

First-Passage Processes in Financial Markets

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Recently, lots of on-line trading services on the internet were constructed by several major banks such as the Sony Bank. They use a trading system in which foreign currency exchange rates change according to a first-passage process (FPP). Automatic FOREX trading systems are popular in Japan where many investors use a scheme called *carry trade* by borrowing money in a currency with low interest rate and lending it in a currency offering higher interest rates. In this paper, we investigate the statistical properties of the FPP from three different points of view, namely, *i) probabilistic-theoretical analysis*, *ii) empirical data analysis of BTP futures* (BTP is the middle and long term Italian Government bonds with fixed interest rates), and *iii) minority game modeling of the market*.

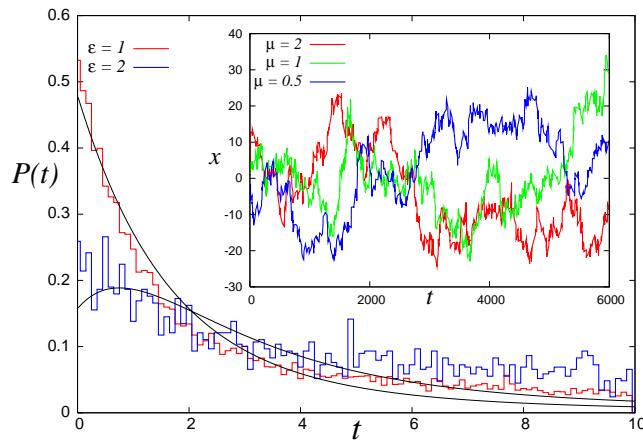


Figure 1: Theoretical prediction of the FPT pdf $P(t)$ (black solid-lines). The inset shows typical behaviour of a mixture of the NCPPs.

finite by taking into account the distribution of θ . In this paper, we assume that the parameters μ, θ and σ are all independent variables: $g(\mathbf{w}) = g(\mu, \theta, \sigma) = f_\mu(\mu)f_\theta(\theta)f_\sigma(\sigma)$ and choose each form of the functions f_μ , etc. reasonably. Then, it should be noted that the parameter σ is not a relevant parameter in the sense that one can naturally rescale the parameters such as $\theta/\sigma \rightarrow \theta, \epsilon/\sigma \rightarrow \epsilon$ to cancel it. For the f_μ , we assume $f_\mu(\mu) = (1/\beta) \sum_{l=\mu_1}^{\mu_2} \delta(\mu - l)$. This choice is justified from the fact that the inverse of relaxation time μ takes the value dependent on the time of the day (*Time-of-the-Day Dependence*). For instance, the μ might take $1/10$ during the period $k \in [0, 6]$ and takes $1/20$ for $k \in [6, 12]$, $1/30$ for $k \in [12, 18]$ and $1/10$ for $k \in [18, 24]$.

In our FPP, the stochastic variable is updated if and only if the time k -dependent stochastic variable x_k underlying the FPP holds $|x_{k+t} - x_k| \geq \epsilon$, where the duration t of the stochastic process.

i) probabilistic-theoretical analysis: We assume that the t is generated by the joint probability density $p(x, t) = \int d\mathbf{w} g(\mathbf{w})p(x, t|\mathbf{w}), \int d\mathbf{w} (\dots) \equiv \int_0^\infty d\mu \int_{-\infty}^\infty d\theta \int_0^\infty d\sigma (\dots)$. $p(x, t|\mathbf{w})$ denotes the one-point probability density for a single normal compound Poisson process (NCPP) starting at the position $x = 0$ for $t = 0$ is given by $p(x, t|\mathbf{w}) = e^{-\mu t} \sum_{n=0}^\infty (\mu t)^n e^{-(x-n\theta)^2/2n\sigma^2} / n! \sqrt{2\pi n\sigma^2}$. We should notice that the average price x of a single NCPP diverges as $\langle x \rangle \equiv \int_0^\infty dt \int_{-\infty}^\infty p(x, t|\mathbf{w}) dx = (\theta/\mu) \sum_{n=0}^\infty n$ for a constant θ , whereas it remains

