

Article

# Decision-Making Processes Underlying Pedestrian Behaviors at Signalized Crossings: Part 2. Do Pedestrians Show Cultural Herding Behavior?

Marie Pelé <sup>1</sup>, Jean-Louis Deneubourg <sup>2</sup> and Cédric Sueur <sup>1,3,\*</sup>

<sup>1</sup> Université de Strasbourg, CNRS, IPHC UMR 7178, 67000 Strasbourg, France; marie.pele@iphc.cnrs.fr

<sup>2</sup> Center for Nonlinear Phenomena and Complex Systems (Cenoli)-CP 231, Université libre de Bruxelles, 1050 Bruxelles, Belgium; Jean-Louis.Deneubourg@ulb.ac.be

<sup>3</sup> Institut Universitaire de France, 75002 Paris, France

\* Correspondence: cedric.sueur@iphc.cnrs.fr; Tel.: +33-388-107-453

Received: 12 September 2019; Accepted: 8 November 2019; Published: 12 November 2019

**Abstract:** Followership is generally defined as a strategy that evolved to solve social coordination problems, and particularly those involved in group movement. Followership behavior is particularly interesting in the context of road-crossing behavior because it involves other principles such as risk-taking and evaluating the value of social information. This study sought to identify the cognitive mechanisms underlying decision-making by pedestrians who follow another person across the road at the green or at the red light in two different countries (France and Japan). We used agent-based modelling to simulate the road-crossing behaviors of pedestrians. This study showed that modelling is a reliable means to test different hypotheses and find the processes underlying decision-making when crossing the road. We found that two processes suffice to simulate pedestrian behaviors: personal motivation and imitation. Importantly, the study revealed differences between the two nationalities and between sexes in the decision to follow and cross at the green and at the red light. Japanese pedestrians showed a greater mimetic behavior at the red light but the process takes into account both the number of crossing and waiting pedestrians, contrary to French citizens. Finally, the simulations are revealed to be similar to observations, not only for the departure latencies but also for the number of crossing pedestrians and the rates of illegal crossings. The conclusion suggests new solutions for safety in transportation research.

**Keywords:** followership; modelling; agent-based model; road crossing; social information; risk taking

---

## 1. Introduction

“We are discreet sheep; we wait to see how the drove is going, and then go with the drove.” [1]

The fundamentally gregarious nature of human beings leads them to follow simple rules in the context of collective phenomena such as panic, pedestrian traffic flow and crowd coordination [2–5]. Indeed, who can resist following a person who is running in a specific direction, even without knowing if it is justifiable? During such collective events, it has often been shown that a number of individuals take the initiative, then others follow [6,7]. This followership is generally defined as a strategy “that evolved for solving social coordination problems in ancestral environments, including in particular the problems of group movement, intragroup peacekeeping and intergroup competition” [8,9]. This followership behavior is particularly interesting in the context of road-crossing behavior because it involves other principles such as risk taking [10,11] and evaluating the value of social information [12,13].

Walking in the street is a necessary daily behavior that is generally considered to be safe in the light of the number of times people cross the road without getting struck by a vehicle. Yet crossing the road causes the largest number of pedestrian accidents and the most severe injuries [14]. Indeed, circumstances such as the urban and social environment can make crossing the road a high-risk behavior. One of these circumstances is the misuse of social information [12,15]. Indeed, social information, i.e. how to trust information gained from people surrounding us, is an important topic in the context of human collective phenomena and followership [16–19]. In the case of road-crossing behaviors at signalized crossings, trusting wrong or unreliable information and following someone crossing at the red light without checking the light color by ourselves (i.e., personal information) leads to an increased risk of accidents and injuries [20]. This probability of following depends on many factors such as gender or spatial proximity [12] and is governed by an amplification process also called information cascade or mimetic process [12,15,21,22]. An information cascade [21,23] occurs at signalized crossings when pedestrians observe the actions of others and then do likewise, despite possible contradictions with their personal information. The mimetic process is quite similar, except that in this case the probability of crossing increases with the number of pedestrians that have already started crossing according to what we call a mimetic coefficient [22]: the higher the number of pedestrians that have already started crossing, the more likely it is that the remaining individuals will follow them, even if the former took the wrong decision (i.e., they crossed at the red light). However, some studies also suggest that behaviors and reactions vary according to the socio-demographic traits of individuals. For example, men show more high-risk behaviors than women, who display more compliance and conformity [24–26]. In the same way, the rate of rule-breaking differs according to the country in which the pedestrian study is done, and this is not due to a difference in the risk of car-pedestrian accidents but to social conventions [27] and cultural differences between societies that are conformist or individualistic [28–30].

In this study, we wanted to highlight the cognitive mechanisms underlying decision-making in pedestrians who follow another person who is crossing at the green or red light. Although many studies have attempted to understand which factors influence the number of rule-breaking incidents at road-crossings, very few have focused on the real decision-making processes of pedestrians facing the situation of following other pedestrians at a signalized crossing and how the decision to cross is affected by changes in the number of pedestrians that have started crossing [31,32]. We compared the behavior according to the country (France and Japan), but also differentiated between crossings according to the gender of individuals. Mathematical equations and survival analyses were used to better understand the cognitive mechanisms underlying the decision-making process [22,33]. However, as these types of analyses may only provide a range of values to explain behaviors rather than a precise figure of the considered parameter, we also used agent-based models to simulate decision-making processes and followership behaviors [34–36]. The use of agent-based models allows to recreate all the variables observed at pedestrian crossings, test them thousands of times and then compare the simulated results with those observed. This kind of analysis is not possible using traditional statistical approaches [36,37].

