

Statistical Theories of Income and Wealth Distribution

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Abstract The distributions of income and wealth in countries across the world are found to possess some robust and stable features independent of the specific economic, social and political conditions of the countries. We discuss a few physics-inspired multi-agent dynamic models along with their microeconomic counterparts, that can produce the statistical features of the distributions observed in reality. A number of exact analytical methods and solutions are also provided.

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1 Introduction

Even from everyday experience, one can understand that almost without any exception income and wealth in a society are unequally distributed among its people and from time immemorial, this inequality has been a constant source of irritation in all societies. There are several non-trivial issues and questions related to this observation. In fact, the issue of inequality in terms of income and wealth has been perhaps the most fiercely debated one in economics. Economists and philosophers have spent much time on the normative aspects of this problem (Sen, 1999; Foucault, 2003; Scruton, 1985; Rawls, 1971). The direct and indirect effects of inequality on the society have also been studied extensively. In particular, the effects of inequality on the growth of the economy (Aghion *et al*, 1999; Barrow, 1999; Benabou, 1994; Forbes, 2000) or on the political-economic scenario (Alesina and Rodrik, 1992; Benabou, 2000; Alesina and Perotti, 1993; Blau and Blau, 1982) have attracted major attention. Relatively less emphasis had been put on the sources of the problem itself. But there remain very important questions that beg to be answered. How are income and wealth distributed? What are the forms of the distributions? Are they universal or do they depend upon the specific conditions in the individual country? And the most important question is, if inequality is universal, as some of its gross features obviously are, then what is the reason for such universality? More than a hundred years back, this problem caught attention of Pareto and he found that wealth distribution follows a power law decay for the richer section of the society (Pareto, 1897). Much later Champernowne also considered this problem systematically and he came up with a probabilistic theory to justify Pareto's claim (Champernowne, 1953). Separately, Gibrat worked on the same problem and he proposed a law of proportionate development (Gibrat, 1931). It was subsequently found in numerous studies that the distributions of income and wealth indeed possess some globally stable and robust features (see e.g., Yakovenko and Rosser, 2009 for a review). In general, the bulk of the distribution of both income and wealth seems to fit both the log-normal and the gamma distributions reasonably well. Economists usually prefer the log-normal distribution (Montroll and Shlesinger, 1982;

Gini, 1921) whereas statisticians (Hogg *et al*, 2007) and more recently physicists (Yakovenko and Rosser, 2009; Chatterjee *et al*, 2005; Chatterjee and Chakrabarti, 2007), tend to rely more on the gamma distribution. And the upper end of the distribution, that is the tail of the distribution, is agreed to be described well by a power law as was found by Pareto. Although the exact nature of such distributions are yet to be finalized, there is a general agreement on the observation that income and wealth distributions show regularities independent of the country-specific conditions and these observed regularities in patterns may be indicative of a *natural law* of economics.

Here, we survey some multi-agent dynamic models inspired by the physics of energy distribution in many-body thermodynamic systems. Specifically, we intend to discuss a very simple microeconomic model with a large number of agents and consider the asset transfer equations among the agents due to trading in such an economy. It will be shown that this type of asset transfer among the agents in an economy closely resembles the process of energy transfer due to collisions among particles in a thermodynamic system like an ideal gas. The steady state distribution for such a system is an exponential one, as was found by Gibbs a hundred years back (see e.g., Yakovenko and Rosser, 2009). We then see that several modified versions of the same model produce gamma function like behavior for the distribution of money among the agents in the economy. A further modification of the model produces a power law for the upper or tail end of the distribution of money, as has been found empirically. Next, we discuss the analytical aspects of the models and provide some exact results and derivations of the same. So far this is the only known class of models which, starting from microeconomics of utility maximization and solving for the resultant dynamical equations in the line of rigorously established statistical physics, can reproduce quite reliably the major features of both of the income and wealth distributions in economies.

This paper is organized as follows. In section 2, we review the data gathered on income and wealth distributions. In section 3, we consider a simple microeconomic framework as our basic model. In the next section, we discuss a number of different modifications of the model focusing on different economic behavioral assumptions that lead to a number of

intriguing results. In section 5, we review some analytical results of the models considered.

2 A Short Review of Data

The distributions of income and wealth have long been subject to detailed empirical analysis and tests. To put the result briefly, these studies (Yakovenko and Rosser, 2009; Chatterjee *et al*, 2005; Chakrabarti *et al*, 2006) indicate that

$$P(m) \sim \begin{cases} m^\alpha \exp(-m/T) & \text{for } m < m_c, \\ m^{-(1+\nu)} & \text{for } m \geq m_c, \end{cases} \quad (1)$$

where P denotes the number density of people with income or wealth m and α , ν denote exponents and T denotes a scaling factor. The power law in income and wealth distribution (for $m \geq m_c$) is named after Pareto and the exponent ν is called the Pareto exponent. The crossover point (m_c) is extracted from the crossover of the Gamma (or log-normal) distribution to the power law tail. The existence of both features in the same distribution was possibly first demonstrated by Montroll and Shlesinger (1982) who observed that while the top 2-3 % of the population (in terms of income) followed a power law with Pareto exponent $\nu \simeq 1.63$, the rest followed a lognormal distribution. That study led economists to fit the region below m_c to a log-normal form, $\log P(m) \propto -(\log m)^2$. This form has indeed been seen in several studies (see e.g., Souma, 2000; Di Matteo *et al*, 2004; Clementi and Gallegati, 2005). But there are enough empirical evidences that the Gamma distribution form Eqn. (1) fits better with the data, (see e.g., the remarkable fit with the Gibbs distribution in Yakovenko and Silva, 2005; Drăgulescu and Yakovenko, 2001; Drăgulescu and Yakovenko, 2001a; see also Purica, 2004). There are many studies concluding that the tail is described well by a power law (see e.g., Souma, 2000; Drăgulescu and Yakovenko, 2001; Drăgulescu and Yakovenko, 2001a; Aoyama *et al*, 2000). Interestingly, the tail of the distribution of income of companies also follows a power law (see e.g., Okuyama *et al*, 1999; Axtell, 2001). It may, however, be noted that the above-mentioned inferences are drawn

primarily from the income tax data which does not encompass the bottom of the distributions and has some other problems like those induced by personal biases of the tax-payers etc. Hence, a combination of household survey data and upper income range data from income tax sources (see e.g., Bach et al, 2009) would encompass the whole spectrum of the population and would provide a more reliable picture of the entire income distribution (see also Atkinson and Brandolini, 2009; Atkinson et al, 2009).

