

Limit Theorems Which Establish a Keeping Intensity Relationship Between Poisson and Ornstein-Uhlenbeck Processes

O.V.RUSAKOV ¹

Saint-Petersburg State University

A number of functional limit theorems which establish a relationships between the Poisson type and Ornstein-Uhlenbeck type processes are given and discussed. Applications to statistics and financial modeling also discussed. We describe a construction for passages to the limit such that the linear time scaling property for the Poisson process directly transits to the same property for the Ornstein-Uhlenbeck process, i.e. $\lambda = \beta$.

As the Ornstein-Uhlenbeck process we understand the stationary, markovian, gaussian process $U(t), t \geq 0$. In the case of U is a centered and normalized process we call it as a standard one. From the definition above it follows that $\text{cov}(U(t_1), U(t_2)) = \exp\{-\beta|t_2 - t_1|\}$, where $\beta > 0$ is a viscosity coefficient.

The Basic Limit Theorem

Let $\{\Pi_i(s)\}$, $i = 1, 2, \dots$, $s \geq 0$, be a sequence of independent identically distributed Poisson processes with the joint intensity $\lambda > 0$.

Let $\{\varepsilon_i^{(j)}\}$, $i = 1, 2, \dots$, $j = 1, 2, \dots$, be a two dimensional array of mutually independent identically distributed random variables (iid rv's), $\mathbf{E}\varepsilon_1^{(1)} = 0$, $\mathbf{D}\varepsilon_1^{(1)} = 1$. As a sampling distribution we consider the simple binomial one: $\varepsilon_1^{(1)} = \pm 1, 1/2$. We suppose that the sequence $\{\Pi_i\}$ is independent of the family $\{\varepsilon_i^{(j)}\}$.

In all text below we suppose that N is a natural number tending to infinity.

We examine limit behavior of sums of the rv's $\{\varepsilon_i^{(j)}\}$ when the upper index $j \equiv j(i)$ is a random variable with values determined by values of the

¹supported in part by the grant of Scientific Schools 4222.2006.1

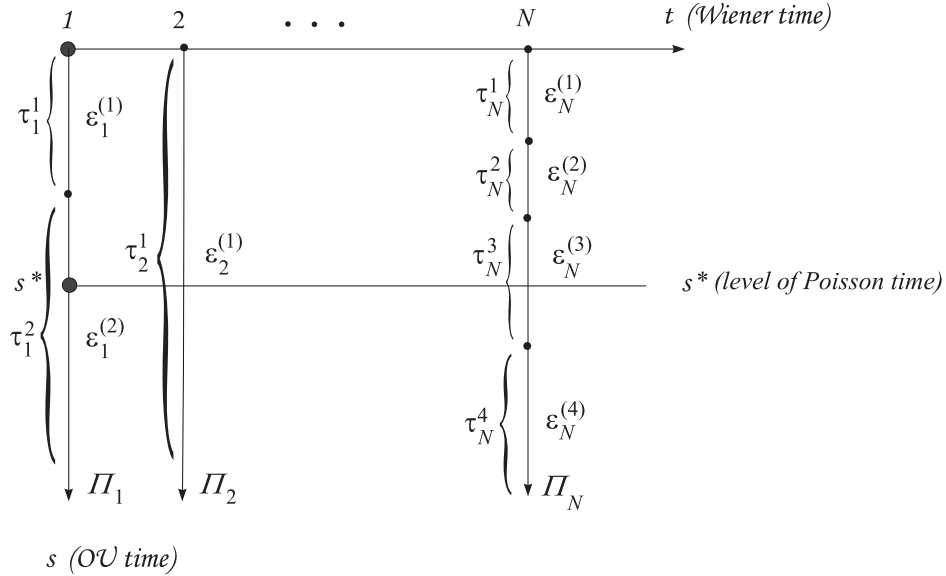
Poisson process Π_i (with the number i). Such kind of constructions has the following interpretation.