Different studies [4,7,12,15] assumed that persons crossing the road after one or several pedestrians had done likewise show an amplification process or a mimetic behavior. This means that the higher the number of pedestrians crossing the road, the higher the probability is that others will follow them. We tested this mimetism hypothesis against the null hypothesis of independence, meaning that the probability of crossing the road is not governed by an amplification process but solely by each pedestrian's intrinsic probability of crossing. This suggests that at least two processes underlying pedestrian decisions: their own motivation, or *intrinsic probability*, and the likelihood that they will be influenced by others, i.e., the *mimetic process* [22,38]. These processes might, however, vary according to individuals and their socio-demographic traits, which leads us to make several assumptions. First, the intrinsic probability of crossing and the mimetic process should both be lower at the red light compared to the green light because a minority, if not a majority of pedestrians respect the rules and consider risks [39–41]. As far as the country or cultural effect are concerned, French pedestrians are expected to show a higher intrinsic probability to cross at the red light but no

difference at the green light, as French people are known to be more individualistic and less conformist than Japanese people [11,15]. The mimetic process should be different between French and Japanese pedestrians, and mimetism in French citizens is expected to be stronger mimetic process at the red light but lower at the green light compared to their Japanese counterparts. However, as the Japanese are more conformist and are aware of group pressure [28–30], they are expected to consider not only the number of pedestrians that are already crossing, but also the number of pedestrians remaining on the sidewalk, whilst French pedestrians are expected to only take the number of crossing pedestrians into account. This process, which takes both the number of moving and of waiting individuals into account, is a “following the majority” process [23,42] that has also been observed in sheep during decisions to move together [43,44]. Men are shown to take more risks [11,25] and be less concerned by group pressure and should, therefore, show higher intrinsic probability than women to cross the road at the red light for instance. Similarly, men are expected to follow the pedestrians who have already started crossing, and are not expected to consider the number of waiting pedestrians, whilst women are more likely to check the behaviors of resting pedestrians as they are more sensitive to group pressure [24,26].

## 2. Materials and Methods

### 2.1. Study Sites

We observed pedestrian behaviors at three sites in Strasbourg, France and at four sites in Nagoya, Japan. Details of each site are given in Table 1. Pictures of each site are available in [15]. These sites all permitted the observation of collective road crossings involving at least 10 pedestrians at a time. The speed of vehicles on each site was limited to 50 kmh<sup>-1</sup>. There was no difference in pedestrian crossing speed between the sites (permutation test for independent samples:  $\max T = 2.22$ ,  $p = 0.168$ ). At some sites, vehicles were allowed to turn left or right despite the green light for pedestrians, but the drivers were aware that crossing pedestrians had priority.

Moreover, turning vehicles travel much slower than vehicles that are heading straight on. However, the driver of an approaching vehicle may be less careful if pedestrians cross at the red light, as he/she has the right of way. The risk to pedestrians is, therefore, much higher when crossing at the red light. There was no button for pedestrians to trigger the green light at any of the sites studied.

**Table 1.** Information about the studied sites in France and in Japan. Road-crossing speed was estimated by scoring the crossing speed of 20 random pedestrians for each site.

Information Type	France-Strasbourg			
Sites	Train Station	Pont des Corbeaux	Place Broglie	
	48.584474, 7.736135	48.579509, 7.750745	48.584559, 7.748628	
Coordinates				
Lanes	2 * 1	2 * 2	2 * 1	
Mean pedestrian flow per h	667	612	850	
Mean road-crossing speed (m.s <sup>-1</sup> )	0.96 ± 0.05	1.11 ± 0.29	1.01 ± 0.16	
Data collection dates	2–7 July 2014	1–25 October 2014	9 February– March 2015	
Information Type	Japan-Nagoya			
Sites	Train Station	Maruei	Excelco	Osu- Kannon

	35.170824,	35.168638,	35.166891,	35.159316,
Coordinates	136.884328	136.905740	136.907284	136.901697
Lanes	2 * 3	1 * 1	2 * 1	2 * 1
Mean pedestrian flow per h	480	645	869	814
Mean road-crossing speed (m.s <sup>-1</sup> )	1.10 ± 0.22	1.15 ± 0.21	0.98 ± 0.21	1.07 ± 0.18
Data collection dates	5 July 2011	13 June–27 January–5 February 2015		

## 2.2. Data Scoring

Data were scored over a six-day period for each site, for 1 h per day and per site. We observed during working days, hours and weeks to ensure that data excluded movements generated by tourism, festivals, etc. This scoring duration is sufficiently ample to provide a large dataset [12,32,45]. Video cameras were set up in order to score the light color and were placed in locations ensuring that crossing pedestrians were visible at all times. Behavioral sampling was used to score the crossing of pedestrians in one direction only, i.e., that recorded by the camera. Pedestrians were not informed about the purpose of the study. As both cities are touristic, pedestrians are accustomed to seeing tourists taking pictures or videos. We did not observe any difference in the way pedestrians behaved when they saw the camera. We did not take any other equipment such as counters or pocket PC in order to avoid influencing pedestrian behavior. When observation of road-crossing behavior was hampered by a visual obstacle (i.e., a car or a truck in front of the video camera), we removed this behavior and the behaviors occurring immediately before and after it from the analyses. We also removed data in which cyclists were among the pedestrians or where tourists were present. Tourists were easily differentiated from citizens, as they were in large groups accompanied by a guide, were dressed differently from citizens and carried specific equipment (guidebook, map, camera, etc.).

