including wage and labor income, asset income, transfers (pensions, unemployment benefits, child support, welfare benefits) as well as income and property taxes. We consider two measures of family income. The first one is primary income, which includes labor and asset income, and unemployment and pension benefits. The second one corresponds to disposable income and is equal to primary income plus redistributive transfers minus taxes. In both cases, we normalize family income by family size using the OECD equivalence scale. To make income measures comparable over time, income is expressed in constant terms (2002 Euros) using the consumer price index.

Consequently, our data set provide a comprehensive measure of one fundamental individual outcome: living standard. It also offer a characterization of an important determinant of this outcome, that most authors would agree to include among the set of relevant individual circumstances. On the contrary, it offers no measure of individual effort or luck and provides only an incomplete description of circumstances. Therefore, the empirical assessment of equality of opportunity that our data allow corresponds to the situation analyzed in section 3.2.

4.1.2 Sample selection rules

Within a given survey wave, changes in the social structure over time imply that the age composition will differ across groups of different social origin. For instance, the rise in the share of higher-grade professionals and the fall in the share of farmers implies that children of higher-grade professionals (respectively farmers) will be younger (resp. older) than the average. To avoid this composition effect, our sample is limited to households whose head was between 30 and 50 years old at the time of the survey. We also exclude households whose head was retired or student at the time of the survey. Another advantage of this sample selection rule is that household income will be more representative of their lifetime income (Grawe 2006). This leads to samples of about 4,000 households in each wave.

The early waves of the survey exhibit a high rate of non-response to the questions pertaining to income earned and taxes paid. In some case, information is missing for one or several income items. In others, respondents only report some income items in bracketed form. In these data, non-response cannot be considered random. It is correlated with the occupation of the head of household as well as with other socio-demographic characteristics, and is stronger for self-employed workers (farmers and artisans) than for wage earners. Hence, ignoring missing data would lead to a biased view of the income prospects conditional on social origin. For change in the occupational classification that occurred in 1982.
this reason, in case of non-response, household income has been imputed, using the simulated residuals method. For observations with missing or bracketed data, we predict income using an estimated income equation. This equation is estimated on those households who report an income (in level or in brackets), and income is regressed on observable characteristics (age, sex, occupational group, last diploma, consumption, nationality, family composition, geographic area of living ...). In case of missing data, we also draw a residual term that is added to the predicted income. This procedure is implemented to impute primary and disposable income. The income distribution estimated using this imputation procedure appear consistent with the distributions obtained from administrative data. This is true, in particular, for farmers’ income that was especially badly reported in the first waves of the survey.

4.2 Statistical Inference: General principles

Here, we explain the general principles of our statistical methodology. The details of the stochastic dominance tests we implement are presented in the Appendix. Our samples allow to build income distribution conditional on social origin and test whether conditions IC1 or IC2 are satisfied. Assessing equality or stochastic dominance relationships is a demanding exercise. It requires that the entire outcome distributions (or some integral of them) be compared for all possible circumstances. One should also bear in mind that these distributions need to be estimated and compared, in our case using samples of relatively small size. Hence, special attention must be paid to the statistical robustness of the conclusions drawn from sample data. In this context, performing parametric tests of stochastic dominance is likely to yield fragile conclusions. On the contrary, our empirical analysis rests on non-parametric stochastic dominance tests developed in Davidson and Duclos (2000), which lead to robust conclusions.

The empirical procedure we implement is the following. For all possible pairs of circumstances c and c′, we perform three tests independently: (1) we test the null hypothesis of equality of the distributions of types c and c′; (2) we test the null of first-order stochastic dominance of the distribution of type c over type c′ and vice-versa; (3) we test the null of second-order stochastic dominance of the distribution of type c over type c′ and vice-versa.

Test (1) corresponds to condition IC1 and tests (2) and (3) to IC2. For any pair of types, detailed equations are available upon request.

As we will discuss, part of the results reported below are driven by changes in the income distribution of farmers, estimated from our data. Consequently, we were concerned that part of these evolutions may reflect spurious changes in income distribution related to changes in reporting behavior. Hence we compared the estimated income distribution of farmers to the ones obtained from agricultural national accounts. As documented in Lefranc et al. (2004), it turns out that our estimates are strongly consistent with those obtained from administrative sources.
we interpret the joint results of these tests in the way summarized below. Of course, this interpretation is only temptative and one should keep in mind the caveats discussed in section 3.4.

- If we fail to reject the null of test (1), we say that equality of opportunity is supported, since IC1 is satisfied.

- Else, if test (2) or (3) accepts dominance of one distribution over the other but not the other way round (e.g. \( F(|c) \gtrless_{\text{SSD}} F(|c') \) and \( F(|c') \nless_{\text{SSD}} F(|c) \)) we say that equality of opportunity is violated, since neither IC1 nor IC2 are satisfied.

- Else, if test (3) rejects dominance of each distribution over the other (i.e. \( F(|c) \nless_{\text{SSD}} F(|c') \) and \( F(|c') \nless_{\text{SSD}} F(|c) \)) we say that weak equality of opportunity is supported, since IC2 is satisfied but not IC1.

- Else, if test (2) or (3) conclude that the two distributions dominate each other (i.e. \( F(|c) \gtrless_{\text{SSD}} F(|c') \) and \( F(|c') \gtrless_{\text{SSD}} F(|c) \)), we give priority to the result of test (1) since it is a more powerful test of equality of distributions for any significance level. Hence, we say that only weak equality of opportunity is supported, since IC2 is satisfied but not IC1.

