Nonlinear Dynamics and Systems Theory, 9(3) (2009) 277-286



# Antagonistic Games with an Initial Phase<sup>†</sup>

# Jewgeni H. Dshalalow\* and Ailada Treerattrakoon

Department of Mathematical Sciences, Florida Institute of Technology, Melbourne, Florida 32901-6975, USA.

Received: March 10, 2009; Revised: June 4, 2009

**Abstract:** We formalize and investigate an antagonistic game of two players (A and B), modeled by two independent marked Poisson processes forming casualties to the players. The game is observed by a third party point process. Unlike previous work on this topic, the initial observation moment is chosen not arbitrarily, but at some random moment of time following initial actions of the players. This caused an analytic complexity unresolved until recently. This, more realistic assumption, forms a new phase ("initial phase") of the game and it turns out to be a short game on its own. Following the initial phase, the main phase of the game lasts until one of the players' cumulative casualties exceed some specified threshold. We investigate the paths of the game in which player A loses the game.

**Keywords:** noncooperative stochastic games; fluctuation theory; marked point processes; Poisson process; ruin time; exit time; first passage time.

Mathematics Subject Classification (2000): 82B41, 60G51, 60G55, 60G57, 91A10, 91A05, 91A60, 60K05.

### 1 Introduction

We model an antagonistic stochastic game by two marked Poisson processes  $\mathcal{A}$  and  $\mathcal{B}$ , each representing casualties incurred to players A and B. The mutual attacks are rendered in accordance with associated Poisson point processes and their marks are distributed arbitrary and position independent. The game is observed by a third party process  $\mathcal{T}$ . Consequently, the information on the game is available upon  $\mathcal{T}$ , thereby forming the embedding  $\mathcal{A}_{\mathcal{T}} \otimes \mathcal{B}_{\mathcal{T}}$ . (The latter is a more general bivariate marked point process with marks being mutually and position dependent.) The game lasts until one of the players

 $<sup>^\</sup>dagger$  This research is supported by the US Army Grant No. W911NF-07-1-0121 to Florida Institute of Technology.

<sup>\*</sup> Corresponding author: eugene@fit.edu

gets "exhausted" or "ruined". This happens whenever the total casualties to the players exceed some specified thresholds. The real exit from the game takes place with a delay in accordance with observations  $\mathcal{T}$ . This is one of the quite common scenarios of games, in which the co-authors [9] (and most recently, the first author [5–8, 12]) have been involved.

A realistic approach to the modeling was rendered through the embedded delayed process  $\mathcal{A}_{\mathcal{T}} \otimes \mathcal{B}_{\mathcal{T}}$  distorting the real time information. However, in the previous models the position of the first observation epoch was placed arbitrarily on the positive time axis with no regard to the start of the conflict. As the result, the initial observation point could have been placed before the game began. In a recent article by Dshalalow and Huang, this deficiency was overcome by placing the first observation at some random time after the conflict has emerged. This alone formed a separate initial phase of the conflict with a joint functional, which included the time of the beginning of the conflict and the amount of casualties to the players, all the way to the first observation. To merge this initial phase with the rest of the game, required some past information (non-Markovian), all resulting in two separate phases, which we thereby have come to identify. From the modeling point of view, the present game is simpler than that of [7], which in contrast, also included a second phase following the initial and first phases.

The first phase of this game ends with player A losing to player B (while in [7] it was not specified who of the two exactly loses, as their casualties were then limited).

Even though our model is not entirely characterized as a sequential game, it comes close enough to this literature [1, 3, 5–7, 11, 12, 14, 15, 18, 21, 24]. The tools we are using in this paper are mainly self-contained and developed methods of fluctuation theory that originated from applications to random walk processes. We hold on classic random walk fluctuation analysis, only in a generalized forms. We mention just a few pieces of literature where applications of the fluctuation theory takes place in the areas such as economics [17] and physics [20]. More on this can be found in [5–9]. Topically, the paper falls into the category of antagonistic stochastic games widely applied to economics [2, 16, 19, 24] and warfare [9, 12, 22, 23]. As in all previous work by the authors and the first author, the results are directly applicable to economics and warfare, in particular, in light of a high volatility of the global economy in the recent months. The latter can be interpreted as an "antagonism" between the economic actions (such as bailout of credit institutions) against the panic of the market.

Another area of applied mathematics that relates to our work includes *hybrid systems* [4, 13], in particular hybrid stochastic games [5]. For more references on this topic see [5].

The layoff of the paper is as follows. Section 2 deals with the formalism of the game. Section 3 takes on the initial phase. Section 4 continues with the game beyond the initial phase until player A is ruined. The merge between the two phases is the main contribution to this section.