It is well known that a time moment for the k -th jump of the Poisson process equals to the sum $\sum_{j=1}^k \tau^j$, where $\{\tau^j\}$, $j = 1, 2, \dots$, — is the sequence of iid rv's having the joint exponential distribution law $\mathcal{Exp}(\lambda)$, $\lambda > 0$. We denote the interval between the two consecutive time moments of the jumps by

$$\Theta^j \triangleq [\tau^{j-1}, \tau^{j-1} + \tau^j)$$

and name it as j -th spacing, $j = 1, 2, \dots$.

Let us denote $\{\Theta_i^j\}$, $i = 1, 2, \dots$, $j = 1, 2, \dots$ the array of spacings corresponding to the processes $\{\Pi_i\}$. So, the enumeration of the upper index j in $\varepsilon_i^{(j)}$ by values of Poisson processes means that we weight the spacing Θ_i^j by the corresponding random weight $\varepsilon_i^{(j)}$.



Pic.1. An Example of the Sequence of the Poisson Processes for a Limit of the Ornstein-Uhlenbeck Type

Theorem 1. Let $\{0, \frac{1}{N}, \frac{2}{N}, \dots, \frac{N-1}{N}, 1\}$ be the nodes of the uniform partition of the interval $[0, 1] \ni s$. In the nodes we define the \sqrt{N} -normalized

cumulative sums over the Poisson processes

$$S_N(s) \triangleq \frac{1}{\sqrt{N}} \sum_{i=1}^N \varepsilon_i^{(\Pi_i(s)+1)}, \quad (0)$$

and in the remaining points $s \in [0, 1]$ we define the values of $S_N(s)$ by the linear interpolation.

The following weak convergence is valid in the space of continuous function $\mathbf{C}_{[0,1]}$, as $N \rightarrow \infty$,

$$S_N(s) \Longrightarrow U(s),$$

where $U(s)$ is the [standard] Ornstein-Uhlenbeck process with the viscosity coefficient λ coinciding with the intensity coefficient of every Poisson process from the sequence $\{\Pi_i(t)\}$.

Next we consider the generalization of $S_N(s)$, defined by (0), to random broken planes. Following theorem 2 from section below allows us the fact of convergency the random broken planes to a field that we name as the Wiener-Ornstein-Uhlenbeck field (WOU field).

Family of the Wiener-Ornstein-Uhlenbeck Fields

Let $W(t, s)$, $t \geq 0$, $s \geq 0$ be a standard Brownian sheet, i.e. a centered gaussian function with the covariance

$$\text{cov}(W(t_1, s_1), W(t_2, s_2)) = \min(t_1, t_2) \min(s_1, s_2).$$

We define WOU field by the Lamperti transform applying to the time scale s ,

$$Z(t, s) \equiv Z_\beta(t, s) = \exp\{\beta s\} W(t, e^{-2\beta s}), \quad (1)$$

where $\beta > 0$ is the viscosity coefficient.

Model Comment 1. We interpret the time t as an extrinsic (observed) time and we name the time s by the intrinsic time. Model interpretation of the intrinsic time is as follows. Let we observe the Brownian motion (Bm) W (log-prices of some financial asset, for instance) and we have a delay with the fixing of the small increments dW . We suppose that during this delay the increments dW can replace with their independent copies. And

the time between the replacements is driven by the Poisson process Π , i.e. the replacements occur in the points of jumps Π . We assume that for every t the delays of observations $dW(t)$ are independent and identically distributed. Thus we will observe another Bm. So, in terms of WOU field the prime Bm $W(t)$ is the $Z(t, 0)$ and the observed after delays Bm is $Z(t, s^*)$, where s^* is the time of the delay (see Pic.1) and β in Z is the intensity of the driving Poisson processes.

The field $Z(t, s)$ is a tensor product of the Wiener process and the Ornstein-Uhlenbeck process, i.e. the WOU field is a continuous gaussian centered function,

$$\text{cov}(Z(t_1, s_1), Z(t_2, s_2)) = \exp\{-\beta|s_2 - s_1|\} \min(t_1, t_2).$$

Remark that this form of the covariance guarantees the existence of the WOU field.