While there is no dearth of empirical analysis on the income distribution, relatively few studies have considered the distribution of wealth due to the lack of an easily available data source. However, Drăgulescu and Yakovenko (2001a), Levy and Solomon (1997), Coelho *et al* (2005), Sinha (2006) have studied wealth distributions extensively. Hegyi *et al* (2007) studied the wealth distribution in Hungarian medieval society. Similar studies are done on the wealth distribution of ancient Egyptian societies (14-th century BC) (Abul-Magd, 2002) as well. The general feature observed in these limited empirical studies of wealth distribution is that of a power law behavior for the wealthiest 5 – 10% of the population, and gamma or log-normal distribution for the rest of the population.

To sum up, numerous investigations during the last ten years revealed that the tail of the income distribution indeed follows a power law with the value of the Pareto exponent ν generally varying between 1 and 3 (Di Matteo *et al*, 2004; Clementi and Gallegati, 2005; Drăgulescu and Yakovenko, 2001a; Levy and Solomon, 1997; Sinha, 2006; Oliveira *et al*, 1999; Aoyama *et al*, 2003, Clementi and Gallegati, 2005a). The rest of the low income population, follow a different distribution which is debated to be either gamma (Drăgulescu and Yakovenko, 2001; Levy and Solomon, 1997; Aoyama *et al*, 2003; Chakrabarti and Marjit, 1995; Ispolatov *et al*, 1998; Drăgulescu and Yakovenko, 2000) or log-normal (Di Matteo *et al*, 2004; Clementi and Gallegati, 2005; Clementi and Gallegati, 2005a).

The striking similarities observed in the income distributions for different countries indicate that probably the same process governs the distributions of assets in different economies though these economies are superficially different. There is a huge literature on modelling the economies in analogy with large systems of interacting particles, by physicists. From that perspective, the economy is often viewed as a thermodynamic system

in which the distribution of income among the agents is readily identified with the distribution of energy among the particles in a gas. In particular, a class of kinetic exchange models have provided a simple mechanism for understanding the unequal distribution of assets. These models have been successful to capture the key factors in economic interactions that result in different economies with different socio-political structures converging to similar forms of unequal distribution of resources (see Chatterjee *et al*, 2005 and Chakrabarti *et al*, 2006, which consists of a collection of large number of technical papers in this field).

3 The Model

We intend to discuss a minimal model to analyze the effects of stochastic trading processes on the asset holding in the steady state of an economy. Chakrabarti and Chakrabarti (2009) considered an N -agent exchange economy. Each of the agents produces a single perishable commodity which is different from all other commodities produced. Money is treated as a non-perishable commodity which facilitates transactions. All commodities along with money can enter the utility function of any agent as arguments. These agents care for their future consumptions and hence they care about their savings in the current period as well. Initially, all of these agents are endowed with an equal amount of money which is assumed to be unity. With successive tradings their money-holding will change with time. As will be shown, the steady state distribution of money among the agents is independent of the initial amount endowment. At each time step, two agents are chosen at random to carry out transactions among themselves following the utility maximization principle. The utility functions are of Cobb-Dauglas type. We also assume that the preference structure of the agents are time-dependent that is the parameters of the utility function vary over time (Lux, 2005; Silver *et al*, 2002). Below, we consider a typical transaction that leads to the dynamics of money among the agents.

Suppose agent 1 produces Q_1 amount of commodity 1 only and agent 2 produces Q_2 amount of commodity 2 only and the amounts of money in their possession at time t are $m_1(t)$ and $m_2(t)$ respectively (clearly, $m_i(0) = 1$ for $i=1,2$). Since neither of the two agents possess the commodity

produced by the other agent, both of them will be willing to trade with each other and buy the other good by selling a fraction of their own productions as well as with the money that they hold. Hence, at each time step there would be a net transfer of money from one agent to the other due to trade. Our focus is on how the amounts money held by the agents change over time due to the repetition of such a trading process. For notational convenience, we denote $m_i(t + 1)$ as m_i and $m_i(t)$ as M_i (for $i = 1, 2$).

Utility functions are defined as follows. For agent 1, $U_1(x_1, x_2, m_1) = x_1^{\alpha_1} x_2^{\alpha_2} m_1^{\alpha_m}$ and for agent 2, $U_2(y_1, y_2, m_2) = y_1^{\alpha_1} y_2^{\alpha_2} m_2^{\alpha_m}$ where the arguments in both of the utility functions are consumption of the first (i.e., x_1 and y_1) and second good (i.e., x_2 and y_2) and amount of money in their possession respectively. For simplicity, we assume that the sum of the powers is normalized to 1 i.e., $\alpha_1 + \alpha_2 + \alpha_m = 1$. Let the commodity prices to be determined in the market be denoted by p_1 and p_2 . Now, we can define the budget constraints as follows. For agent 1 the budget constraint is $p_1 x_1 + p_2 x_2 + m_1 \leq M_1 + p_1 Q_1$ and similarly, for agent 2 the constraint is $p_1 y_1 + p_2 y_2 + m_2 \leq M_2 + p_2 Q_2$. In this set-up, we get the market clearing price vector (\hat{p}_1, \hat{p}_2) as $\hat{p}_i = (\alpha_i / \alpha_m)(M_1 + M_2) / Q_i$ for $i = 1, 2$ (see Chakrabarti and Chakrabarti, 2009).