#### 2 A Formal Description of the Model

The results of Sections 2 and 3 are based on Dshalalow and Huang [7]. To make it self-contained we follow the initial phase of [7].

Let  $(\Omega, \mathcal{F}(\Omega), \mathfrak{F}_t, P)$  be a filtered probability space and let  $\mathcal{F}_A, \mathcal{F}_B, \mathcal{F}_S \subseteq \mathcal{F}(\Omega)$  be independent sub- $\sigma$ -algebras. We suppose that

$$\mathcal{A} := \sum_{j \ge 1} d_j \varepsilon_{r_j} \text{ and } \mathcal{B} := \sum_{k \ge 1} z_k \varepsilon_{w_k}$$
(2.1)

are  $\mathcal{F}_{\mathcal{A}}$ -measurable and  $\mathcal{F}_{\mathcal{B}}$ -measurable marked Poisson random measures ( $\varepsilon_a$  is a point mass at a) with respective intensities  $\lambda_A$  and  $\lambda_B$  and position independent marking. The random measures are specified by the transforms

$$Ee^{-u\mathcal{A}(\cdot)} = e^{\lambda_A |\cdot|[h_A(u)-1]}, \ h_A(u) = Ee^{-ud_1}, \ Re(u) \ge 0,$$
 (2.2)

$$Ee^{-v\mathcal{B}(\cdot)} = e^{\lambda_B |\cdot|[h_B(v)-1]}, \ h_B(v) = Ee^{-vz_1}, \ Re(v) \ge 0,$$
 (2.3)

where  $|\cdot|$  is the Borel-Lebesgue measure and  $d_j$  and  $z_k$  are nonnegative r.v.'s representing the successive strikes of players B and A against each other, respectively, while  $r_j$  and  $w_k$  are the times of the strikes.

The game starts with hostile actions initiated by one of the players A or B at  $r_1$  or  $w_1$ . The players can exchange with several more strikes before the first information is noticed by an observer at time  $t_0$ . We therefore assume that

$$t_0 \ge \max\{r_1, w_1\}. \tag{2.4}$$

The initial observation time  $t_0$  will be formalized below. All forthcoming observations will be rendered in accordance with a point process

$$T_0 = \sum_{i \ge 0} \varepsilon_{t_i} = \varepsilon_{t_0} + S, \text{ with } S = \sum_{i \ge 1} \varepsilon_{t_i},$$
  
$$0 < t_0 < t_1 < \dots < t_n < \dots (t_n \to \infty, \text{ with } n \to \infty).$$
 (2.5)

We introduce the extension of  $\mathcal{T}$ :

$$\mathcal{T} := \varepsilon_{t_{-1}} + T_0, \text{ with } t_{-1} := \min\{r_1, w_1\},$$
(2.6)

such that the tail  $S = \sum_{i\geq 1} \varepsilon_{t_i}$  of  $T_0$  is  $\mathcal{F}_S$ -measurable. The increments  $\Delta_1 := t_1 - t_0$ ,  $\Delta_2 := t_2 - t_1$ ,  $\Delta_3 := t_3 - t_2$ , ... are all independent and identically distributed, and all belong to the equivalence class  $[\Delta]$  of r.v.'s with the common Laplace-Stieltjes transform

$$\delta(\theta) := E e^{-\theta \Delta}.\tag{2.7}$$

Now we define the initial observation as

$$t_0 = \max\{r_1, w_1\} + \Delta_0, \tag{2.8}$$

where  $\Delta_0 \in [\Delta]$  and  $\Delta_0$  is independent from the rest of the  $\Delta$ 's.  $t_0$  is included in  $T_0$  of equation (2.5) and because it contains some of the  $\mathcal{A}$  and  $\mathcal{B}$ ,  $T_0$  is not  $\mathcal{F}_S$ -measurable. However,  $T_0$  is a delayed renewal process, while  $\mathcal{T}$  is not.

We assign to  $t_{-1}$  the genuine start of the game at time min $\{r_1, w_1\}$  of (2.6). That is,

$$t_{-1} = \min\{r_1, w_1\}. \tag{2.9}$$

Now, since  $t_{-1}$  and  $t_0 - t_{-1}$  are dependent (through  $r_1$  and  $w_1$ ), the extended process  $\mathcal{T}$  of (2.6) is not a renewal process, and not even a delayed renewal, as it was in [5, 6, 8, 9, 12].