Lastly, one should note that, given our interpretation, conclusions of test (2) and (3) cannot contradict since the null of (2) is included in the null of (3). Thus the conjunction of the results of the three tests interpreted in this way cannot be inconsistent.

4.3 Results

In this section, we first report the results of the tests of equality and stochastic dominance, for the income distributions conditional on our partition of social origin. While we use all available waves, we only report in the main tables and figures the results for 1979 and 2000, since our discussion mostly focus on the initial and terminal waves. Results for other years are reported in the appendix.

Three main conclusions emerge from this analysis. First, equality of opportunity is not satisfied: for most pairwise comparisons of types, we find evidence of stochastic dominance relationships; overall, a clear hierarchy of the different groups of social origin emerges. Second, the pattern of inequality of opportunity is stable over time: the relative ranking of types remains almost constant across the period 1979-2000. Third, the degree of inequality of opportunity decreases over time: while the ranking of types is unchanged, the income distributions of the different types come closer together over the period.
We then examine what factors account for inequality of opportunity. As discussed in decision theory, two factors may contribute to stochastic dominance between two income lotteries: differences in the expected return of the two lotteries and differences in the degree of risk of the two lotteries. In our case, the expected return corresponds to the mean income for each type and risk corresponds to within type inequality. Our results indicate that the degree of risk is very similar for all types. This is true over the entire period. On the contrary, the returns differ markedly across types. Since the evolution of these returns is a key determinant of the narrowing of the income prospect gap, we finally analyze the determinants of the changes in these returns over time.

4.3.1 A reduction in the degree of inequality of opportunity

The cumulative distribution functions and the generalized Lorenz curves conditional on social origin are given in Figures[2] and[3] for 1979 and 2000. For 1979, a particularly clear ranking of social types emerges. Children of higher-grade professionals stand out as the most advantaged type: their conditional distribution dominates by far those of other social groups. Children of lower-grade professionals come next, followed by children of artisans and children of non-manual workers. In fact, the income distributions of the latter two groups seem very close, especially in the first half of the distribution. Lastly, at the bottom of the social hierarchy, come the children of manual workers and the children of farmers. The income distribution of the children of farmers, in 1979, is, by far, dominated by all other social backgrounds.

This “visual” ranking is strongly supported by the results of the tests of equality and stochastic dominance. These results are presented in Table[1] for primary and disposable income. In 1979, in all but one pair-wise comparisons, the equality of the conditional income distributions is rejected. Without ambiguity, this indicates that EOP-S is not satisfied. The only two types who apparently face equal opportunities are the children of non-manual employees and of artisans, although one should keep in mind, here and in the rest of the paper, that IC1 is only a necessary condition for EOP-S. Furthermore in all other pair-wise comparisons, the tests indicate that one distribution dominates the other. Hence we cannot even dismiss the case where EOP-W2 is violated. One should also note that in all these cases, stochastic dominance is satisfied at the first order, which implies that a ranking of social types can be achieved without assuming risk aversion. Lastly, in 1979, the impact of taxes and transfer on inequality of opportunity, is very limited. The gap between the generalized Lorenz curves is slightly lower for disposable income than for primary income (see figure[3]), but stochastic dominance relationships are not affected
by redistribution.

The pattern of stochastic dominance relationships exhibits small changes between 1979 and 2000. The results of the tests for primary income are given in Table 1. Four important features can be underlined. First, the dominant position of the children of higher-grade professionals remains unchallenged during the entire period: in every wave their income distribution dominates those of all other groups. Second, the hierarchy of intermediate groups tends to weaken. This is in great part due to an improvement of the relative ranking of the children of artisans: in 1979, this type was dominated by children of lower-grade professionals and their opportunities were equal to those offered to children of non-manual workers; this is no longer the case after 1994. In 2000, the opportunities offered to children of artisans are equal to those offered to children of lower-grade professionals and dominate those of children of non-manual workers. In this intermediate group, one can also notice a change in the relative ranking of children of lower-grade professionals and non-manual workers. As the table in appendix demonstrates in 1989 and 1994, the income distribution of these two groups are equal, although it is no longer true in 2000. The third important phenomenon that takes place over this period occurs at the bottom of the hierarchy. In 2000 the conditional distribution of the children of farmers dominates the distribution of children of manual workers. This group is now dominated by all the others social backgrounds. Overall, a three-levels hierarchy persists over the entire period. It is dominated by the children of higher grade professionals. In the middle comes an intermediate group that includes the children of lower-grade professionals, artisans and non-manual workers. At the bottom, come the children of farmers and manual workers.

While the ranking of social backgrounds has changed, one may nevertheless be tempted to conclude that equality of opportunity has not made any good progress over the period 1979-2000. In all waves, equality of opportunity is rejected in at least 80% of all pair-wise comparisons. In 2000, IC1 is only satisfied in one case and IC2 prevails in another one. Hence, a strict ranking of all types is almost always possible.

IC1 is clearly violated in France. Given proposition 2, this implies that EOP-S is violated, for anyone who agrees that the fraction of any income determinants that is correlated with father’s occupation should be treated as a circumstance and not as effort, in the practical definition of equality of opportunity IC2 is also clearly violated. Then under the same proviso, given proposition 3, we cannot even dismiss that one type is dominated by some other type for every level of effort.