By substituting the demand functions of x_i , y_i and p_i for $i = 1, 2$ in the money demand functions, we get the most important equation of money exchange in this model. We make a restrictive assumption that α_1 in the utility function can vary randomly over time with α_m remaining constant. It readily follows that α_2 also varies randomly over time with the restriction that the sum of α_1 and α_2 is a constant $(1 - \alpha_m)$. In the money demand equations derived from the above-mentioned problem, we substitute α_m by λ and $\alpha_1 / (\alpha_1 + \alpha_2)$ by ε to get the money evolution equations as

$$\begin{aligned} m_1(t + 1) &= \lambda m_1(t) + \varepsilon(1 - \lambda)(m_1(t) + m_2(t)) \\ m_2(t + 1) &= \lambda m_2(t) + (1 - \varepsilon)(1 - \lambda)(m_1(t) + m_2(t)) \end{aligned} \tag{2}$$

where $m_i(t) \equiv M_i$ and $m_i(t + 1) \equiv m_i$. For a fixed value of λ , if α_1 is a random variable with uniform distribution over the domain $[0, 1 - \lambda]$, then ε is also uniformly distributed over the domain $[0, 1]$. For the limiting

value of α_m in the utility function (i.e., $\alpha_m \rightarrow 0$ which implies $\lambda \rightarrow 0$), we get the money transfer equation describing the random sharing of money without savings (see Chakrabarti and Chakrabarti, 2009 for derivation and a discussion in details).

A noteworthy feature of this model is that the exchange equations are not sensitive to the level of production that is even if for some reason the level of production alters (due to production shock) the form of the transfer equations will remain the same provided the form of the utility function remains the same. Also, the model captures the possibility of coupling in the evolution of assets (money). This set of equations forms the basis of our subsequent analysis.

4 Stochastic Models

As is shown above, the results of the economic activities (production, trade and consumption) is represented by a pair of asset transfer equations (see Eqn. (2)). What we basically do is to study the steady state behavior of some modifications of this pair of equations. In the following models, one considers a closed economic system where total money M and total number of agents N is fixed. It is assumed that the system is conservative and no migration occurs. Each agent i possesses money $m_i(t)$ at time t . In any trading, a pair of agents i and j exchange their money (Chakrabarti and Marjit, 1995; Ispolatov *et al*, 1998; Drăgulescu and Yakovenko, 2000; Chakraborti and Chakrabarti, 2000; Chakrabarti and Chakrabarti, 2009) such that their total money is (locally) conserved and none end up with negative money ($m_i(t) \geq 0$, i.e., debt not allowed):

$$m_i(t+1) = m_i(t) + \Delta m; \quad m_j(t+1) = m_j(t) - \Delta m \quad (3)$$

following local conservation:

$$m_i(t) + m_j(t) = m_i(t+1) + m_j(t+1); \quad (4)$$

time (t) changes by one unit after each trading.

4.1 Model A: Random Sharing of Money

The simplest model considers random sharing of the total money between the trading partners. Assuming $\lambda \rightarrow 0$ in Eqn. (2), we get

$$\begin{aligned} m_i(t+1) &= \varepsilon[m_i(t) + m_j(t)] \\ m_j(t+1) &= (1 - \varepsilon)[m_i(t) + m_j(t)] \end{aligned} \quad (5)$$

for the i -th and the j -th agent, where ε is a random fraction uniformly distributed between 0 and 1. This is the simplest set of equations of money transfer among the agents. Interestingly, the same set of equations represents transfer of energy among particles due to collisions in an ideal gas except that there all m_i 's (money) in Eqn. (5) are substituted by the colliding particles' energies. Note that money remains conserved in this model. While deriving the probability density function, we must account for all possible divisions of the total amount of money i.e., $m_i(t) + m_j(t)$. Clearly all trading actions must satisfy the condition that

$$P(m_i)P(m_j) = P(\varepsilon[m_i + m_j])P((1 - \varepsilon)[m_i + m_j]) \quad (6)$$

for all ε , $0 \leq \varepsilon \leq 1$. However, if we consider the distinct possibility that the entire amount of money accrues to one individual only and the other becomes pauper, we can solve the model very easily. Using that particular kind of trading, we get

$$P(m_i)P(m_j) = P(m_i + m_j)P(0). \quad (7)$$

Clearly the steady state ($t \rightarrow \infty$) distribution of money is a Gibbs (exponential) distribution:

$$P(m) = P(0) \exp(-m/T); T = M/N. \quad (8)$$

Hence, no matter how uniform or justified the initial distribution is, the eventual steady state correspond to the exponential distribution where most of the people have got very little money. This steady state result is seen to be very robust. Several variations of the mode of trading, and of the 'trading network' (on which the agents can be put at the nodes and

each agent trade with its ‘neighbors’ only), whether compact, fractal or small-world like (Oliveira *et al*, 1999) leaves the distribution unchanged. There are still other studies where variations like random sharing of an amount $2m_2$ only (not of $m_1 + m_2$) when $m_1 > m_2$ (trading at the level of the relatively poorer agent in the trade), lead even to a drastic situation: all the money in the market drifts to one agent and the rest become truly pauper (Hayes, 2002; Chakraborti, 2002).

4.2 Model B: With Constant λ

We now consider Eqn. (2) with constant λ for the i -th and j -th agents:

$$m_i(t+1) = \lambda m_i(t) + \varepsilon(1-\lambda)(m_i(t) + m_j(t))$$

$$m_j(t+1) = \lambda m_j(t) + (1-\varepsilon)(1-\lambda)(m_i(t) + m_j(t)).$$

Note that λ acts as a savings factor in this model, where each trader at time t saves a fraction λ of its money $m_k(t)$ (for $k = i, j$) and trades randomly with the rest (see Chakraborti and Chakrabarti, 2000).

The market (non-interacting at $\lambda = 1$) becomes ‘interacting’ for any non-vanishing $\lambda (< 1)$. For fixed λ (same for all agents), the steady state distribution $P(m)$ of money is exponentially decaying on both sides of the mode of the distribution i.e., the most-probable amount of money per agent. The mode also shifts away from $m = 0$ (for $\lambda = 0$) to M/N as $\lambda \rightarrow 1$ (Fig. 1). This self-organizing feature of the market, induced by sheer self-interest of saving by each agent without any global perspective, is very significant since the fraction of people below a particular poverty line decrease as the saving fraction λ increases and most people end up with some finite, non-zero fraction of the average money in the market (Chakraborti and Chakrabarti, 2000). Although this fixed saving propensity does not give yet the Pareto-like power-law distribution, the Markovian nature of the scattering or trading processes (see Eqn. (7)) is effectively lost. Indirectly through λ , the agents get to know (start interacting with) each other and the system co-operatively self-organises towards a stable form with a non-vanishing most-probable amount of money-holding (see Fig. 1).