## 2.3. Research Ethics

Our methodological approach solely involved anonymous observations and anonymous data scores. Our protocol followed the ethical guidelines of our institutions (Institut Pluridisciplinaire Hubert Curien, Strasbourg, France and Primate Research Institute, Kyoto University, Japan) and we received ethical approval from these institutions to carry out our study. All data were anonymous, and individuals were given sequential numerical identities according to the time of the road crossing and the arrival/departure order of crossing. Pedestrians had the possibility to be informed about the study by an information medium in their language (Japanese or French). They were also provided with an email address and phone number to contact our institution at a later date if desired. Persons who refused to participate in the study were removed from the data (i.e., we deleted the crossing concerned).

## 2.4. Data Analysis

This study focuses solely on following pedestrians (at the green and the red light) and not the first pedestrians to go, which have been described in a first paper, “Part 1” [46]. This approach was chosen because the two types of decisions (departing first and following) are underpinned by very different processes [9,47,48]. The complete six-hour data set was analyzed for each site. We scored the behaviors of following pedestrians when at least two pedestrians crossed the road simultaneously (i.e., when the time between the two departures was lower than the mean road-crossing time indicated for each site).

All indicated times are in hundredths of a second. We decided to limit our analysis to the crossing behaviors of the first 10 pedestrians at the green or at the red light, mainly due to the difficulty of analyzing the time and order of crossing when more than 10 pedestrians were crossing the road collectively. Similarly, we only analyzed data for pedestrians who were present at the

crossing when the light color changes, or who either decreased their walking speed or stopped as they approached the crossing, as we wanted to assess how specific factors such as the light color, waiting time, and number of pedestrians influenced their decisions to cross the road.

We scored road crossings for 2569 followers, of which 1838 crossed at the green light and 731 at the red one. 902 crossings of followers were scored in France and 1667 in Japan.

For each following pedestrian, we scored the following variables (see [15] for a visual explanation of the different scored variables):

- The light color when crossing (red or green).
- The departure time,  $\Delta T_j$ , i.e., the period between the previous light color change and the moment the pedestrian  $j$  starts crossing the road. This variable is positive for pedestrians crossing after the light (for pedestrians) has turned red but negative for pedestrians crossing before the light turns red.
- The departure latency  $\Delta T_{j,j-1}$ , i.e., the time elapsed between the departure of pedestrian  $j$  and previous pedestrian  $j - 1$ .
- The departure order of pedestrians, where the first pedestrian to leave the kerb is ranked as 1, the second as 2, and so on. Here, we then focused on pedestrians of ranks 2 to  $n$ ,  $n$  being the number of following pedestrians in a crossing event, with a maximum threshold set at  $n = 10$ .
- The gender (male or female).
- The age, estimated at 10-year intervals from 0–9, 10–19, ..., to 80–89. However, given the number of data and the analyses we carried out, it was not possible to analyze the effect of age (per interval) on the decision-making processes. Indeed, the number of pedestrians below 20 and above 60 years old was not sufficient to calculate mean departure latencies per pedestrian crossing rank (less than about 3 data per age category per rank).
- The country (France or Japan).
- The waiting time, i.e., the time between the moment a pedestrian stops at the light and the moment he/she starts crossing the road.

## 2.5. Mathematical Analyses

Survival analysis [33,49] was used to study the distributions of departure latencies for all followers. Survival analysis indicates how the ratio of observations decreases from 1 (all observations/data) to 0 (none) according to a response variable. First, curve estimation tests identified which type of function was followed by these distributions, namely linear (meaning that the probability of crossing is time dependent), exponential (the probability of crossing is constant over time) or sigmoid (the probability of crossing depends on a time threshold that is directly correlated to the response variable) (see [11,22,38]).

When trying to identify the process underlying following behaviors in pedestrians, we had two hypotheses, namely the independence hypothesis and the mimetism hypothesis [21,22,50]. Explanations of these two different hypotheses are given just below, but they can only be considered once we have evaluated whether the intrinsic possibility of crossing for each pedestrian is constant per time unit. This intrinsic motivation is given by studying the distribution of departure latencies for the first pedestrian who crosses. As this distribution corresponded to an exponential distribution (see Supplementary material and Table 2), the departure probability of a first pedestrian to cross  $\psi_{01}$  is the log gradient of this exponential distribution, that is, the inverse function of the mean departure latency  $\Delta T_1$  for the first departing individual [38,50,51]:

$$\psi_{01} = \sum_{i=1}^n \lambda_i, \quad (1)$$

where  $\psi_{01}$  is the probability of seeing a first pedestrian crossing the road. We based this probability on the departure latencies for a pedestrian crossing at the green light, but for pedestrians crossing at the red light, we based this probability on waiting time.