Thus, in general, the distribution of μ converges to the above form $f_\mu(\mu)$ in the long-observation limit $k \rightarrow \infty$ of the market price x . When we assume that the θ obeys a normal Gaussian $\mathcal{N}(0, 1)$, we have the first-passage time (FPT) probability distribution function analytically as $P(t) = \beta^{-1} \sum_{l=\mu_1}^{\mu_\beta} e^{-lt} \sum_{n=0}^{\infty} \left(\mu - \frac{n}{t} \right) \frac{(lt)^n}{n!} \int_{-\infty}^{\infty} D\theta H_n(\theta, \epsilon)$ where $D\theta \equiv d\theta e^{-\theta^2/2}/\sqrt{2\pi}$ and we defined $H_n(\theta, \epsilon)$ as the difference between two error functions: $H(\sqrt{n}\theta - \epsilon/\sqrt{n}) - H(\sqrt{n}\theta + \epsilon/\sqrt{n})$. In Figure 1, the results for several widths ϵ of the rate-window are shown. We find that the analytical results of $P(t)$ agree with computer simulations. For $\epsilon < \sqrt{\pi/2}$, we have the mixture of exponentials with effective relaxation time Δ^{-1} as an approximation:

$$P(t) \simeq \frac{1}{\beta} \left(1 - \epsilon \sqrt{\frac{2}{\pi}} \right) \sum_{l=\mu_1}^{\mu_\beta} l e^{-lt} + \epsilon \Delta \sqrt{\frac{2}{\pi}} e^{-\Delta t}, \quad \Delta \equiv -t^{-1} \log e^{-lt} \sum_{n=0}^{\infty} \frac{(lt)^n}{n! \sqrt{n(n+1)}} > 0$$

From this expression, we find that in the $\epsilon \rightarrow 0$ limit, the mixture of exponential distributions that describes the point process for the market rate x underlying behind the FPP is recovered. For this FPT distribution, we evaluate the relevant statistics, namely, the residual life time w [4] and the Gini index [5]. In this paper, we calculate the w numerically and discuss the condition on which the so-called inspection paradox: $w > \int_0^\infty dt t P(t)$ takes place. **ii) Empirical data analysis:** To investigate to what extent the FPP changes the stochastic properties of financial markets, we investigate the Kullback-Leibler (KL) distance between the distribution of duration for the BTP future (from 20/01/1997 to 04/06/1997 with the skewness 0.087 and the kurtosis 23.75 for the return distribution) and the distribution of the first-passage time generated from the rate window with a width a . We calculate the KL distance K as a function of a . We find that the K monotonically increases as a increases. For instance, $K = 0.405643$ with $\langle t \rangle = 117.07$, $w = 63645.62$ for $a = 1$ and $K = 18.97$ with $\langle t \rangle = 19253.09$, $w = 71680.57$ for $a = 6$. This result tells us that the statistical properties of the FPP for the BTP future might change from the underlying stochastic process, namely, the BTP future itself. **iii) Minority game modeling:** To model the underlying stochastic process behind the FPP, the minority game might be one of the possible candidates. In our computer simulations, At each round l of the game, each trader i ($i = 1, \dots, N$) decides his (or her) decision: $b_i(l) = +1$ (*buy*) or $b_i(l) = -1$ (*sell*) to joint the minority group. Then, we evaluate the total bit $A(l) = (1/\sqrt{N}) \sum_{i=1}^N b_i(l)$ for each round l . Then, the $A(l)$ follows complicated stochastic process as shown in Figure 2 and one might consider the FPP by regarding the total bit ('price') $A(l)$ as the underlying stochastic process. We evaluate the first passage time distribution and discuss the stochastic properties. The result will be reported in the conference.

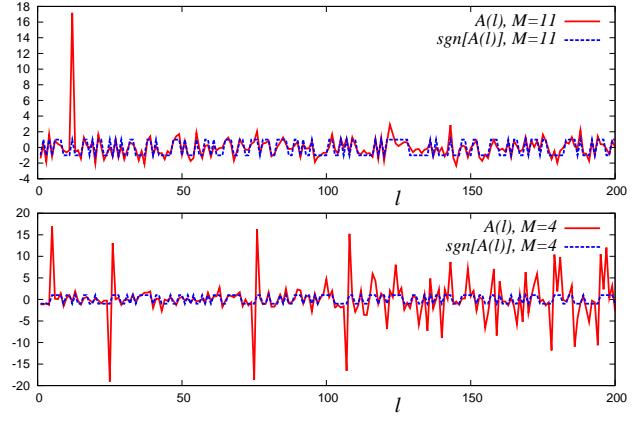


Figure 2: Typical behaviour of the total bit $A(l)$ where l is regarded as 'time step'. M denotes the market history length available to the traders.

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