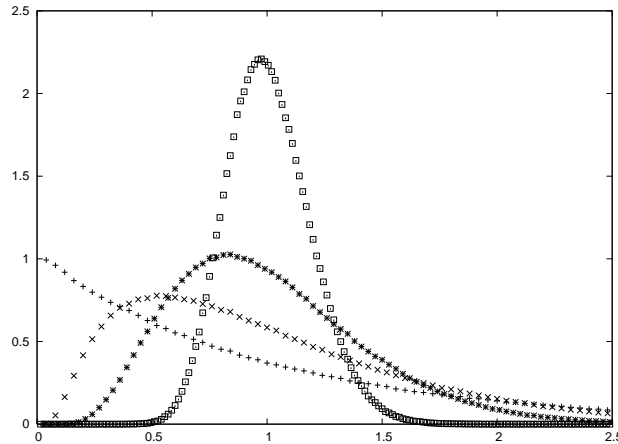


Figure 1:

Steady state money distribution $P(m)$ (y-axis) is plotted against money m (x-axis) for the model with uniform savings. +, x, * and \square denotes distributions with $\lambda = 0, 0.3, 0.6$ and 0.9 respectively. All data sets shown are for average money per agent $M/N = 1$ and $N = 100$.

Angle (Lux, 2005; Angle, 1986; Angle, 2006) proposed an early version of the above model several years back in sociology journals. Angle's 'Inequality Process' is described by the following equations:

$$\begin{aligned}
 m_i(t+1) &= m_i(t) + D_t w m_j(t) - (1 - D_t) w m_i(t) \\
 m_j(t+1) &= m_j(t) + (1 - D_t) w m_i(t) - D_t w m_j(t)
 \end{aligned}
 \tag{9}$$

where w is a fixed fraction and D_t takes value 0 or 1 randomly. The numerical simulation results of Angle's model fit well to Gamma distributions.

In the model with uniform savings, the distribution of the monetary assets shows a self organizing feature. A peaked distribution with a most probable value indicates an economic scale. Empirical findings in homogeneous groups of individuals as in waged income of factory labourers in UK and USA (Willis and Mimkes, 2004) and data from population survey in USA among students of different school and colleges support this observation (Angle, 2006).

4.3 Model C: With Distributed λ

In reality, people face different constraints resulting in different patterns of saving or at a more basic level, their attitudes towards savings may not be the same i.e., the parameters of their utility functions may differ from one person to another. This in turn implies that the saving parameter λ is very heterogeneous. To imitate this situation, we allow λ to be distributed within the population (Chatterjee *et al*, 2003; Chakrabarti and Chatterjee, 2004; Chatterjee *et al*, 2004). The evolution of money in such a trading can be written as:

$$\begin{aligned} m_i(t+1) &= \lambda_i m_i(t) + \varepsilon [(1 - \lambda_i) m_i(t) + (1 - \lambda_j) m_j(t)] \\ m_j(t+1) &= \lambda_j m_j(t) + (1 - \varepsilon) [(1 - \lambda_i) m_i(t) + (1 - \lambda_j) m_j(t)]. \end{aligned} \quad (10)$$

The trading rules are same as before, except that

$$\Delta m = \varepsilon(1 - \lambda_j) m_j(t) - (1 - \lambda_i)(1 - \varepsilon) m_i(t) \quad (11)$$

here; where λ_i and λ_j are the saving propensities of agents i and j . The agents have fixed (over time) saving propensities, distributed independently, randomly and uniformly (white) within an interval 0 to 1: agent i saves a random fraction λ_i ($0 \leq \lambda_i < 1$) and this λ_i value is quenched for each agent (λ_i are independent of trading or t). Studies show that for uniformly distributed saving propensities, $\rho(\lambda) = 1$ for $0 \leq \lambda < 1$, one gets eventually $P(m) \sim m^{(1+\nu)}$, with $\nu = 1$ (see Fig. 2). The eventual deviation from the power law in $Q(m)$ in the inset of Fig. 2 is due to the exponential cutoff contributed by the rare statistics for high m value.

It may be noted in this connection that such a dispersion in the savings propensity λ is often seen in most of the economies with 0.2 as the most probable value (Purica, 2004). However, as will be shown in Sec. 5, the Pareto distribution results in only when the dispersion of λ continues up to unity.

A direct analytical derivation of the pareto law found above, is provided in section 5.3. It is seen that the variation in ε plays no role in it. The key factor is the distribution of the savings propensity λ . Referring to section 3, we can define the utility functions as follows. For agent 1,

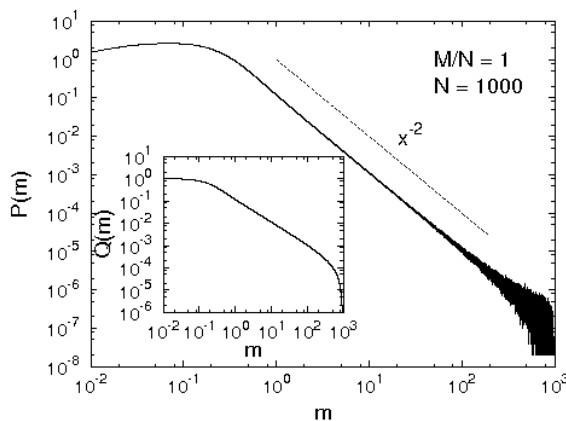


Figure 2:

Steady state money distribution $P(m)$ for the distributed λ model with $0 \leq \lambda < 1$ for a system of $N = 1000$ agents with the average money per agent $M/N = 1$. A power-law is observed with $1 + \nu = 2$.