We consider the diagonal process $Z(t) \triangleq Z(t, t)$ of the following peculiar interest. The fact is that the diagonal process $Z(t)$ is a zero mean a gaussian markovian process with the linearly increasing variance $\mathbf{D}Z(t) = t$.

Model Comment 2. The process $Z(t)$ can play a role of an alternative to the Brownian motion. The covariance of the diagonal process coincide with covariance of the product independent Ornstein-Uhlenbeck and Wiener processes $\text{cov}(Z(t_1), Z(t_2)) = \exp\{-\beta|t_1 - t_2|\} \min(t_1, t_2)$. Such type of covariance directly corresponds to the stochastic volatility models in finance. An effect is that applying the diagonal process we work with the normal distribution for log-prices whereas in the stochastic volatility models the log-prices follows to a complicate distribution with practically very sensitive and unstable parameters.

The essential defect (single, probably) of the diagonal process is that it is not a homogeneous one. The conditional moments of $Z(t)$ are as follows

$$\begin{aligned} \mathbf{E} \{Z(t)|Z(u) = z\} &= ze^{-\beta(t-u)}, \\ \mathbf{D} \{Z(t)|Z(u) = z\} &= t - ue^{-2\beta(t-u)}, \end{aligned} \tag{2}$$

for $u \leq t$.

The simplest way to characterize the diagonal process is by the following Lamperti type transform for the standard Bm W ,

$$Z(t) \stackrel{distr}{=} \exp\{\beta t\} W(te^{-2\beta t}). \quad (3)$$

Proposition 1 below characterizes the diagonal process in the moving average form and in the form of the Langevin stochastic differential equation.

Proposition 1. There exists the Bm $B(u)$ such that

$$Z(t) = e^{-\beta t} \int_0^t e^{\beta u} \sqrt{1 + 2\beta u} dB(u), \quad (4)$$

and

$$dZ(t) = -\beta Z(t)dt + \sqrt{1 + 2\beta t} dB(t) \quad (5)$$

with the starting point $Z(0) = 0$.

We consider the following generalization of the WOU field to the non-homogeneous case named the n-hWOU field. Let $f(s)$, $s \geq 0$, $f(0) = 1$ be the positive, strictly monotone, increasing, and differential function. We define the n-hWOU field by the following generalization of the transform of Lamperti type

$$Z_f(t, s) = f(s)W\left(t, \frac{1}{f^2(s)}\right). \quad (6)$$

It is easy to calculate the covariance for the n-hWOU field,

$$\text{cov}(Z_f(t_1, s_1), Z_f(t_2, s_2)) = \frac{f(s_1)f(s_2)}{f^2(\min(s_1, s_2))} \min(t_1, t_2). \quad (7)$$

Proposition 2. There exists the Bm $B(u)$ such that the corresponding diagonal process $Z_f(t) \stackrel{\Delta}{=} Z_f(t, t)$ follows to the equality

$$Z_f(t) = \frac{1}{f(t)} \left(\int_0^t f(u) \sqrt{1 + 2u (\ln f(u))'} dB(u) \right). \quad (8)$$

Proposition 3 establishes the relationships between the diagonal processes for fields $Z_f(s, t)$ and $Z(s, t)$.

Proposition 3. From the form of the covariance of the $Z_f(s, t)$ it is directly follows,

$$Z_f(t, t) \stackrel{distr}{=} Z(\ln f(t), t). \quad (9)$$

Model Comment 3. Proposition 3 includes that for investigation of distribution properties of the processes $Z(t, g(t))$, where $g(t)$ is a positive increasing function, it is sufficient to examine the diagonal process $Z_f(t)$ of n-hWOU field with $f = e^g$. In model sense for the intrinsic time the function $g(t)$ regularizes non-homogeneity of the delays in observations of $dW(t)$.

The equation (9) obviously implies that $\mathbf{D}Z_f(t, t) = t$.