It should be clear that  $t_0$  depends upon  $r_1$  and  $w_1$  and thus on  $\mathcal{A}$  and  $\mathcal{B}$ , which makes  $T_0 \mathcal{A} \otimes \mathcal{B}$ -measurable. Define the continuous time parameter process

$$(\alpha(t), \beta(t)) := \mathcal{A} \otimes \mathcal{B}([0, t]), \ t \ge 0, \tag{2.10}$$

to be adapted to the filtration  $(\mathfrak{F}_t)_{t\geq 0}$ . Also introduce its embedding over  $T_0$ :

$$(\alpha_j, \beta_j) := (\alpha(t_j), \beta(t_j)) = \mathcal{A} \otimes \mathcal{B}([0, t_j]), \ j = 0, 1, \dots,$$
(2.11)

which forms observations of  $\mathcal{A} \otimes \mathcal{B}$  over  $T_0$ , with respective increments

$$(\xi_j, \eta_j) := \mathcal{A} \otimes \mathcal{B}((t_{j-1}, t_j]), \ j = 1, \dots$$
(2.12)

In addition, let

$$(\xi_0, \eta_0) := \mathcal{A} \otimes \mathcal{B}((\max\{r_1, w_1\}, t_0])$$

$$(2.13)$$

to be used later on.

Introduce the embedded bivariate marked random measures

$$\mathcal{A}_{T_0} \otimes \mathcal{B}_{T_0} := (\alpha_0, \beta_0) \varepsilon_{t_0} + \sum_{j \ge 1} (\xi_j, \eta_j) \varepsilon_{t_j}, \qquad (2.14)$$

where the marginal marked point processes

$$\mathcal{A}_{T_0} = \alpha_0 \varepsilon_{t_0} + \sum_{i \ge 1} \xi_i \varepsilon_{t_i} \text{ and } \mathcal{B}_{T_0} = \beta_0 \varepsilon_{t_0} + \sum_{i \ge 1} \eta_i \varepsilon_{t_i}$$
(2.15)

are with position dependent marking and with  $\xi_j$  and  $\eta_j$  being dependent. For the forthcoming sections we introduce the Laplace-Stieltjes transform

$$g(u, v, \theta) := Ee^{-u\xi_j - v\eta_j - \theta\Delta_j}, \ Re(u) \ge 0, \ Re(v) \ge 0, \ Re(\theta) \ge 0, \ j \ge 1,$$
(2.16)

which will be evaluated as the follows:

$$E[e^{-u\xi_{j}-v\eta_{j}-\theta\Delta_{j}}] = E[e^{-\theta\Delta_{j}}E[e^{-u\xi_{j}-v\eta_{j}}|\Delta_{j}]]$$

$$= E[e^{-\theta\Delta_{j}}E[e^{-u\mathcal{A}((t_{j-1},t_{j}])}|\Delta_{j}]E[e^{-v\mathcal{B}((t_{j-1},t_{j}])}|\Delta_{j}]]$$

$$= E[e^{-\theta\Delta_{j}} \cdot e^{\lambda_{A}\Delta_{j}(h_{A}(u)-1)} \cdot e^{\lambda_{B}\Delta_{j}(h_{B}(v)-1)}]$$

$$= E[e^{-\{\theta+\lambda_{A}(1-h_{A}(u))+\lambda_{B}(1-h_{B}(v))\}\Delta_{j}}]$$

$$= \delta(\theta^{*}), \ j = 1, 2, \dots, \qquad (2.17)$$

with

$$\theta^* := \theta + \lambda_A (1 - h_A(u)) + \lambda_B (1 - h_B(v)), \qquad (2.18)$$

and  $\delta$  defined in (2.7).

## 3 The Initial Phase of the Game

The entire game will include the recording of the conflict between players A and B known to an observer upon process  $\mathcal{T}$  (informally,  $\{t_{-1}, t_0, t_1, \ldots\}$ ) from its inception upon  $t_{-1}$ followed by the initial observation at time  $t_0$ .  $\mathcal{T}$  is defined below. The actual start of the game at  $t_{-1}$  is unknown to the observer, as this moment takes place prior to  $t_0$ . From the construction of the extended game, the point process  $\mathcal{T}$  is obviously "doubly delayed" (in light of its attachment  $t_{-1}$ ). The information on  $t_{-1}$  will be used in section 4 during the merging process. The initial phase of the game is specified as follows. Define the respective damages to the players at  $t_{-1}$  as

$$(\xi_{-1},\eta_{-1}) := (\alpha_{-1},\beta_{-1}) := (\alpha(t_{-1}),\beta(t_{-1})) = (d_1 \mathbf{1}_{\{r_1 \le w_1\}}, z_1 \mathbf{1}_{\{r_1 \ge w_1\}}).$$
(3.1)