$U_1(x_1, x_2, m_1) = x_1^{\alpha_1} x_2^{\alpha_2} m_1^{\alpha_m}$ and for agent 2, $U_2(y_1, y_2, m_2) = y_1^{\beta_1} y_2^{\beta_2} m_2^{\beta_m}$. Again for simplicity, we assume that the sums of the powers are normalized to 1 i.e., $\alpha_1 + \alpha_2 + \alpha_m = 1$ and $\beta_1 + \beta_2 + \beta_m = 1$. We make a crucial assumption that $\alpha_m \simeq \beta_m$ and $\alpha_m, \beta_m \rightarrow 1$. Also, we assume that $\alpha_1 = \alpha_2 \simeq \beta_1 = \beta_2$. Then by doing the same exercise as in section 3 and denoting α_m and β_m by λ_1 and λ_2 respectively, we can approximate the money evolution equations in the following form,

$$\begin{aligned} m_1(t+1) &= \lambda_1 m_1(t) + \frac{1}{2} ((1 - \lambda_1) m_1(t) + (1 - \lambda_2) m_2(t)) \\ m_2(t+1) &= \lambda_2 m_2(t) + \frac{1}{2} ((1 - \lambda_1) m_1(t) + (1 - \lambda_2) m_2(t)). \end{aligned} \tag{12}$$

Note that ε is constant here (equals to 1/2) and λ is the variable. The above set of equations also produce the Pareto distribution in the steady state (see sections 5.2 and 5.3 for analytical derivations of the same).

4.4 Model D: Taxation and Redistribution

This model was studied by Drăgulescu and Yakovenko (2001) and Guala (2009). We return to Eqn. (5) which captures the process of random sharing of money. In this model trading process takes place in two steps. In the first step, we assume that prior to trade a fixed fraction τ of money is taxed from both of them. Random sharing occurs with the rest of the money. In the second step, the total amount of money taxed is distributed equally among all the agents in the economy.

$$\begin{aligned} m_i(t + 1/2) &= \varepsilon(1 - \tau)(m_i(t) + m_j(t)) \\ m_j(t + 1/2) &= (1 - \varepsilon)(1 - \tau)(m_i(t) + m_j(t)) \end{aligned} \tag{13}$$

For all k ,

$$m_k(t + 1) = m_k(t + 1/2) + \tau \frac{(m_i(t) + m_j(t))}{N}. \tag{14}$$

This model also gives rise to gamma-like features in the steady state distribution. But it has a peculiarity in that it shows transition from exponential to gamma function as τ goes up and then after a threshold, it returns to an exponential for higher values τ . Guala (2009) shows that the optimal tax rate is about 0.325 where optimality refers to equality among the agents.

4.5 Model E: Risk Aversion and Insurance

Chakrabarti and Chakrabarti (2009) proposed this model with a different interpretation. In this model, the money transfer process takes place in two steps. The process is again governed by Eqn. (5). We assume that the agents are risk-averse. Hence, prior to trade they reach an agreement that whoever will be the winner, shall transfer a fraction f of his *excess* of money to the loser. This is akin to an insurance where the agents sacrifice higher gains to avoid losses. In the first step, the agents trade in an absolutely random fashion. This step follows from Eqn. (2) above if we

consider that $\lambda \rightarrow 0$. Hence,

$$\begin{aligned} m_i(t + 1/2) &= \varepsilon[m_i(t) + m_j(t)] \\ m_j(t + 1/2) &= (1 - \varepsilon)[m_i(t) + m_j(t)]. \end{aligned} \tag{15}$$

The agents agree to split the *excess* amount of money. Hence the agent with more money, transfers a fraction f of the excess to the agent with less money. It is reasonable to assume that $0 \leq f \leq 0.5$. If $m_i(t + 1/2) \geq m_j(t + 1/2)$, excess of money, $\delta = m_i(t + 1/2) - m_j(t + 1/2)$. Hence,

$$\begin{aligned} m_i(t + 1) &= m_i(t + 1/2) - (f\delta) \\ m_j(t + 1) &= m_j(t + 1/2) + (f\delta). \end{aligned} \tag{16}$$

This process is repeated at each time step until the system reaches a steady state and the distribution $p(m)$ of money among the agents in the steady state are studied. Substituting for δ , $m_i(t + 1/2)$ and $m_j(t + 1/2)$ in the above equation, we get the reduced equations

$$\begin{aligned} m_i(t + 1) &= g[m_i(t) + m_j(t)] \\ m_j(t + 1) &= (1 - g)[m_i(t) + m_j(t)]. \end{aligned} \tag{17}$$

The expression of g in the above equations is $g = f + (1 - 2f)\varepsilon$. It may be noted that g is a linear transformation of an uniformly distributed variable ε . Hence, g is also uniformly distributed and its domain is $[f, 1 - f]$. With rising values of f , this model shows a transition from pure exponential (for $f = 0$) to a Δ distribution (for $f = 0.5$) in money holding. Gamma like distributions emerge for values of f between the two extremes.

There are other useful stochastic models which are also able to generate exponential or gamma function like probability density functions for the distribution of money. Studies of these models agree with the general form of the steady state distributions obtained here. However, these models have no theoretical support from economics. One can see for example Scalas *et al*, 2006; Garibaldi *et al*, 2007 and Kar-Gupta, 2006 for a detailed discussion on this type of models.

5 Analytical Studies

There have been a number of attempts to study the uniform savings model (Model B, Sec. 4.2) analytically (see e.g., Das and Yarlalagadda, 2005), but no closed form expression for the steady state distribution $P(m)$ has yet been arrived at. The exact distributions for the model with taxation (Model D) and the model with risk averse agents (Model E) are also unknown whereas the model with distributed savings (Model C) has been solved in several ways (Chatterjee *et al*, 2005a; Chatterjee *et al*, 2005b; Mohanty, 2006; Repetowicz *et al*, 2005; Richmond *et al*, 2005). Below, we discuss a very simple method of obtaining the moments of the steady state distributions of the models B and D up to any order without knowing the actual distribution.