The following limit theorem 2 allows us the fact of convergency random broken planes to the [homogeneous] WOU field.

Theorem 2. Let us split the rectangle $[0, T] \times [0, 1] \ni (t, s)$, $T < \infty$ into rectangular parts by the nodes $\left\{ \left(\frac{iT}{N}, \frac{j}{N} \right) \right\}; i, j \in \{0, 1, \dots, N\}$. We define the random broken rectangles as follows

$$S_N(t, s) \equiv \frac{1}{\sqrt{N}} \sum_{i=1}^{[Nt]} \varepsilon_i \left(\Pi_i \left(\frac{[Ns]}{N} \right) + 1 \right),$$

where $[Nt]$ and $[Ns]$ denote the integer parts. Then the following weak convergence is valid in the space of step functions $\mathbf{D}_{[0, T] \times [0, 1]}^{\mathbf{Q}}$ equipping with the uniform metric and having jumps in a countable fixed set, as $N \rightarrow \infty$,

$$S_N(t, s) \implies Z(t, s),$$

where $Z(t, s)$, $(t, s) \in [0, T] \times [0, 1]$, is the WOU field.

The following theorem 3 states a convergency of non-homogeneous Poisson processes (with spacings weighted by the iid rv's $\{\varepsilon\}$) to the non-homogeneous Ornstein-Uhlenbeck (n-hOU) process.

Non-homogeneous Ornstein-Uhlenbeck Process as a Limit of Weighted Non-homogeneous Poisson Processes

We define the n-hOU process by generalization of Lamperti transform,

$$U_f(s) = f(s)W \left(\frac{1}{f^2(s)} \right), \quad (10)$$

where $f(s)$, $s \geq 0$, $f(0) = 1$ is the positive, strictly monotone, increasing, and differential function. Obviously, the n-hOU process defined by (10) is a truncation of the n-hWOU field defined by (6).

In particular case, $f(s) = \exp\{-\beta s\}$ we have in (10) the standard Ornstein-Uhlenbeck process with the viscosity coefficient $\beta > 0$. The fact is that the process $U_f(s)$ is a stationary one iff the function f is the exponential.

The equivalent definition of the n-hOU process we give by the following generalization of the Langevin stochastic differential equation (SDE),

$$dX(t) = -H(t)X(t)dt + \sqrt{2H(t)}dB(t), \quad (11)$$

where the viscosity parameter is as follows

$$H(t) = \frac{d \ln f(t)}{dt}, \quad t \geq 0,$$

and the initial value $X(0) \in \mathcal{N}(0, 1)$ does not depends on the Bm $B(t)$.

Let $\{\Pi_i^f(s)\}$, $i = 1, 2, \dots$ $s \geq 0$, be a sequence of independent identically distributed non-homogeneous Poisson processes (Cox processes) with the joint intensity function $\Lambda(s) = \ln f(s)$, where $f(s)$, $s \geq 0$, $f(0) = 1$ is the positive, strictly monotone, increasing, and differential function.

Theorem 3. Let $\{0, \frac{1}{N}, \frac{2}{N}, \dots, \frac{N-1}{N}, 1\}$ be nodes of the uniform partition of the interval $[0, 1] \ni s$. In the nodes we define the \sqrt{N} -normalized cumulative sums of the variables $\{\varepsilon\}$ indexed by the non-homogeneous Poisson processes

$$S_N^f(s) \equiv \frac{1}{\sqrt{N}} \sum_{i=1}^N \varepsilon_i^{(\Pi_i^f(s)+1)},$$

and in the remaining points $s \in [0, 1]$ we define the values of $S_N^f(s)$ by the linear interpolation.

The following weak convergence is valid in the space of continuous function $\mathbf{C}_{[0,1]}$, as $N \rightarrow \infty$,

$$S_N^f(s) \Longrightarrow U_f(s),$$

where $U_f(s)$ is the non-homogeneous Ornstein-Uhlenbeck process satisfied (10) with the viscosity function

$$H(s) = \frac{d\Lambda(s)}{ds}.$$