$n$  is the numbers of waiting individuals, here  $n = N$  being the maximum number of pedestrians we analyzed for each crossing, meaning 10.

$\lambda_i$  is the probability of individual  $i$  being the first person to cross. Here, we analyzed this probability according to the gender and the country (Table 2).

The probability  $\lambda_i$  of the individual  $i$  to be the first to cross is, therefore:

$$\lambda_i = \frac{\psi_{01}}{n} = \frac{\psi_{01}}{10}. \tag{2}$$

and,

$$\psi_{01} = \frac{1}{\Delta\tau_1}. \tag{3}$$

In a mimetic process where the departure probability is proportional to the number of pedestrians already moving  $j$ , the probability  $\psi_j$  per unit time that one of the  $n$  waiting agents became the  $j$ th following pedestrian is:

$$\psi_j = (\lambda + C(j - 1))n, \tag{4}$$

where  $C$  is the mimetic coefficient per individual. Here, we analyzed this probability depending on the gender and the country (results are given in Table 2).

The departure latency  $\Delta T_{j,j-1}$  of the follower  $j$  was:

$$\Delta T_{j,j-1} = \frac{1}{(\lambda + C(j - 1))(n - (j - 1))} \tag{5}$$

or

$$\frac{1}{\Delta T_{j,j-1}} = \psi_{j,j-1} = (\lambda + C(j - 1))(n - (j - 1)) = (\lambda - C)(n + 1) + j(2C + Cn - \lambda) - Cj^2. \tag{6}$$

This is a parabolic law  $z + wj + yj^2$  where,

$$z = (\lambda - C)(n + 1),$$

$$w = (2C + Cn - \lambda),$$

$$y = C.$$

Fitting the distribution of departure probability per pedestrian departure rank with a parabolic curve allows us to calculate a range of values for the mimetic coefficient  $C$  (See Supplementary Material for calculation details and “Results” for the statistical values of parabolic curve estimation tests).

**Table 2.** Values of departure probability for the first pedestrian to leave the kerb  $\psi_{01}$ , intrinsic departure probability of following pedestrians  $\lambda$ , mimetic coefficient  $C$  and crossing rule according to the light color, country and gender.  $D$  indicates the dataset (number of observed pedestrians) for each category. Range of calculated  $C$  is given by resolving by parabolic Equation (6). The best mimetic coefficient  $C$  is the one obtained after implementing the range of calculated  $C$  in the model and confronting simulated and observed data. The crossing rule is considered to be that the probability to join depends on the number of already crossing pedestrians (Equation (4)), or is calculated as the number of crossing pedestrians divided by the number of waiting ones (Equation (7)). Values are small because the time step we used is the hundredth of second. To obtain an idea in seconds, multiply by 100.

Light Color	Country	Gender	$D$	$\psi_{01}$	Range of Calculated $C$	Best $C$	Crossing Rule
Green	France	Man	98	0.013	0.0001–0.0026	0.0006	Crossing
Green	France	Woman	130	0.013	0.0005–0.0031	0.0012	Crossing
Green	Japan	Man	638	0.016	0.0006–0.0026	0.0016	Crossing
Green	Japan	Woman	972	0.017	0.0004–0.0031	0.00135	Crossing
Red	France	Man	290	0.0008	0.00047–0.0006	0.0005	Crossing

Red	France	Woman	384	0.0006	0.00083–0.0002	0.0008	Crossing
Red	Japan	Man	35	0.0003	0.0056–0.011	0.005	Crossing/Waiting
Red	Japan	Woman	22	0.0003	0.0021–0.0032	0.005	Crossing/Waiting

## 2.6. Modelling

A modelling approach using multi-agent based models was used to determine whether pedestrians cross according to a mimetic process (mimetism hypothesis) or independently of external influence (independence hypothesis), and to ensure the precise determining of the mimetic coefficient for each gender and country according to the range of values provided by the survival analyses.

We implemented the distribution of the number of waiting pedestrians  $N$  in the model. There is no difference in this distribution between the Japanese and French sites (Mann–Whitney test =  $W = 764,990$ ,  $p = 0.464$ ,  $M_{\text{France}} = 12.87 \pm 6.14$ ,  $M_{\text{Japan}} = 12.38 \pm 4.59$ ).

At the start of a simulation, all agents ( $N$ ) were at the kerb on the same side of the road (we consider only one side) and had to move to the opposite sidewalk. We implemented the departure probability  $\lambda_i$  of each agent, according to his or her gender and his or her country (see Table 2). Pedestrians of the same gender and from the same country all have the same probabilities. The departure probability of the first departing individual, the only individual whose decision to move would not be influenced by the other group members, was identical for the two versions of the model tested here, i.e., the independence hypothesis and the mimetic process hypothesis. We assumed in the model that all agents are aware of the state (waiting or crossing) of all other agents at any given time.

### i. Independence hypothesis

The first hypothesis assumed that individuals were independent: the departure probability of a pedestrian was not influenced by the departure of other pedestrians. As a consequence, the individual departure probability remained constant, whatever the rank of the crossing pedestrian.