The mathematical structures of the discrete and continuous (both in time and space) versions of the kinetic exchange equations under strict and not-so-strict conservation laws are well-studied. The major findings are the precise analysis of the Boltzmann-type equations resulting from the binary collision models (Markowich, 2007), links between the steady state distributions and a number of particular asymptotes (Markowich, 2007; Toscani and Brugna, 2009), possible extensions to incorporate multiple interacting species, generation of bimodal distribution of income/opinion formation, effects of taxes etc (Doring *et al*, 2008; Toscani, 2009). However, we stick to the descriptive part of the ideal gas like markets only. Below follows some non-rigorous but useful technical results.

5.1 Moments of the Distribution

We denote *expectation* or average of a variable x by $E(x)$ and the central moment of order $n > 1$ (μ_n) of a variable x as

$$E(x - E(x))^n = E \left(\sum_{l=0}^n \binom{n}{l} x^l E(-x)^{(n-l)} \right). \quad (18)$$

For $n = 2$, $E(x - E(x))^n$ corresponds to the variance of x and is denoted by $V(x)$. Since the systems are conservative and the initial endowments were

unity for all agents, it is obvious that $E(m_i)$ would be unity. So we can write the n -th moment of the distribution of money without subscript as

$$E(m-1)^n = E\left(\sum_{l=0}^n \binom{n}{l} m^l (-1)^{(n-l)}\right). \quad (19)$$

We assume that m_i and m_j are independent variables. Using the money-transfer equation in Eqn. (19), one can find out μ_n iteratively for any n (i.e., if the moments up to $(n-1)$ -th order are known then it is possible to find the n -th moment by the above equation). For example, we find out the second moment of the steady state distribution of model B in section 4.2 by assuming that the first moment is set to unity i.e., the average money $M/N = 1$. Note that for the i -th agent, the time evolution of money is

$$m_i(t+1) = \lambda m_i(t) + \varepsilon(1-\lambda)[m_i(t) + m_j(t)].$$

By taking expectations on both sides we get $E(m) = 1$. Also, in the steady state, $V(m_i) = E(x^2) - [E(x)]^2$ where $x = \lambda m_i + \varepsilon(1-\lambda)(m_i + m_j)$. Using the fact that $E(x) = 1$, we get

$$V(m_i) = \lambda^2 E(m_i^2) + (1-\lambda)^2 E(\varepsilon^2 [m_i + m_j]^2) + 2\lambda(1-\lambda)E(\varepsilon)E(m_i[m_i + m_j]) - 1.$$

Using symmetry in the agent indices i and j , one gets $E(\varepsilon^2 [m_i + m_j]^2) = [V(\varepsilon) + 1/4][2V(m) + 4]$. Since $V(\varepsilon) = 1/12$ (as ε is uniformly distributed), we get the following equation after rearranging terms

$$V(m) = \lambda^2 [V(m) + 1] + \frac{2}{3}(1-\lambda)^2 [V(m) + 2] + \lambda(1-\lambda)[V(m) + 2] - 1.$$

Simplifying the above expression we get the result for $\lambda \neq 1$,

$$V(m) = \frac{(1-\lambda)}{(1+2\lambda)}. \quad (20)$$

Chakrabarti and Chakrabarti (2009) also discusses the second moment of the distribution of the model E.

Patriarca *et al* (2004) claimed through heuristic arguments (based on numerical results) that the distribution of model B is a close approximate form of the Gamma distribution

$$P(m) = Cm^\alpha \exp[-m/T] \quad (21)$$

where $T = 1/(\alpha + 1)$ and $C = (\alpha + 1)^{\alpha+1}/\Gamma(\alpha + 1)$, Γ being the Gamma function whose argument α is related to the savings factor λ as:

$$\alpha = \frac{3\lambda}{1 - \lambda} \quad (22)$$

which implies $T = (1 - \lambda)/(1 + 2\lambda)$ and it may be noted that in the case considered here (with $M/N = 1$), T happens to be equal to the variance of the distribution itself. The same value of the parameter is obtained through moment calculation above (Richmond *et al*, 2005; Chakrabarti and Chakrabarti, 2009). Also, when compared with Eqn. (1), $m_c \rightarrow \infty$. The qualitative argument forwarded here Patriarca *et al* (2004) is that, as λ increases, effectively the agents (particles) retain more of its money (energy) in any trading (scattering). This can be taken as implying that with increasing λ , the effective dimensionality increases and temperature of the scattering process changes. This result has also been supported by numerical results in Bhattacharya *et al* (2005). However, Repetowicz *et al* (2005) and Richmond *et al* (2005) analyzed the moments, and found that moments up to the third order agree with those obtained from the form of the Eqn. (21), and discrepancies start from fourth order onwards. Hence, the actual form of the distribution for this model still remains to be found out.

It is seen that the values of the parameters of the distribution derived by the above-mentioned technique (without any distributional assumption), are identical to those found by Patriarca *et al* (2004). Although this technique enables one to derive the exact values of the moments up to any order, it has an obvious drawback that it can not be applied to the models where one does not get any *representative* equation. For example, this technique does not apply to model D (sec. 4.4).

We review now some of the analytical results on the steady state distribution $P(m)$ of money resulting from the equations Eqn. (10) representing the trading and money dynamics (Model C, Sec. 4.3) in the distributed savings case.

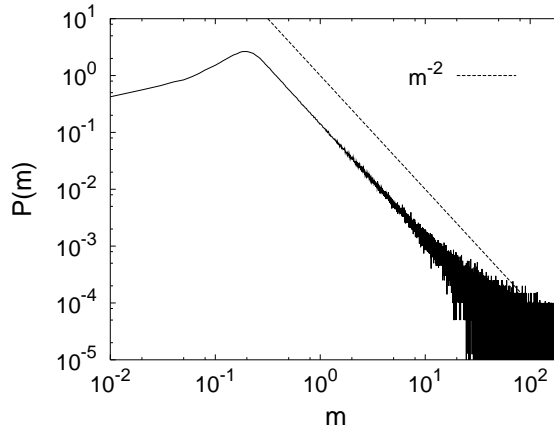


Figure 3:

Steady state money distribution $P(m)$ against m in a numerical simulation of a market with $N = 200$, following equations Eqn. (10) with $\varepsilon = 1/2$. The dotted line corresponds to $m^{-(1+\nu)}$; $\nu = 1$. The average money per agent $M/N = 1$.