Therefore, the embedded process  $\sum_{k\geq -1} \varepsilon_{t_k}(\alpha_k, \beta_k)$  satisfies the extended initial conditions

$$\mathcal{A}_{t_{-1}} \otimes \mathcal{B}_{t_{-1}} = (\alpha_{-1}, \beta_{-1}) = (d_1, 0), \text{ on trace } \sigma\text{-algebra } \mathcal{F}(\Omega) \cap \{r_1 < w_1\}, \quad (3.2)$$

$$\mathcal{A}_{t_{-1}} \otimes \mathcal{B}_{t_{-1}} = (\alpha_{-1}, \beta_{-1}) = (0, z_1), \text{ on } \mathcal{F}(\Omega) \cap \{r_1 > w_1\},$$
(3.3)

$$\mathcal{A}_{t_{-1}} \otimes \mathcal{B}_{t_{-1}} = (\alpha_{-1}, \beta_{-1}) = (d_1, z_1), \text{ on } \mathcal{F}(\Omega) \cap \{r_1 = w_1\}.$$
(3.4)

The extended version of the game is defined as the bivariate marked point process

$$\mathcal{A}_{\mathcal{T}} \otimes \mathcal{B}_{\mathcal{T}} := (\xi_{-1}, \eta_{-1})\varepsilon_{t_{-1}} + (\alpha_0 - \xi_{-1}, \beta_0 - \eta_{-1})\varepsilon_{t_0} + \sum_{j \ge 1} (\xi_j, \eta_j)\varepsilon_{t_j}$$
(3.5)

(embedded over  $\mathcal{T}$ ).

As we will see it in the next section, the game will require knowledge of  $\mathcal{A}_{\mathcal{T}} \otimes \mathcal{B}_{\mathcal{T}}$  at  $t_{-1}$  and  $t_0$ . Consequently, we begin to work on the functional

$$\phi_0 := \phi_0(a_0, b_0, \vartheta_0, u_0, v_0, \theta_0) = E[e^{-a_0\alpha_{-1} - u_0\alpha_0 - b_0\beta_{-1} - v_0\beta_0 - \vartheta_0 t_{-1} - \theta_0 t_0}]$$
(3.6)

that describes what we call, the *initial phase* of the game. The following theorem is due to Dshalalow and Huang [7].

**Theorem 3.1** The functional  $\phi_0$  of the initial phase of the game satisfies the following formula:

$$\phi_0 = \frac{\lambda_A \lambda_B \delta(\theta_0^*)}{\vartheta_0 + \theta_0 + \lambda_A + \lambda_B} \left( \frac{1}{\theta_A + \lambda_B} h_A(a_0 + u_0) h_B(v_0) + \frac{1}{\theta_B + \lambda_A} h_A(u_0) h_B(b_0 + v_0) \right),\tag{3.7}$$

where

$$\theta_0^* := \theta_0 + \lambda_A (1 - h_A(u_0)) + \lambda_B (1 - h_B(v_0)), \tag{3.8}$$

$$\theta_A := \theta_0 - \lambda_A (h_A(u_0) - 1), \tag{3.9}$$

$$\theta_B := \theta_0 - \lambda_B (h_B(v_0) - 1), \tag{3.10}$$

$$\delta(\theta) := E[e^{-\theta\Delta_0}], \ \Delta_0 \in [\Delta].$$
(3.11)

# 4 The Main Phase of the Game

After passing the initial phase, the game continues with its status registered at epochs  $\mathcal{T}$  and it ends when at least one of the players sustains damages in excess of thresholds M or N. To further formalize the game past  $t_0$  we introduce the following random *exit* indices

$$\mu := \inf \{ j \ge 0 : \alpha_j = \alpha_0 + \xi_1 + \ldots + \xi_j > M \},$$
(4.1)

$$\nu := \inf \{ k \ge 0 : \beta_k = \beta_0 + \eta_1 + \ldots + \eta_k > N \}.$$
(4.2)

Related on  $\mu$  and  $\nu$  are the following r.v.'s:

 $t_{\mu}$  is the nearest observation epoch when player A's damages exceed threshold M,

 $t_{\nu}$  is the first observation of  $\mathcal{T}$  when player B's damages exceed threshold N. Apparently,  $\alpha_{\mu}$  and  $\beta_{\nu}$  are the respective cumulative damages to players A and B at their ruin times. We will be concerned, however, with the ruin time of player A and thus restrict our game to the trace  $\sigma$ -algebra  $\mathcal{F}(\Omega) \cap \{\mu < \nu\}$ . Accordingly, we will target the following functional

$$\phi_{\mu} := \phi_{\mu}(a, b, \vartheta, u, v, \theta) = E[e^{-a\alpha_{\mu-1} - u\alpha_{\mu} - b\beta_{\mu-1} - v\beta_{\mu} - \vartheta t_{\mu-1} - \theta t_{\mu}} \mathbf{1}_{\{\mu < \nu\}}].$$
(4.3)