Under this hypothesis, the probability that one of the  $n$  waiting agents (e.g. individual  $i$ ) would become a crossing pedestrian  $j$  per unit time was  $\lambda_i$ . According to Equations (1) and (2), the departure latency of the crossing pedestrian  $j$  was the inverse function of the sum of the  $\lambda_i$  of  $n$  waiting agents:

$$\Delta T_{j,j-1} = \frac{1}{\sum_{i=1}^N \lambda_i}. \quad (7)$$

In our case, the probabilities were identical, with  $\lambda_1 = \dots = \lambda_N = \lambda$  for pedestrians of same country and same gender.

### ii. Mimetic process hypothesis

The second hypothesis specified that pedestrians would be influenced by seeing others crossing, thanks to a mimetic process. To test this hypothesis, we added a mimetic coefficient  $C$  to the above version of the model (independence hypothesis). The probability per unit time that one of the  $n$  waiting agents would become the crossing pedestrian  $j$  under the anonymous mimetic hypothesis was obtained from Equation (4) and its departure latency obtained from Equation (5). The range of different mimetic coefficients found in Table 2 was tested. In this version of the model and according to Equation (4), waiting pedestrians had the same probability of crossing but would now be differentiated according to their gender and their country. In Equation (4), the probability that an individual will cross is only influenced by the number of already crossing individuals  $j$ . This is the most parsimonious hypothesis, which also has the lowest number of parameters [21,44,52]. However, when the number of individuals already moving is not enough to explain the probability of departures, it is necessary to take into account both the number of moving individuals  $j$  (here, the pedestrians that have already started crossing) and the number of individuals that are not moving  $n$  (here, the waiting pedestrians) [44]. This has been applied in data analysis for collective movements of sheep [43].

In this last case, the probability to see a pedestrian crossing the road is, therefore:

$$\psi_j = \left( \lambda + C \left( \frac{j-1}{n} \right) \right) n = (\lambda n + C(j-1)). \quad (8)$$

If the first rule (Equation (4)) did not fit with our observed data, we tested the second rule (Equation (7)).

The different versions of the model, corresponding to each hypothesis, were implemented in Netlogo 6.0.6 [36,53]. The models we developed are modified versions of an existing model from [38] and is adapted to Netlogo 6.0.6. Models can be found in the Supplementary material.

The model is stochastic: a number between 0 and 1 was randomly attributed to each waiting pedestrian at each time step (a hundredth of second). If this number was lower than the theoretical departure probability of each agent, the pedestrian would start to cross; if this number was higher than the theoretical departure probability, the agent did not move and continued to wait. The country, the gender, the pedestrian departure rank and the departure latency of each crossing pedestrian agent were scored for each simulated crossing event. To be consistent with the experimental situation, we stopped a simulation after a specific time threshold when no agent followed the departure of the first or last crossing pedestrian. This threshold depends not only on the light color but also on the country. When crossing at the green light, the time threshold is defined as the time the light remains green; 1 min 45 s is the maximum green light time between the sites and 23 s the minimum time. Knowing that the maximum departure time of a following pedestrian was 9.76 s, this threshold is more than sufficient to avoid committing any errors. At the red light, French and Japanese pedestrians showed two different behaviors (see Part 1, [46]). Indeed, when crossing illegally, French pedestrians crossed throughout the duration of the red light (about 80 s) whilst the Japanese only crossed when the pedestrian light was close to changing from red to green (about 4 s). We then set these thresholds for French (80 s) and Japanese (4 s) pedestrians, respectively. We stopped the simulation when all the  $N$  agents had crossed the road or when the time threshold had been reached, but considered a maximum of 10 pedestrians for each crossing event in order to compare simulations with our observed data. We set the number of simulations to 1000 for each hypothesis and for each set of tested parameters (Light color \* Country \* gender \* hypotheses, with the range of mimetic coefficients values for the mimetism hypothesis), or a total of 70,000 simulations.

## 2.7. Statistical Analyses

In order to know whether departure time survival distributions follow exponential law (see Supplementary Material) and the distributions of departure latencies according to pedestrian crossing rank follow parabolic law, linear regression was used to compare theoretical data to observed data with adjusted  $R^2$ . The best fitting distribution were chosen according to F-statistics. The Kolmogorov–Smirnov test was used to compare the distribution of simulated data to observed data. This test was revealed to be the best means to compare two distributions. In a Kolmogorov–Smirnov test, the higher the p-value, the lower the D-statistics, and the better the fitting. Analyses were performed in R 3.3.2, with  $\alpha$  set at 0.05.

## 3. Results

### 3.1. Analyses of Departure Latencies According to Pedestrian Departure Rank

The survival curves for departure latencies of following pedestrians at the green light ( $y = 0.7736 * \exp^{-0.029x}$ ,  $R^2 = 0.98$ ,  $F = 88,800$ ,  $p < 0.00001$ ) and waiting time at the red light both fit with exponential distributions ( $y = 1.0606 * \exp^{-0.0005x}$ ,  $R^2 = 0.97$ ,  $F = 23,540$ ,  $p < 0.00001$ ). The probability that a pedestrian will follow another at the green light is 0.029, a value that is about 22 and 17.5 times higher than the probability to be the first to leave the kerb in France and Japan, respectively (see Table 3). The probability that a pedestrian will follow another at the red light is 0.0005, i.e., about 7 and 16 times higher than the probability to be the first to leave the kerb in France and Japan, respectively. Several initial conclusions can be drawn from this result. First, the substantially higher probability indicates a mimetic process. Second, this mimetic process seems to differ according to the light color and the

country: French citizens show less mimetism at the red light compared to the green light, but more cases of first pedestrians crossing at the red light in France compared to Japan. However, the same mimetic process is observed for Japanese pedestrians crossing at the red light and the green light, with much fewer pedestrians crossing first at the red light.