31 This would apply to determinants as diverse as hours of work, work diligence, genes, preferences, knowledge, education, connections, beliefs,...
Figure 2: Income distributions by social background - disposable income

Notes: The occupational group refers to social origin. H-grade prof.: higher-grade professionals; L-grade prof.: lower-grade professionals; Non-man. workers: non-manual workers.
Figure 3: Generalized Lorenz curves in 1979 and in 2000

A- 1979, primary income

B- 1979, disposable income

C- 2000, primary income

D- 2000, disposable income

Notes: Income in Euros 2002 is represented on the y-axis. The occupational group refers to social origin.


Despite this stability in the ranking of social backgrounds, the comparison of the generalized Lorenz curves for the first and the last wave (see figure 3) clearly indicates that the income distributions of the different types have come closer together between 1979 and 2000. This suggests that the change at work is more cardinal than ordinal. While inequality of opportunity continues to prevail, the degree of inequality of opportunity seems to weaken.

A cardinal assessment of this decrease in inequality of opportunity is now proposed in Figure 4. We compute an inequality of opportunity index consistent with our stochastic dominance approach and derived from the Gini coefficient. The detailed presentation of this so-called Gini-opportunity index (GO) and its properties is given in Lefranc et al. (2008). Over the

\[32\] The definition of this index comes as follows. Let us rank all types according to twice the area under its
Table 1: Stochastic dominance tests

<table>
<thead>
<tr>
<th></th>
<th>1979</th>
<th>2000</th>
</tr>
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<tbody>
<tr>
<td>Farmers</td>
<td>-</td>
<td>&lt;1</td>
</tr>
<tr>
<td>Artisans</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>-</td>
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<th>1979</th>
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<td>H-grade prof.</td>
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<td>L-grade prof.</td>
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<tr>
<td>Non-man. workers</td>
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Each element in the table indicates the result of the comparison of the income distribution of the groups in row and column using the tests presented in section 4.2:

- >: the row dominates the column for order i stochastic dominance;
- <: the column dominates the row for order i stochastic dominance;
- =: the distributions are equal;
- ?: the distributions cannot be ranked using first and second order stochastic dominance.
Chi two: Index of social immobility, Gini: index of income inequality, GO: index of inequality of opportunity.

Period we observe a sharp drop of the GO index at a decreasing rate. The level of inequality of opportunity in 2000 is 40% lower than in 1979. The same trend does not occur for the income inequality measured by the Gini coefficient which shows a much less drop. Interestingly, social immobility measured by the Chi-two index witnesses a different evolution. After a sharp drop up to 1995, social immobility starts to rise again slightly in the last period contrary to the GO. This divergence demonstrates that changes in inequality of opportunity do not only mirror changes in social immobility. We expand on that point in the last subsection.

We now turn to the analysis of this cardinal change. To do so, we analyze the degree of risk and the return attached to the different conditional income distributions.

Conditional GL curve, starting from the smallest one. For type \( s \), whose population share is \( p_s \), this area is equal to

\[
\mu_s(1 - G_s),
\]

with \( \mu \) denoting the mean and the \( G \) the Gini coefficient. The Gini-opportunity index is defined by:

\[
GO(x) = \frac{1}{\mu} \sum_{i=1}^{k} \sum_{j>i} p_i p_j (\mu_{j} (1 - G_j) - \mu_{i} (1 - G_i)).
\]

\( 33CHI2 = \sum_{j=1}^{6} \frac{(p_{jk} - p_j p_k)^2}{p_j p_k} \) where \( p_{jk} \) represents the proportion of individuals of social origin \( j \) and social destination \( k \).
4.3.2 The risk of social lotteries

As already discussed, stochastic dominance relationships among income lotteries can arise because of differences in their expected return or in their degree of risk. In the present context, the expected return is defined as the mean income conditional on social origin and the risk corresponds to within-type inequalities. Note that in the general model of section 2, the income lotteries offered to individuals correspond to the income distribution conditional on their circumstances and effort. Return and risk should be computed from these distributions. In our case, we do not observe individual effort. Hence, within-type inequalities will reflect the joint influence of luck and effort. This remark should be kept in mind throughout this section.

The degree of risk of the different social lotteries can be analyzed using the Lorenz curves of the distributions of income conditional on social origin. Consider two lotteries $A$ and $B$. The lottery $A$ is less risky than $B$ if its Lorenz curve is always above the curve of $B$. In this case, lottery $A$ is said to Lorenz-dominate lottery $B$. If the two Lorenz curves are identical then the two social lotteries are equally risky. Consequently, to compare the degree of risk of the different social lotteries, we can resort to the same testing procedure as for stochastic dominance.

Table 2 displays the results of the tests of Lorenz dominance. The Lorenz curves of the different social lotteries are very similar. For primary income, tests conclude to the equality of the Lorenz curves, in 9 comparisons (out of 15) in 1979, 5 in 1984, 12 in 1989, 11 in 1994 and 10 in 2000. The tests conclude to dominance in only one case in 1979, 3 in 1984, 2 in 1994 and 0 in 1989 and 2000. Hence, the different social lotteries exhibit very similar degrees of risk.

A slightly different picture emerges from the analysis of the Lorenz curves of disposable income, for the end of the period. First, in 2000 (as well as in 1994), the Lorenz curve of the children of manual workers dominates that of all most other groups, with the exception of the children of non-manual workers and lower-grade professionals. In other words, the income distribution for children of manual workers exhibits the smallest degree of risk. Second, at the end of the period, the income distribution of children of farmers and, to a lesser extent, artisans, tend to exhibit more risk than those of children of wage earners.