To calculate a tractable form of  $\phi_{\mu}$  we will use the bivariate Laplace-Carson transform

$$\mathcal{LC}_{pq}(\cdot)(x,y) := xy \int_{p=0}^{\infty} \int_{q=0}^{\infty} e^{-xp-yq}(\cdot)d(p,q), \ Re(x) > 0, \ Re(y) > 0,$$
(4.4)

with the inverse

$$\mathcal{LC}_{xy}^{-1}(\cdot)(p,q) = \mathcal{L}_{xy}^{-1}(\cdot \frac{1}{xy}), \qquad (4.5)$$

where  $\mathcal{L}^{-1}$  is the inverse of the bivariate Laplace transform.

**Theorem 4.1** The functional  $\phi_{\mu}$  of the game on trace  $\sigma$ -algebra  $\mathcal{F}(\Omega) \cap \{\mu < \nu\}$  satisfies the following formula:

$$\phi_{\mu} = \mathcal{L}\mathcal{C}_{xy}^{-1} \left( (\Phi_0^1 - \Phi_0) + \frac{\Phi_0^*}{1 - g} (G^1 - G) \right) (M, N), \tag{4.6}$$

where

$$G := g(u+x, v+y, \theta), \tag{4.7}$$

$$G^{1} := g(u, v + y, \theta),$$
 (4.8)

$$\Phi_0^* := \phi_0(0, 0, 0, a + u + x, b + v + y, \vartheta + \theta), \tag{4.9}$$

$$\Phi_0 := \phi_0(a, b, \vartheta, u + x, v + y, \theta), \tag{4.10}$$

$$\Phi_0^1 := \phi_0(a+x, b, \vartheta, u, v+y, \theta), \tag{4.11}$$

with g and  $\phi_0$  of (2.16) and (3.7), respectively.

**Proof**: First we modify (4.1) and (4.2) for the random exit indices  $\mu$  and  $\nu$  which depend on parameters M and N, now to depend on p and q (being arbitrary nonnegative real numbers), respectively, and working with them as parametric families of r.v.'s:

$$\mu(p) := \inf \{ j \ge 0 : \alpha_j = \alpha_0 + \xi_1 + \ldots + \xi_j > p \}, \ p \ge 0,$$
(4.12)

$$\nu(q) := \inf \{k \ge 0 : \beta_k = \beta_0 + \eta_1 + \ldots + \eta_k > q\}, \ q \ge 0.$$
(4.13)

The functional  $\phi_{\mu}$  will now change to

$$\Phi_{pq} = E[e^{-a\alpha_{\mu(p)-1} - u\alpha_{\mu(p)} - b\beta_{\mu(p)-1} - \nu\beta_{\mu(p)} - \vartheta t_{\mu(p)-1} - \theta t_{\mu(p)}} \mathbf{1}_{\{\mu(p) < \nu(q)\}}].$$
(4.14)

This will follow the paths of the game on the trace  $\sigma$ -algebra  $\mathcal{F}(\Omega) \cap \{\mu(p) < \nu(q)\}$  and yield:

$$\Phi_{pq} = \sum_{j \ge 0} \sum_{k>j} E[e^{-a\alpha_{j-1} - u\alpha_j - b\beta_{j-1} - v\beta_j - \vartheta t_{j-1} - \theta t_j} \mathbf{1}_{\{\mu(p) = j, \nu(q) = k\}}].$$
(4.15)

By Fubini's theorem, and that

282

$$\mathcal{LC}_{pq}\left(\mathbf{1}_{\{\mu(p)=j,\nu(q)=k\}}\right)(x,y) = (e^{-x\alpha_{j-1}} - e^{-x\alpha_j})(e^{-y\beta_{k-1}} - e^{-y\beta_k}),$$

(which can be readily shown) we have

$$\mathcal{LC}_{pq}(\Phi_{pq})(x,y) = \sum_{j\geq 0} \sum_{k>j} E[e^{-a\alpha_{j-1}-u\alpha_j-b\beta_{j-1}-v\beta_j-\vartheta t_{j-1}-\theta t_j} \\ \times (e^{-x\alpha_{j-1}}-e^{-x\alpha_j})(e^{-y\beta_{k-1}}-e^{-y\beta_k})].$$
(4.16)

We distinguish two cases.