We then analyzed whether the departure latencies according to the pedestrian departure rank followed a parabolic curve, which indicates that a mimetic process underlies the decision to cross. Results are given in Table 3. Whilst mimetic process is seen to govern the behavior of Japanese pedestrians, this mimetism does not seem to be as consistent in the behavior of French pedestrians.

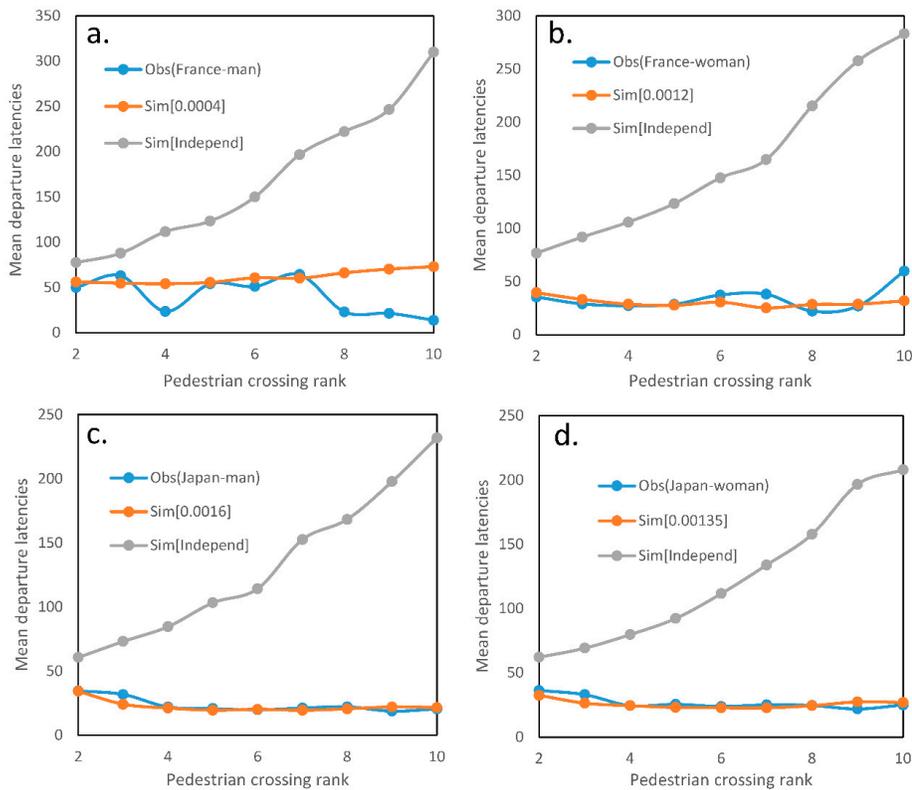
**Table 3.** Statistical values of parabolic curve estimation test with the departure latencies according to the pedestrian departure rank, for each light color, gender and country.

Country	Gender	Green Light			Red Light		
		df	F	p	df	F	p
France	Man	7	2.79	0.139	7	15.37	0.006
France	Woman	7	4.01	0.08	7	1.448	0.268
Japan	Man	7	42.02	0.0003	7	5.93	0.026
Japan	Woman	7	40.1	0.0004	4	7.81	0.039

### 3.2. Comparison between Observed Crossing and Simulated Crossing

The modelling of crossing behaviors reveals whether an individual bases their crossing decision purely on their own motivation (independence hypothesis) or if this decision solely relies on mimetic behaviors (mimetic process hypothesis).