Overall, these results suggest that social origin mostly influences the distribution of outcomes as a scale factor. On the other hand, the combined influence of effort and luck is such that the relative prospects are roughly similar across types. This implies that income conditional on social origin can be represented by the following multiplicative model:

\[ \text{Effort can nevertheless be seen as a source of moral hazard and is therefore akin to a form of risk.} \]
## Table 2: Lorenz dominance tests

### A- Primary Income

<table>
<thead>
<tr>
<th></th>
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</thead>
<tbody>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Farmers</td>
<td>-</td>
<td>?</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>&lt;</td>
</tr>
<tr>
<td>Artisans</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>?</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>?</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td><strong>2000</strong></td>
<td></td>
<td></td>
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<tr>
<td>Farmers</td>
<td>-</td>
<td>=</td>
<td>?</td>
<td>=</td>
<td>=</td>
<td>&lt;</td>
</tr>
<tr>
<td>Artisans</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td>H-grade prof.</td>
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<td>-</td>
<td>=</td>
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<td>=</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
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<td>=</td>
</tr>
<tr>
<td>Non-man. workers</td>
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<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
</tr>
</tbody>
</table>

### B- Disposable Income

<table>
<thead>
<tr>
<th></th>
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<th></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>1979</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Farmers</td>
<td>-</td>
<td>?</td>
<td>=</td>
<td>=</td>
<td>&lt;</td>
<td>&lt;</td>
</tr>
<tr>
<td>Artisans</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>?</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td><strong>2000</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Farmers</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>?</td>
<td>&lt;</td>
<td>&lt;</td>
</tr>
<tr>
<td>Artisans</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>&lt;</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
<td>&lt;</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
<td>=</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>=</td>
<td>=</td>
</tr>
</tbody>
</table>


Each element in the table indicates the result of the comparison of the Lorenz curves of the groups in row and column using the criterion of Lorenz dominance:

- '>', the row dominates the column for Lorenz dominance;
- '<', the column dominates the row for Lorenz dominance;
- '=': the Lorenz curves are equal;
- '?': the Lorenz curves cannot be ranked using Lorenz dominance.
\[ y_{ic} = E(y \mid c)\epsilon_i \] (4)

Where \( y_{ic} \) represents income of individual \( i \) with social origin \( s \), \( E(y \mid c) \) is the income mean conditional on \( c \) and \( \epsilon_i \) a random term whose distribution is independent of social origin.

The strong similarity in the degree of risk attached to the different social backgrounds explains why in most cases, first-order stochastic dominance is a sufficient criterion for ranking conditional income distributions. In fact, in this context, equality of opportunity can be assessed by relying solely on comparisons of mean conditional incomes. This situation is a priori quite rare.

4.3.3 The expected return of social lotteries

We now compare the expected returns of the different social lotteries. These expected returns can be summarized by the mean income conditional on social origin.

Trends in conditional mean income Mean incomes conditional on social origin are given in Table 3. This table confirms the evolution apparent in Figure 3: mean conditional incomes tend to converge between 1979 and 2000. The ratio between the mean income of the most advantaged group (children of higher-grade professionals) and the least advantaged one (children of farmers in 1979 and of manual workers in 2000) falls from 1.89 to 1.46 for disposable income and from 2.04 to 1.63 for primary income. The advantage of children of higher-grade professionals falls by one half relative to the least advantaged group and diminishes relative to every other social background. More generally, the increase in the mean income of the different social groups is inversely related to their initial rank.

The increase in mean income is particularly strong for children of non-wage earners: children of farmers increase their mean income by 34.6%; children of artisans by 8.5%. As a result children of non-wage earners improve their relative position relative to all other social backgrounds. This contributes to the fall in the degree of inequality of opportunity since the initial ranking of both groups was relatively low: in 1979, children of artisans ranked third and children of farmers ranked last. The results for non-wage earners should be interpreted with caution due to the high rate of non-response discussed in section 4.1.2. However, the growth in non-wage earn-

---

35 The rare exceptions arise when comparing children of non-wage earners to children of wage earners. Due to the higher risk in the distribution for the former group at the end of the period, expected return and degree of risk sometimes point to opposite directions. For instance in 1994, as will be discussed in the next section, the mean disposable income is 5% larger for children of artisans than for children of lower-grade professionals. However, the stochastic dominance test turns out to be inconclusive due to the higher degree of risk of the lottery of children of artisans.
Table 3: Mean income conditional on social origin or destination

A- Mean income conditional on social origin

<table>
<thead>
<tr>
<th></th>
<th>Primary Income</th>
<th>Disposable Income</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1979</td>
<td>2000</td>
</tr>
<tr>
<td>Farmers</td>
<td>12 874</td>
<td>17 541</td>
</tr>
<tr>
<td>Artisans</td>
<td>19 295</td>
<td>20 174</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>26 375</td>
<td>24 543</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>21 225</td>
<td>20 511</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>17 379</td>
<td>17 720</td>
</tr>
<tr>
<td>Mean Income</td>
<td>16 503</td>
<td>18 313</td>
</tr>
</tbody>
</table>