(i) Case j = 0. This case will include the entire information on the initial phase observed at  $t_0$  and prior to  $t_0$ , including  $t_{-1}$ . In a few lines below, we are going to implement the result of Theorem 3.1 and utilize all necessary versions of the functional  $\phi_0$  :

$$\sum_{k>0} E[e^{-a\alpha_{-1}-u\alpha_{0}-b\beta_{-1}-v\beta_{0}-\vartheta t_{-1}-\theta t_{0}}(e^{-x\alpha_{-1}}-e^{-x\alpha_{0}})(e^{-y\beta_{k-1}}-e^{-y\beta_{k}})]$$

$$=\sum_{k>0} E[e^{-a\alpha_{-1}-u\alpha_{0}-b\beta_{-1}-v\beta_{0}-\vartheta t_{-1}-\theta t_{0}}(e^{-x\alpha_{-1}}-e^{-x\alpha_{0}})\times e^{-y\beta_{0}}e^{-y(\eta_{1}+ldots+\eta_{k-1})}(1-e^{-y\eta_{k}})]$$

$$=\left\{E[e^{-(a+x)\alpha_{-1}-u\alpha_{0}-b\beta_{-1}-(v+y)\beta_{0}-\vartheta t_{-1}-\theta t_{0}}]\times \sum_{k>0} E[e^{-y(\eta_{1}+...+\eta_{k-1})}(1-e^{-y\eta_{k}})]\right\}$$

$$=\left\{\phi_{0}(a+x,b,\vartheta,u,v+y,\theta)-\phi_{0}(a,b,\vartheta,u+x,v+y,\theta)\right\}\times \sum_{k>0} [g(0,y,0)]^{k-1}(1-g(0,y,0))$$

$$=\Phi_{0}^{1}-\Phi_{0},$$
(4.17)

where the summation over k > 0 converges to 1 as per Lemma 1 of Dshalalow and Huang [5]: the associated convergence of  $\sum_{k>0} [g(0, y, 0)]^{k-1}$  is guaranteed provided that Re(y) > 0. The last line in (4.17) is due to notation (4.9-4.11).

(ii) Case j > 0. This case also contains parts of functional  $\phi_0$  in the information related to the reference point  $t_0$ .

Transformation (4.16) for this case is

,

$$\begin{split} \sum_{j>0} \sum_{k>j} E[e^{-a\alpha_{j-1}-u\alpha_{j}-b\beta_{j-1}-v\beta_{j}-\vartheta t_{j-1}-\theta t_{j}}(e^{-x\alpha_{j-1}}-e^{-x\alpha_{j}})(e^{-y\beta_{k-1}}-e^{-y\beta_{k}})] \\ &= \sum_{j>0} \sum_{k>j} \left\{ E[e^{-(a+u+x)\alpha_{j-1}-(b+v+y)\beta_{j-1}-(\vartheta+\theta)t_{j-1}}] \\ &\times E[e^{-u\xi_{j}}(1-e^{-x\xi_{j}})e^{-(v+y)\eta_{j}-\theta\Delta_{j}}]E[e^{-y(\eta_{j+1}+...+\eta_{k-1})}(1-e^{-y\eta_{k}})] \right\} \\ &= \sum_{j>0} \left\{ E[e^{-(a+u+x)\alpha_{0}-(b+v+y)\beta_{0}-(\vartheta+\theta)t_{0}}] \\ &\times E[e^{-(a+u+x)(\xi_{1}+...+\xi_{j-1})-(b+v+y)(\eta_{1}+...+\eta_{j-1})-(\vartheta+\theta)(\Delta_{1}+...+\Delta_{j-1})}] \\ &\times E[e^{-u\xi_{j}}(1-e^{-x\xi_{j}})e^{-(v+y)\eta_{j}-\theta\Delta_{j}}] \sum_{k>j} E[e^{-y(\eta_{j+1}+...+\eta_{k-1})}(1-e^{-y\eta_{k}})] \right\}, \end{split}$$

where the third factor can be written as

$$E[e^{-u\xi_j - (v+y)\eta_j - \theta\Delta_j}] - E[e^{-(u+x)\xi_j - (v+y)\eta_j - \theta\Delta_j}] = G^1 - G^1$$

(as per notation (4.7-4.8)) and the summation over k > j converges to 1, for Re(y) > 0, as per Lemma 1 of [5]. Then, after some algebra in (4.18) and the use of notation (4.7-4.8)and (4.18), we arrive at

$$\phi_0(0,0,0,a+u+x,b+v+y,\vartheta+\theta) \cdot \sum_{j>0} g^{j-1} \cdot (G^1 - G)$$

$$= \Phi_0^* \cdot \sum_{j>0} g^{j-1} \cdot (G^1 - G) = \frac{\Phi_0^*}{1-g} (G^1 - G),$$
(4.19)

with the convergence of  $\sum_{j>0} g^{j-1}$  under the condition that the parameters of g satisfy

$$Re(a + u + x) > 0, \ Re(b + v + y) > 0, \ Re(\vartheta + \theta) > 0,$$
 (4.20)

with any two of the three strict inequalities relaxed with  $\geq$ .