5.2 Distribution of Money Difference

In the process defined by Eqn. (10), the total money ($m_i + m_j$) of the pair of agents i and j remains constant, while the difference Δm_{ij} evolves as

$$\begin{aligned}
 (\Delta m_{ij})_{t+1} &\equiv (m_i - m_j)_{t+1} = \left(\frac{\lambda_i + \lambda_j}{2} \right) (\Delta m_{ij})_t \\
 &+ \left(\frac{\lambda_i - \lambda_j}{2} \right) (m_i + m_j)_t \\
 &+ (2\varepsilon - 1)[(1 - \lambda_i)m_i(t) + (1 - \lambda_j)m_j(t)].
 \end{aligned}
 \tag{23}$$

Numerically, as shown in Fig. 2, we observe that the steady state money distribution in the market becomes a power law, following such tradings when the saving factor λ_i of the agents remain constant over time but varies from agent to agent widely. As shown in the numerical simulation results for $P(m)$ in Fig. 3, the law, as well as the exponent, remains unchanged

even when $\varepsilon = 1/2$ for every trading (Chatterjee *et al*, 2004). Clearly, the third term in Eqn. (23) is zero for $\varepsilon = 1/2$. Even in the case where $\varepsilon \rightarrow 1$, the third term in the above equation becomes unimportant for the critical behavior. For simplicity, we concentrate on this case, where the above evolution equation for Δm_{ij} can be written in a more simplified form as

$$(\Delta m_{ij})_{t+1} = \bar{\lambda}_{ij}(\Delta m_{ij})_t + \tilde{\lambda}_{ij}(m_i + m_j)_t, \quad (24)$$

where $\bar{\lambda}_{ij} = \frac{1}{2}(\lambda_i + \lambda_j)$ and $\tilde{\lambda}_{ij} = \frac{1}{2}(\lambda_i - \lambda_j)$. As such, $0 \leq \bar{\lambda} < 1$ and $-\frac{1}{2} < \tilde{\lambda} < \frac{1}{2}$.

The steady state probability distribution D for the modulus $\Delta = |\Delta m|$ of the mutual money difference between any two agents in the market can be obtained from Eqn. (24) in the following way provided Δ is very much larger than the average money per agent $= M/N$. This is because, using Eqn. (24), large Δ can appear at $t + 1$, say, from ‘scattering’ from any situation at t for which the right hand side of Eqn. (24) is large. The possibilities are (at t) m_i large (rare) and m_j not large, where the right hand side of Eqn. (24) becomes $\simeq (\bar{\lambda}_{ij} + \tilde{\lambda}_{ij})(\Delta_{ij})_t$; or m_j large (rare) and m_i not large (making the right hand side of Eqn. (24) becomes $\simeq (\bar{\lambda}_{ij} - \tilde{\lambda}_{ij})(\Delta_{ij})_t$); or when m_i and m_j are both large, which is a much rarer situation than the first two and hence is negligible. Consequently for large Δ , the distribution $D(\Delta)$ satisfies

$$\begin{aligned} D(\Delta) &= \int d\Delta' D(\Delta') \langle \delta(\Delta - (\bar{\lambda} + \tilde{\lambda})\Delta') + \delta(\Delta - (\bar{\lambda} - \tilde{\lambda})\Delta') \rangle \\ &= 2 \left\langle \left(\frac{1}{\lambda} \right) D \left(\frac{\Delta}{\lambda} \right) \right\rangle, \end{aligned} \quad (25)$$

where the δ functions take care of the Δ values permitted by Eqn. (24) and we have used the symmetry of the $\tilde{\lambda}$ distribution and the relation $\bar{\lambda}_{ij} + \tilde{\lambda}_{ij} = \lambda_i$, and have suppressed labels i, j . Here $\langle \dots \rangle$ denote average over λ distribution in the market, and δ denotes the δ -function. Taking now a uniform random distribution of the saving factor λ , $\rho(\lambda) = 1$ for $0 \leq \lambda < 1$, and assuming $D(\Delta) \sim \Delta^{-(1+\nu)}$ for large Δ , we get

$$1 = 2 \int_0^1 d\lambda \lambda^\nu = 2(1 + \nu)^{-1}, \quad (26)$$

giving an unique value of $\nu = 1$. This also indicates that the money distribution $P(m)$ in the market also follows a similar power law variation, $P(m) \sim m^{-(1+\nu)}$ and $\nu = 1$. Distribution of Δ from numerical simulations also agree with this result.

Chatterjee *et al* (2005) and Chatterjee *et al* (2005a) analysed the master equation for the kinetic exchange process and found its solution for a special case. For a pioneering study of the kinetic equations for the two-body scattering process and a more general solution, see Repetowicz *et al* (2005) and Richmond *et al* (2005).

5.3 Average Money at any Saving Propensity and the Distribution

Patriarca *et al* (2005) and Patriarca *et al* (2006) studied the relationship between a particular saving factor λ and the average money held by an agent characterized by that savings factor and these numerical studies revealed that the product of the average money and the unsaved fraction remains constant i.e.,

$$\langle m(\lambda) \rangle (1 - \lambda) = C \quad (27)$$

where C is a constant; here $\langle x \rangle$ denotes ensemble average over x for a particular value of λ . Mohanty (2006) justifies this result rigorously using a mean-field type approach. It is assumed that the distribution of money of a single agent over time is stationary, which means that the time averaged value of money of any agent remains unchanged independent of the initial value of money. Assuming that the i -th agent interacts with all agents over time and taking the expectation of Eqn. (10), one can write

$$\langle m_i \rangle = \lambda_i \langle m_i \rangle + \langle \varepsilon \rangle \left[(1 - \lambda_i) \langle m_i \rangle + \left\langle \frac{1}{N} \sum_{j=1}^N (1 - \lambda_j) m_j \right\rangle \right]. \quad (28)$$