B- Mean income conditional on social destination

<table>
<thead>
<tr>
<th></th>
<th>Primary Income</th>
<th>Disposable Income</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1979</td>
<td>2000</td>
</tr>
<tr>
<td>Farmers</td>
<td>9 367</td>
<td>17 858</td>
</tr>
<tr>
<td>Artisans</td>
<td>17 090</td>
<td>19 833</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>28 513</td>
<td>30 642</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>19 048</td>
<td>20 185</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>14 009</td>
<td>13 327</td>
</tr>
<tr>
<td>Manual workers</td>
<td>12 264</td>
<td>13 113</td>
</tr>
<tr>
<td>Mean Income</td>
<td>16 503</td>
<td>18 313</td>
</tr>
</tbody>
</table>


Decomposition This fall in the dispersion of mean incomes conditional on social origin may be related to fairly distinct phenomena. First it may originate from an increase in social mobility: if mobility increases, conditional mean incomes converge, as the distributions of social class destinations come closer together. Second, it may come from a reduction in the dispersion of mean incomes conditional on social origin and social destination. We refer to the first effect as the mobility effect, and to the second as the return effect.

The return effect can itself be decomposed into two effects, an absolute and a relative one. The absolute effect corresponds to a reduction in the spread of mean incomes conditional on...
social destination. The relative effect is detected by computing the ratio of the mean income conditional on social origin and social destination over the mean income conditional on social destination. Not all occupations classified under the same destination yield the same income. Coming from a high social origin may help occupy better paid jobs within each social destination. This phenomenon may lead to differences in mean income conditional on social origin and destination, once destination has been controlled. If these differences decrease over time, a relative return effect has been detected.

Formally, the mean income in year $t$, conditional on social origin $j$ can be expressed as:

$$
\bar{y}_j^t = \sum_k \alpha_{jk}^t r_{jk}^t \bar{y}_k^t
$$

where $k$ denotes social destination, $\alpha_{jk}^t$ is the probability of destination $k$ conditional on origin $j$, $r_{jk}^t = \frac{y_{jk}^t}{\bar{y}_k^t}$ where $\bar{y}_{jk}^t$ denotes the mean conditional on social origin and destination and $\bar{y}_k^t$ the mean conditional on social destination. A change in $\alpha$ reflects a mobility effect, a change in $\bar{y}_k^t$, an absolute return effect, and a change in $r_{jk}^t$ a relative return effect. We begin with a description of the direction of these effects over the period we study before moving to a decomposition analysis à la Oaxaca-Blinder.

The evolution of social mobility is summarized by the mobility matrices given in Table 4. The matrices indicate a rise in social mobility for several social groups. For example, children of higher-grade professionals see their probability of becoming higher-grade professionals fall from 51% in 1979 to 44% in 2000, while in the meantime the proportion of higher-grade professionals in the total population has increased. Children of farmers experience a large increase in upward mobility: while their probability of becoming farmer remains unchanged, the probability that they become higher-grade or lower-grade professionals increases between 1979 and 2000 and the probability that they become manual or non-manual workers decreases. Again, the mobility effect accounts for part of the change in mean incomes conditional on social origin.

As documented in Table 3, mean incomes conditional on social destination tend to partly converge between 1979 and 2000. Higher-grade professionals experience slower income growth than other groups and non-wage earners experience faster growth. This indicates that the return effect narrows the dispersion of mean incomes conditional on social origin. However, it only accounts for part of the evolution. The observed reduction in mean income gaps is much stronger when conditioning on social origin than when conditioning on social destination. For instance the ratio between the mean disposable income of higher-grade professionals and manual workers only falls from 2.04 to 1.92. When conditioning on social origin instead of destination,
Table 4: Occupational group transition matrices

<table>
<thead>
<tr>
<th></th>
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<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>A- 1979</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Farmers</td>
<td>.225</td>
<td>.074</td>
<td>.047</td>
<td>.116</td>
<td>.14</td>
<td>.396</td>
<td>.216</td>
</tr>
<tr>
<td>Artisans</td>
<td>.009</td>
<td>.234</td>
<td>.192</td>
<td>.219</td>
<td>.147</td>
<td>.195</td>
<td>.139</td>
</tr>
<tr>
<td>H-grade prof.</td>
<td>.009</td>
<td>.033</td>
<td>.51</td>
<td>.261</td>
<td>.133</td>
<td>.05</td>
<td>.069</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>.000</td>
<td>.045</td>
<td>.287</td>
<td>.401</td>
<td>.138</td>
<td>.127</td>
<td>.073</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>.006</td>
<td>.057</td>
<td>.144</td>
<td>.295</td>
<td>.197</td>
<td>.298</td>
<td>.111</td>
</tr>
<tr>
<td>Manual workers</td>
<td>.005</td>
<td>.083</td>
<td>.064</td>
<td>.209</td>
<td>.164</td>
<td>.474</td>
<td>.389</td>
</tr>
<tr>
<td>Total</td>
<td>.053</td>
<td>.093</td>
<td>.134</td>
<td>.218</td>
<td>.156</td>
<td>.343</td>
<td>1.00</td>
</tr>
</tbody>
</table>

| B- 2000 |         |          |               |               |                 |                |       |
| Farmers | .218    | .049     | .100          | .159          | .146            | .325           | .105  |
| Artisans | .004    | .118     | .205          | .251          | .191            | .229           | .122  |
| H-grade prof. | .005    | .044     | .415          | .318          | .119            | .097           | .150  |
| L-grade prof. | .001    | .055     | .222          | .349          | .175            | .186           | .085  |
| Non-man. workers | .007    | .056     | .137          | .275          | .232            | .290           | .184  |
| Manual workers | .005    | .047     | .066          | .192          | .198            | .490           | .351  |
| Total | .028    | .058     | .166          | .243          | .184            | .318           | 1.00  |