With the cases j = 0 and j > 0 combined together, we will arrive at

$$\mathcal{LC}_{pq}(\Phi_{pq})(x,y) = (\Phi_0^1 - \Phi_0) + \frac{\Phi_0^*}{1-g}(G^1 - G).$$
(4.21)

Remark 4.1 For the particular case

$$\varphi_{\mu} = \varphi_{\mu}(u, v, \vartheta) = E[e^{-u\alpha_{\mu} - v\beta_{\mu} - \theta t_{\mu}} \mathbf{1}_{\{\mu < \nu\}}]$$
(4.22)

of the functional  $\phi_{\mu}$  we get from (4.21)

$$\mathcal{LC}_{pq}(\varphi_{pq})(x,y) = \Phi_0^1 - \Phi_0 \frac{1 - G^1}{1 - G}, \qquad (4.23)$$

where  $\varphi_{pq}$  is the corresponding marginal reduction of  $\Phi_{pq}$  while the rest of the marginal functionals  $G, G^1, \Phi_0$ , and  $\Phi_0^1$  will shrink but for convenience carry the same characters:

$$G = g(u+x, v+y, \theta), \tag{4.24}$$

$$G^{1} = g(u, v + y, \theta), \qquad (4.25)$$

$$G = g(u + x, v + y, \theta),$$

$$G^{1} = g(u, v + y, \theta),$$

$$\Phi^{*}_{0} = \Phi_{0} = \phi_{0}(0, 0, 0, u + x, v + y, \theta),$$

$$(4.26)$$

$$U^{1} = \phi_{0}(u, 0, 0, u + x, v + y, \theta),$$

$$(4.26)$$

$$\Phi_0^1 = \phi_0(x, 0, 0, u, v + y, \theta).$$
(4.27)

Explicitly,

$$\mathcal{LC}_{pq}(\varphi_{pq})(x,y) = \phi_0(x,0,0,u,v+y,\theta) - \phi_0(0,0,0,u+x,v+y,\theta) \frac{1-g(u,v+y,\theta)}{1-g(u+x,v+y,\theta)},$$
(4.28)

where from (3.7-3.10) and (2.18), the marginal versions of  $\phi_0$  needed for (4.28) are

$$\phi_0(x,0,0,u,v,\theta) = E[e^{-x\alpha_{-1}-u\alpha_0-v\beta_0-\theta t_0}]$$

$$= \frac{\lambda_A \lambda_B \delta(\theta^*)}{\theta + \lambda_A + \lambda_B} \left( \frac{1}{\theta_A + \lambda_B} h_A(x+u) h_B(v) + \frac{1}{\theta_B + \lambda_A} h_A(u) h_B(v) \right), \quad (4.29)$$

$$\phi_0(0,0,0,u,v,\theta) = E[e^{-u\alpha_0-v\beta_0-\theta t_0}]$$

$$= \frac{\lambda_A \lambda_B \delta(\theta^*)}{\theta + \lambda_A + \lambda_B} \left( \frac{1}{\theta_A + \lambda_B} h_A(u) h_B(v) + \frac{1}{\theta_B + \lambda_A} h_A(u) h_B(v) \right), \tag{4.30}$$

and

$$\theta_0^* := \theta + \lambda_A (1 - h_A(u)) + \lambda_B (1 - h_B(v)), \tag{4.31}$$

$$\theta_A := \theta - \lambda_A (h_A(u) - 1), \tag{4.32}$$

$$\theta_B := \theta - \lambda_B (h_B(v) - 1). \tag{4.33}$$

**Concluding Remarks.** In this paper, we study fully antagonistic stochastic games of two players (A and B) (initiated in [5-7]), modeled by two independent marked Poisson processes recording times and quantities of casualties to the players. The game is observed by a third party renewal point process upon which the information is gathered (and a decision about upcoming steps can be made or modified). Unlike previous work in [5, 6, 8, 9], the initial observation moment is not arbitrarily chosen, but it is placed at random following some initial actions of the players. This caused an analytic complexity which was unresolved until recently. Due to this more realistic assumption a new phase in the game emerged, which we name the "initial phase". This initial phase turned out to be a short game on its own. Following the initial phase, the main phase of the game lasts until one of the players is ruined. This takes place when the cumulative casualties of a losing player exceed some specified threshold. We investigate the paths of the game in which player A loses the game. The general formulas are obtained in closed forms. In [10] we will render calculation for a variety of special cases.