The last term on the right can be replaced by the average over the agents (denoted by a constant C) and since ε is assumed to be distributed randomly and uniformly in $[0, 1]$, so that $\langle \varepsilon \rangle = 1/2$, Eqn. (28) reduces to

$$(1 - \lambda_i) \langle m_i \rangle = C.$$

Since the right side is free of any agent index, it suggests that this relation is true for any arbitrary agent, i.e., $\langle m_i \rangle (1 - \lambda_i) = \text{constant}$, where λ_i is the saving factor of the i th agent (as in Eqn. (27)) and what follows is:

$$d\lambda \propto \frac{dm}{m^2}. \quad (29)$$

Here, m represents $\langle m_i \rangle$ defined above. An agent with a particular saving propensity factor λ therefore ends up with a characteristic average money m given by Eqn. (27) such that one can in general relate the distributions of the two:

$$P(m) dm = \rho(\lambda) d\lambda. \quad (30)$$

This, together with Eqn. (27) and Eqn. (28) gives (Mohanty, 2006)

$$P(m) = \rho(\lambda) \frac{d\lambda}{dm} \propto \frac{\rho(1 - \frac{c}{m})}{m^2}, \quad (31)$$

giving $P(m) \sim m^{-2}$ for large m for uniform distribution of savings factor λ , i.e., $\nu = 1$; and $\nu = 1 + \delta$ for $\rho(\lambda) = (1 - \lambda)^\delta$. This study therefore explains the origin of the universal ($\nu = 1$) as well as the non-universal ($\nu = 1 + \delta$) Pareto exponent values in the distributed savings model.

6 Summary and Discussion

Income and wealth distributions across the population in many countries are found to possess some robust characteristics. As has been discussed in section 2, it is empirically found that the bulk (about 90%) of the population fits a gamma like distribution: after an initial steep rise, an exponential decay is seen in the number of persons with income/wealth. There are considerable deviations from exponential decay in the high income/wealth range and the income and wealth data in that range (for the top 5-10% of the population in any country) fit well to Pareto distribution (power law) with the value of the exponent ranging between 1 and 3.

As has been discussed in section 3, the simple exercise of utility maximization (with a well known utility function) in a bilateral trading framework gives rise to a pair of money exchange equations. The system

depicted by this set of equations is conserved and the coupling behaviour is captured well by the same set of equations. This has led to a completely new, statistical formulation of the models of market economies. The dynamics of money in such a model, reveals interesting features about the steady state distribution of money among the interacting agents. Self-organisation is a key emerging feature of these kinetic exchange models when saving factors are introduced. In the model with uniform savings (see Sec. 4.2), the Gamma-like distribution of money shows stable distribution with a most-probable value indicative of an economic scale dependent on the saving propensity, λ . Empirical observations in homogeneous groups of individuals supports this theoretical prediction (see e.g., Willis and Mimkes, 2004; Angle, 2006). The moments of the distribution can be found (see Sec. 5) very easily.

Next, the saving propensity is assumed to vary from agent to agent (see Sec. 4.3). The emergence of a power law tail in money-holding is apparent in cases where the saving factor does not change with tradings or time t for the same agent (i.e., where each trader has a different characteristic saving propensity). The money exchange equations can be cast into a master equation, and the solution to the steady state money distribution giving the Pareto law with $\nu = 1$ have been derived using several approaches (see Sec. 5). Then we discuss two different models focusing on different economic institutions that can also give rise to the same gamma function-like behavior for the distribution of assets in the steady state. The first one considers taxation (see Sec. 4.4) whereas the second one considers insurance against losses (see Sec. 4.5). The moments of the resulting distributions of the last model can be found up to any order (see Sec. 5). The possibilities of emergence of self organizations in markets, evolution of the steady state distributions, emergence of Gamma-like distribution for the bulk and the power law tail, are seen to be captured well by this class of market models.

It has been debated for long whether the distributions derived from these models are representing the income distributions or wealth distributions or simply, the distributions of a conserved asset called money. Given that we have considered this class of models in the framework of general equilibrium theory, we opt for the last interpretation. Clearly,

money is treated as a commodity here which has no storage cost. An important role of money in this model is that it is an asset which transfers purchasing power to the future (see section 3). For the sake of clarity, it may be mentioned that the distributions derived so far, are concerned with money only. Strictly speaking, there is no income (no wage earned from labor supply) in these models and neither is there any wealth accumulation (no capital stock). The reason is that the production side is completely ignored in these models. This may be considered as a future direction of research to consider a model of production and to derive income/wealth distribution directly from that framework. However, the essential nature of both income and wealth and their distributions are captured very well by this class of models. In short, in this class of models money (asset) works as a proxy for income/wealth. Since the distributions derived for money compares extremely well with the empirical data of income/wealth, we believe that these models provide important insights for income/wealth distributions as well.

This class of models has also been criticized for assuming that the *law of conservation* holds. Gallegati *et al* (2006) notes that “in industrialized capitalist economies, income is most definitely not conserved”. While this observation is certainly true that income and wealth in an economy grows over time, it does not contradict the models stated above. The growth of income and wealth over time, is by definition a time-series observation whereas the models presented here tries to explain cross-section observations (data taken at a single instance or within a very short period of time). The main argument in favour of the ideal gas like models is that billions of small transactions that take places in a very short span of time can generate the essential stochastic features of the kinetic exchange models and the corresponding distributions.

Though the models considered above (Sec. 4) follow from established principles of the utility maximization paradigm (Sec. 3) and the analysis of their kinetics (Sec. 5) have a rigorous foundation based on hundred years’ old statistical physics, they are not matured enough yet to be put to use in practice directly. Nevertheless, they present a workable and tractable approach for analysing a statistically large economy. They illuminate the statistical effects of a number of mechanisms and institutions of the

economy and reproduce the distributions of assets seen in reality quite reliably; as such they may provide a new foundation of macroeconomics (Lux and Westerhoff, 2009). In future, policy making may also benefit from such detailed understanding of the mechanisms by which distributions of income and wealth emerge out of collective exchanges (Hogan, 2005).

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