Notes: the table gives the distribution of social destination (in column) conditional on social origin. The rows and columns labelled ‘Total’ give the column (resp. row) marginal distribution. Example: in 2000 sample, 10.5% of the population are children of farmers, 2.8% are farmers, and 4.9% of the children of farmers are artisans. H-grade prof.: higher-grade professionals. L-grade prof.: lower-grade professionals. Non-man. workers: non-manual workers.
Table 5: Relative return effect

<table>
<thead>
<tr>
<th></th>
<th>A- 1979</th>
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<tbody>
<tr>
<td>Farmers</td>
<td>.95</td>
<td>.94</td>
<td>1.31</td>
<td>1.39</td>
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<tr>
<td>Artisans</td>
<td>.82</td>
<td>1.08</td>
<td>1.09</td>
<td>.95</td>
<td>1.03</td>
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<tr>
<td>H-grade prof.</td>
<td>.93</td>
<td>1.05</td>
<td>1.15</td>
<td>1.08</td>
<td>.96</td>
</tr>
<tr>
<td>L-grade prof.</td>
<td>.96</td>
<td>.98</td>
<td>1.35</td>
<td>1.10</td>
<td>1.06</td>
</tr>
<tr>
<td>Non-man. workers</td>
<td>.95</td>
<td>1.13</td>
<td>1.33</td>
<td>1.08</td>
<td>1.06</td>
</tr>
<tr>
<td>Manual workers</td>
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<td>1.33</td>
<td>1.33</td>
<td>1.08</td>
<td>1.06</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
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<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Farmers</td>
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<td>.94</td>
<td>1.03</td>
<td>.96</td>
</tr>
<tr>
<td>Artisans</td>
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<td>1.06</td>
<td>1.05</td>
<td>.97</td>
<td>1.07</td>
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<td>1.05</td>
<td>1.04</td>
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<td>1.05</td>
<td>.97</td>
<td>1.04</td>
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<td>.93</td>
<td>.94</td>
<td>.99</td>
<td>1.05</td>
<td>.97</td>
</tr>
</tbody>
</table>

Notes: the table gives the mean income conditional on social origin and social destination divided by the mean income of destination.

To detect a relative return effect, we compute the ratio of the mean income conditional on social origin and destination to the mean income conditional on social destination.\(^{36}\) Table 5 shows that for almost every social destination, coming from a farmer or blue-collar background represents a handicap in 1979. It is no more true for descendants of farmers in 2000. On the opposite, originating from a high or low grade professional background represents a significant advantage which has decreased sharply at least for the former category over the period. On the whole, the relative return effect may be registered as a source of decrease in EOP.

The contribution of the mobility, the absolute and the relative return effects to the observed evolution can be identified using the Oaxaca-Blinder (1973) decomposition applied to equation 5. As is well-known, the decomposition is not unique. To assess the contribution of changes in \(\alpha\), \(r\) or \(y\), there are several possible choices of a reference period for the factors held constant. Taking the arithmetic average on all six possible decompositions, the change in mean income conditional on social origin between \(t\) and \(t'\) can written as:

\[
\Delta y_j = \frac{1}{6} (\Delta \alpha_j A + \Delta r_j B + \Delta \bar{y}_j C)
\]

\(^{36}\)Farmers have been excluded from social destinations because of problems of statistical inference.
Table 6: Decomposition of the evolution of the mean income

<table>
<thead>
<tr>
<th></th>
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<th></th>
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<th></th>
<th></th>
</tr>
</thead>
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<td>4481</td>
<td>1 553</td>
<td>-1 457</td>
<td>-521</td>
<td>863</td>
<td>1 118</td>
</tr>
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</table>

<table>
<thead>
<tr>
<th>Mobility effect</th>
<th>Change in mean income (1979-2000)</th>
</tr>
</thead>
<tbody>
<tr>
<td>704</td>
<td>1 153</td>
</tr>
<tr>
<td>(%)</td>
<td>-28</td>
</tr>
<tr>
<td>Relative return effect</td>
<td>Change in mean income (1979-2000)</td>
</tr>
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<td>968</td>
<td>21</td>
</tr>
<tr>
<td>(%)</td>
<td>-95</td>
</tr>
<tr>
<td>Absolute return effect</td>
<td>Change in mean income (1979-2000)</td>
</tr>
<tr>
<td>2808</td>
<td>62</td>
</tr>
<tr>
<td>(%)</td>
<td>1677</td>
</tr>
</tbody>
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with

\[
A = 2\bar{r}_j \hat{y}_k + \bar{r}_j \hat{y}_k + \bar{r}_j \hat{y}_k + 2\bar{r}_j \hat{y}_k;
\]

\[
B = 2\alpha_j \hat{y}_k + \alpha_j \hat{y}_k + \alpha_j \hat{y}_k + 2\alpha_j \hat{y}_k;
\]

\[
C = 2\alpha_j \bar{r}_j + \alpha_j \bar{r}_j + \alpha_j \bar{r}_j + 2\alpha_j \bar{r}_j
\]

where \(\alpha_j\), \(\bar{r}_j\) and \(\hat{y}_k\) denote the vectors \((\alpha_{j1} \cdots \alpha_{jK})\), \((r_{j1} \cdots r_{jK})\) and \((\hat{y}_{1} \cdots \hat{y}_{K})\). The terms A, B and C arise from the averaging procedure. The first term on the right hand-side of equation 6 is the mobility effect, the second is the relative return effect and the third one the absolute return effect.