#### References

- Altman, E. and Gaitsgory, V. A hybrid (differential-stochastic) zero-sum game with fast stochastic part. In: *New Trends in Dynamic Games* (ed. by Olsder, G.J.), Birkhäuser, 1995, 46–59.
- Bagwell, K. Commitment and Observability in Games. Games and Economic Behavior 8 (2) (1995) 271–280.
- [3] Brandts, J. and Solàc, C. Reference Points and Negative Reciprocity in Simple Sequential Games. Games and Economic Behavior 36 (2) (2001) 138–157.
- [4] Collins, P. Chaotic dynamics in hybrid systems. Nonlinear Dynamics and Systems Theory 8 (2) (2008) 169–194.
- [5] Dshalalow, J.H. and Huang, W. On noncooperative hybrid stochastic games. Nonliear Analysis: Special Issue Section: Analysis and Design of Hybrid Systems 2 (3) (2008) 803– 811.
- [6] Dshalalow, J.H. and Huang, W. A stochastic game with a two-phase conflict. Jubilee Volume: Legacy of the Legend, Professor V. Lakshmikantham. Cambridge Scientific Publishers, Chapter 18, (2009) 201–209.
- [7] Dshalalow, J.H. and Huang, W. Sequential antagonistic games with initial phase (jointly with Weijun Huang). To appear in *Functional Equations And Difference Inequalities and Ulam Stability Notions*, Dedicated to Stanislaw Marcin ULAM, on the occasion of his 100-th birthday anniversary. In Press.
- [8] Dshalalow, J.H. and Ke, H-J. Layers of noncooperative games. Nonliear Analysis, Series A. In press.
- [9] Dshalalow, J.H. and Treerattrakoon, A. Set-theoretic inequalities in stochastic noncooperative games with coalition. *Journal of Inequalities and Applications*. Art. ID 713642, 14 pp. (2008).

- [10] Dshalalow, J.H. and Treerattrakoon, A. Operational calculus in noncooperative stochastic games, *Nonlinear Dynamics and Systems Theory* (accepted for publication).
- [11] Exman, I. Solving sequential games with Boltzmann-learned tactics. In: Lecture Notes In: Computer Science, 496, 216–220. Proceedings of the 1st Workshop on Parallel Problem Solving from Nature, Springer-Verlag London, UK, 1990.
- [12] Huang W. and Dshalalow, J.H. Tandem Antagonistic Games, Nonliear Analysis, Series A, in press.
- [13] Khusainov, D., Langerak, R., Kuzmych, O. Estimations of solutions convergence of hybrid systems consisting of linear equations with delay. *Nonlinear Dynamics and Systems Theory* 7(2) (2007) 169–186.
- [14] Kobayashi, N. Equivalence between quantum simultaneous games and quantum sequential games. Submitted to *Quantum Physics*.
- [15] Kohler, D.A. and Chandrasekaran, R. A Class of Sequential Games. Operations Research, INFORMS 19(2) (1971) 270–277.
- [16] Konstantinov, R.V. and Polovinkin, E.S. Mathematical simulation of a dynamic game in the enterprise competition problem. *Cybernetics and Systems Analysis* 40 (5) (2004) 720–725.
- [17] Kyprianou, A.E. and Pistorius, M.R. Perpetual options and Canadization through fluctuation theory. Ann. Appl. Prob. 13 (3) (2003) 1077–1098.
- [18] Radzik, T. and Szajowski, K. Sequential Games with Random Priority. Sequential Analysis 9(4) (1990) 361–377.
- [19] Ragupathy, R. and Das, T. A stochastic game approach for modeling wholesale energy bidding in deregulated power markets. *IEEE Tras. on Power Syst.* **19** (2) (2004) 849–856.
- [20] Redner, S. A Guide to First-Passage Processes. Cambridge University Press, Cambridge, 2001.
- [21] Siegrist, K. and Steele, J. Sequential Games. J. Appl. Probab. 38(4) (2001) 1006–1017.
- [22] Shashikin, V.N. Antagonistic game with interval payoff functions. Cybernetics and Systems Analysis 40(4) (2004) 556–564.
- [23] Shima, T. Capture Conditions in a Pursuit-Evasion Game between Players with Biproper Dynamics. Journal of Optimization Theory and Applications 126(3) (2005) 503–528.
- [24] Wen, Q. A Folk Theorem for Repeated Sequential Games. The Review of Economic Studies 69(2) (2002) 493–512.