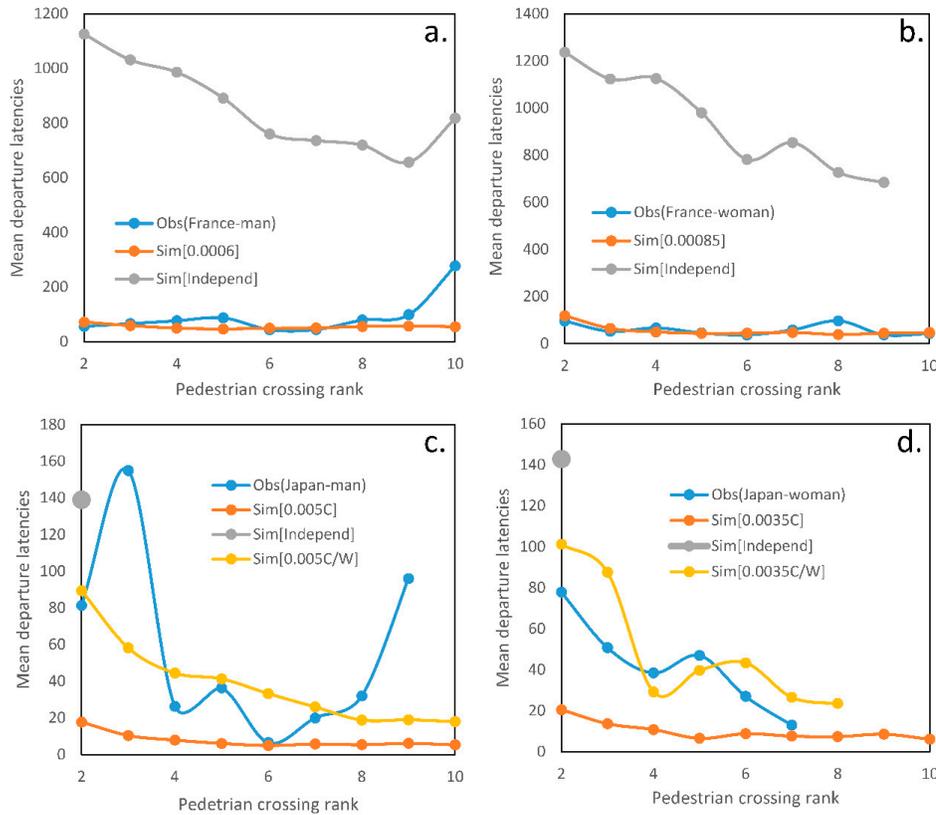
Crossings at the green light: only the number of pedestrians that were already crossing was used, and this was sufficient to explain the distribution of observed departure latencies (Table 3). Results are illustrated in Figure 1. No confirmation of the independence hypothesis was found for either country or gender, and results from simulations were consistently different from observational data ( $p \leq 0.00001$ ,  $D = 1$ ). Values of mimetic coefficients for French men ranged from 0.0005 to 0.001 ( $p \geq 0.126$ ,  $D \leq 0.556$ ) and did not provide different distributions of simulated departure latencies compared to observed data with a best mimetic coefficient of 0.0006 ( $p = 0.352$ ,  $D = 0.444$ , Figure 1a). Concerning French women, distributions of simulated departure latencies with mimetic coefficients ranging from 0.0009 to 0.0015 were not statistically different from the observed data ( $p \geq 0.126$ ,  $D \leq 0.556$ ) with a best mimetic coefficient of 0.00012 ( $p = 0.730$ ,  $D = 0.333$ , Figure 1b). Simulations for Japanese men showed similar distributions of departure latencies to observational data, with mimetic coefficients going from 0.0015 to 0.0019 ( $p \geq 0.126$ ,  $D \leq 0.556$ ) with a best value of 0.0016 ( $p = 0.989$ ,  $D = 0.222$ , Figure 1c). Finally, simulations of crossings for Japanese women are not different from observed data, with mimetic coefficients ranging from 0.0012 to 0.0015 ( $p \geq 0.126$ ,  $D \leq 0.556$ ) and a best mimetic coefficient of 0.00135 ( $p = 0.989$ ,  $D = 0.222$ , Figure 1d).



**Figure 1.** Distribution of mean departure latencies (hundredth of second) at the green light according to pedestrian crossing rank for observed data (blue), simulated data of the independence hypothesis (grey) and simulated data for the best mimetic coefficient (orange). (a) For French men; (b) for French women; (c) for Japanese men; and (d) for Japanese women.

Crossings at the red light: although the number of pedestrians already crossing was sufficient to explain the departure latencies of French sites, the departure probability had to be modelled as dependent on  $\frac{\text{the number of already crossing pedestrians}}{\text{the number of waiting pedestrians}}$  in order to fit the simulations to observation data in Japanese sites (Table 3). While all the waiting agents crossed in the green-light model, some simulations for the red-light model were different, with the number of crossing agents sometimes stopping before the threshold of 10 was reached. This result was important to explain the data. The independence hypothesis was not confirmed, either because the distribution of simulated departure latencies was different from the observed data ( $p \leq 0.00001$ ,  $D = 1$  for French pedestrians), or because the simulations only reached one follower in Japanese pedestrians, which is far from the result we obtained with observations (nine followers for Japanese men and seven followers for Japanese women). For French men, departure latencies of mimetic coefficients [0.0006; 0.0005] were not statistically different from observed departure latencies ( $p \geq 0.126$ ,  $D \leq 0.556$ ; 0.0006 being the best mimetic coefficient:  $p = 0.352$ ,  $D = 0.444$ , Figure 2a). Concerning French women, distributions of simulated departure latencies with mimetic coefficients going from 0.0006 to 0.00085 are not statistically different from observed data ( $p \geq 0.126$ ,  $D \leq 0.556$ ), with a best mimetic coefficient of 0.00085 ( $p = 0.730$ ,  $D = 0.333$ , Figure 2b). As previously noted, the number of pedestrians already crossing was not enough as a single process in our model to explain the departure latencies of Japanese pedestrians ( $p < 0.028$ ,  $D > 0.722$ ). However, the distributions of departure latencies in simulations that include the number of pedestrians already crossing and the number of waiting pedestrians (Table 3) are not different from the observed departure latencies ( $p \geq 0.108$ ,  $D \leq 0.667$ ),

with 0.005 as the best mimetic coefficient for Japanese men ( $p = 0.860$ ,  $D = 0.264$ , Figure 2c) and 0.0035 for Japanese women ( $p = 0.872$ ,  $D = 0.285$ , Figure 2d).