The results of the decompositions for the mean disposal incomes are given in Table 6. The mobility and relative return effects contribute negatively to the change in the mean conditional income for every category except for the children of farmers. On the contrary and as could have been expected every social origin displays a positive absolute return effect, while its importance is broadly comparable across social groups, the absolute return effect is much more important for the children of farmers.

This pattern of results shows that mobility between categories as measured by the mobility effect and mobility within categories, measured by the relative return effect, are positively correlated. Adding these two effects in a broad mobility concept, the following ranking of social groups appears: children of farmers experience a large positive increase in their mean income due to mobility. Next, the children of artisans display a slight decrease due to the same effect. Children of non-manual workers witness a larger negative drop in their mean income and then
finally the children of lower-grade professionals and higher-grade professionals are much more negatively hurt by mobility within and between groups.

5 Conclusion

While different ethical positions can be defended, regarding how to substantively define equality of opportunity, our analysis indicates that any definition relies on a partition of the determinants of individual outcomes into three distinct groups: effort, which includes the determinants that are seen as a legitimate source of outcome differences; circumstances, which consist of the determinants that should not lead, other things equal, to differences in outcome; luck, which comprises the determinants that are seen as a fair source of inequality provided that they are even-handed, with respect to circumstances. In this perspective, the generic model of equality of opportunity developed in this paper appears as a general model, that encompasses a variety of specific conceptions of equality of opportunity, depending on the precise empirical characterization of the above three sets of factors. This model also makes clear that, once these different sets have been delineated, there may still be several ways to define equality of opportunity, which correspond to the strong and weak criteria introduced here. These two criteria are expressed in terms of stochastic dominance because luck is akin to a form of risk in the realization of income that cannot be ignored.

Whatever the precise conception adopted, empirically assessing equality of opportunity turns out to be a data-demanding exercise. To many, equality of opportunity appears as a more desirable social objective than equality of outcome because it takes into account the determinants of observed outcomes. The obvious drawback of this conception is that making equality of opportunity judgments ideally requires that all the relevant determinants of outcome be observable. Of course, this requirement will rarely be met and assessing equality of opportunity will most likely take place under imperfect information on the relevant determinants. In this paper, we exhibit two testable conditions of equality of opportunity that can be used in this context. Because these two conditions are expressed in terms of stochastic dominance, our general model, incorporating luck, shapes our empirical strategy in a very distinctive manner, compared to the rest of the literature.

The first criterion gives a necessary condition for the strong form of equality of opportunity. The other is a sufficient condition for the weak form. Only in very restrictive cases is it possible to exhibit a necessary and sufficient condition. Hence, in almost all cases, imperfect information implies that equality of opportunity cannot be fully assessed. This is probably the price to pay...
if we are willing to develop a rich enough view. It is also important to emphasize that these two conditions still allow, in our empirical application, to provide a valuable assessment of equality of opportunity in France. Even with limited information on the determinants of outcomes, these conditions indicate that equality of opportunity is clearly violated. This conclusion would therefore remain unchanged if we had access to a richer data set.

On the empirical side, this paper reveals that social inheritance is a deeply rooted source of inequality in France over the period 1979-2000. Differences in social origin translate into significant gaps of living conditions. Equality of opportunity in income acquisition does not prevail, neither for primary income nor for disposable income. However, the degree of inequality of opportunity tends to decrease. During this period, the average gap between the most advantaged social group, the children of higher-grade professionals, and the least advantaged one fall by one half.\footnote{This conclusion confirms those of Vallet (2004) who notices a slight increase in social mobility in France.} The explanation of this evolution is addressed in a companion paper (Lefranc et al. (2006)). We show that this reduction in inequality of opportunity does not arise from a decrease in the degree of transmission of economic advantage from one generation to the next. Over time, children tend to face more equal opportunities because of a fall in inequality of outcomes among their parents.

Lastly, this paper has underlined an important phenomenon: the risk of social lotteries appears very similar across the different groups of social origin. As a first approximation, the influence of social origin on opportunities for income, in France, can be summarized by a scale factor: individual income is determined by the product of a random variable - distributed independently of social origin- and the mean income conditional on social origin. Whether a similar determination of income opportunities is also at work in other countries is a question that would be worth investigating. Future research should also analyze the theoretical explanation and consequences of this important stylized fact.

References


Appendix
A- Statistical tests

Stochastic dominance relationships can be easily expressed and statistically tested by making use of poverty indices. Formally the poverty index at the order $g$, for a poverty line $z$, is defined by:

$$D^g(z) = \int_0^z (z - x)^{g-1}dF(x)$$

For example $D^1(z)$ is the proportion of people whose income is below $z$; $D^2(z)$ measures for the mean poverty gap, i.e. the mean amount that should be given to people below the poverty line $z$, to reach this threshold.

The link between poverty indices and stochastic dominance has been established by Foster and Shorrocks (1988). They have shown that stochastic dominance at the order $g$ expresses in vectorial notations. Let $\delta$ denote the vector of poverty indices $(\hat{D}^g(z_1), \ldots, \hat{D}^g(z_k))$ and $\Sigma$ its asymptotic variance-covariance matrix.\(^{38}\)

The hypothesis of dominance at the order $g$ between two distributions $F_A$ and $F_B$ can be expressed in vectorial notations. Let $\delta = (D_B^g - D_A^g)$ the difference of the vectors of poverty indices. Stochastic dominance at the order $g$ can be test as : $H_0 : \delta \in \mathbb{R}^k_+$ versus $H_1 : \delta \notin \mathbb{R}^k_+$. The test statistic is constructed from the estimated vector $\hat{\delta} = (\hat{D}_B^g - \hat{D}_A^g)$, whose asymptotic variance-covariance matrix is given by $\sum_{AB} \sum_{BA}$ under the hypothesis of independence between distributions $A$ and $B$.

The null hypothesis is defined by a set of $k$ constraints. Two approaches can be followed to conduct this test. The first consist in testing each of the $k$ constraints separately. The intersection of $k$ sub-hypothesis is tested in each point where the distributions are compared. This kind of test, called "intersection-union", is used for example in Bishop et al. (1992). However, Dardanoni et Forcina (1999) and Davidson and Duclos (2000) have demonstrated that this test procedure has relatively low power, since it ignores the covariance structure of the vector $\delta$. The second approach, that will be followed here, amounts to simultaneously test the $k$ constraints, using a Wald test and explicitly taking into account the covariance structure of the estimated poverty indices differences vector. The general principle of the test amounts to compare the distances between the estimated vector $\hat{\delta}$ to the sets defining respectively the null

\(^{38}\)In order to use non-stochastic poverty lines, these percentiles are computed from administrative tax records (the Revenus Fiscaux data) that are independent of the BDF survey.

\(^{39}\)In our application this formula has been adapted to account for sample weights.

\(^{40}\)See Davidson and Duclos (2000), theorem 1. p.1441 for the expression of this matrix.
hypothesis and the alternative hypothesis. Of course, Wald tests can also be used to test the equality of two distributions.

**Equality tests** The Wald test for the equality of two distributions is relatively easy to implement and resort to a $\chi^2$ test. The null hypothesis is given by $H_0: \delta = 0$. One can show (Beach & Davidson 1983, Davidson & Duclos 2000) that under the null hypothesis, the vector $\hat{\delta}$ is asymptotically normal and we have:

$$\hat{\delta} \sim N(0, \frac{\Sigma_A}{N_A} + \frac{\Sigma_B}{N_B})$$

Hence the test statistic $T_1$ has the following asymptotic distribution under the null hypothesis:

$$T_1 = \hat{\delta}' (\frac{\Sigma_A}{N_B} + \frac{\Sigma_B}{N_B})^{-1} \hat{\delta} \sim \chi^2_k$$

**Stochastic dominance tests** Stochastic dominance tests are more complex to implement since in this case, the set corresponding to the null hypothesis is defined by an inequality constraint. The hypothesis are given by: $H_0: \delta \in \mathbb{R}^k_+$ against $H_1: \delta \notin \mathbb{R}^k_+$. The Wald test statistic with such constraints has been developed by Kodde and Palm (1986) and Wolak (1989). For this set of hypotheses, the test statistic $T_2$ is defined by:

$$T_2 = \min_{\delta \in \mathbb{R}^k_+} ||\hat{\delta} - \delta||$$

with $||x|| = x'\Sigma^{-1}x$. Kodde and Palm (1986) have shown that the statistic $T_2$ is distributed as a mixture of $\chi^2$ distributions:

$$T_2 \sim \chi^2_k = \Sigma_{j=0}^k w(k, k - j, \Sigma) Pr(\chi^2_j \geq c)$$

with $w(k, k - j, \Sigma)$ the probability that $k - j$ elements of $\hat{\delta}$ are strictly positive. The distribution of the $\chi^2_k$ distribution have not been tabulated, but lower and upper bounds of critical values are available. When these bounds do not allow to reach a conclusion we estimate the critical values of the statistic $T_2$ using Monte-Carlo simulation.\(^{41}\)

---

\(^{41}\)We draw 10,000 multivariate normal vectors with mean 0 and covariance matrix $\Sigma$, and compute the proportion of vectors with $j$ positive elements (for $j \in (0, k)$). This proportion is an estimate of the weight $w(k, j, \Sigma)$.
## B- Summary statistics and results for intermediate waves

Table A-1: Sample summary statistics

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Table A-2: Stochastic dominance tests 1984-1994

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Notes: The occupational group refers to social origin. H-grade prof.: higher-grade professionals. L-grade prof.: lower-grade professionals. Non-man. workers: non-manual workers. Each element in the table indicates the result of the comparison of the income distribution of the groups in row and column using the tests presented in section 4.2:

$>_{i}$: the row dominates the column for order $i$ stochastic dominance;

$<_{i}$: the column dominates the row for order $i$ stochastic dominance;

$:=$: the distributions are equal;

$:?$: the distributions cannot be ranked using first and second order stochastic dominance.
Table A-3: Lorenz dominance tests 1984-1994

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Notes: The occupational group refers to social origin. H-grade prof.: higher-grade professionals. L-grade prof.: lower-grade professionals. Non-man. workers: non-manual workers. Each element in the table indicates the result of the comparison of the Lorenz curves of the groups in row and column using the criterion of Lorenz dominance: 
>:_ the row dominates the column for Lorenz dominance;
<_: the column dominates the row for Lorenz dominance;
=: the Lorenz curves are equal;
?: the Lorenz curves cannot be ranked using Lorenz dominance.