**Figure 2.** Distribution of mean departure latencies (hundredth of second) at the red light according to pedestrian crossing rank for observed data (blue), simulated data of the independence hypothesis (grey) and simulated data for the best mimetic coefficient (orange, and solely involves pedestrians that are already crossing for Japanese sites). Finally, yellow curves indicate the ratio of crossing/waiting pedestrians for the Japanese sites. (a) For French men; (b) for French women; (c) for Japanese men; and (d) for Japanese women. For (c) and (d) the independence hypothesis curve is only represented by a dot at the pedestrian crossing rank 2. We never observed a rank 3 for the independent hypothesis in Japanese pedestrians.

Finally, for the rate of illegal crossings in simulations, the mimetism hypothesis—taking the number of crossing pedestrians into account for French sites and the number of crossing and waiting pedestrians into account for the Japanese sites—is the only hypothesis that provides similar results to observed rule breaking (Table 4). The simulated rate is the same as for the observed rate in all groups except French women, where we have 12% less rule breaking in observation data.

**Table 4.** Rate of illegal crossings for observations and simulations. C means model with only crossing pedestrian as rule, C/W means number of crossing pedestrians divided by number of waiting pedestrians rule.

Hypotheses	France		Japan	
	men	women	men	women
Observed	0.46	0.38	0.02	0.02
Independence hyp.	0.24	0.19	0.0006	0.001

Mimetic hyp. (C)	0.50	0.50	0.12	0.04
Mimetic hyp. (C/W)	NA	NA	0.02	0.02

#### 4. Discussion

This study showed that modelling is useful to test different hypotheses and find the processes underlying decision-making when pedestrians cross the road. Even if modelling remains an abstraction of reality, it identified exactly which parameters are important to fit the simulations to observations, and revealed in this study the differences between the Japanese and the French and between women and men in their decision to follow and cross at the green and red lights. Finally, the model highlighted not only similar departure latencies but also a similar number of crossing pedestrians and a similar rate of illegal crossings between simulated and observed data.

In order to fit the simulated departure latencies to the observed departure latencies, we had to include different parameters in the model, certain of which varied according to the gender and the country. It is important to make the agent-based model reflect the site variables as closely as possible to be sure that simulation results are comparable to empirical observation data [36,54,55]. This avoids false positives or false negatives that can occur through incorrect model parameters. This matters particularly when studying such risky, and even lethal, behaviors such as crossing illegally. First, the number of waiting pedestrians was implemented. The distribution was identical for French and Japanese sites, confirming that our sites were comparable (as Table 1 shows), and that the number of waiting pedestrians was a necessary variable in the model to fit the simulated number of crossing pedestrians to the observed data. We had to model the threshold time rules of crossing that we observed according to the light color but also according to the country. This threshold time necessity in the model confirms that decision making differs according to the light color but also according to the country, with no crossing at the red light for Japanese pedestrians until the light for vehicles turns orange (see Part 1, [46]). This effect of traffic light sequences has already been described in different studies [20,56]. Finally, we had to implement the intrinsic probability to cross and the mimetic coefficient to differentiate between the independence hypothesis and the mimetic process hypothesis, respectively.

The intrinsic probability, meaning our own motivation without being influenced by others, was revealed to differ according to the parameters we studied. First, the intrinsic probability is higher at the green light than at the red light, which is both understandable and reassuring. This shows that the probability of deciding to cross based solely on our personal information is lower at the red light compared to the green light. As we found in a previous study ([15,46]), French pedestrians cross at the red light more often than Japanese pedestrians. The current study finds that the intrinsic probability to cross at the red light is higher at French sites and conversely, the intrinsic probability to cross at the green light is higher at Japanese sites. However, this difference is not just related to the rate of illegal crossings, but is also explained by departure times. Japanese pedestrians start crossing sooner than their French counterparts at the green light, probably because they are more attentive to the light change, meaning that they trust their personal information more than they trust social information [12,15]. This may also be because Nagoya is a more time-pressured city than Strasbourg. In stark contrast, the French cross more and much faster at the red light. Intrinsic probability is, therefore, representative of individual risk-taking levels [10,30,32]. We found a slight difference between French men and women crossing at the red light: here, women seem to have a lower probability of crossing based solely on their personal information at the red light, which confirms previous studies [11,26,57]. However, this intrinsic probability alone was not enough to explain the following behaviors of pedestrians, meaning that a mimetic or amplification process underlies most of the crossing decisions made by following pedestrians.

The implementation of a mimetic process in this model makes it possible to fit simulated data to empirical data and to explain them: the higher the number of individuals crossing, whatever the light color, the higher the probability is that other individuals will cross. This copying behavior, referred to as “sheep” or “herd” behavior, is well known in human beings in domains such as financial

markets, fashion, purchasing or crowd behavior [5,58–62]. This process seems to be deeply rooted in human behavior due to their gregariousness and sociality, and can be observed in many other social animal species [63–65]. Mimetism is lower for French pedestrians crossing at the red light, but they only take the number of crossing pedestrians into consideration when making their decision. By contrast, Japanese pedestrians show greater levels of mimetism at the red light, but they take both the number of pedestrians already crossing and the number of waiting pedestrians into consideration when making their decision to cross. These results confirm the more individualistic nature of French society and the more collectivist and conformist behavior in Japan [11,66,67]. Indeed, the different processes we observed in this study show that the Japanese consider the behavior of other pedestrians: the Japanese tendency to take the behavior of others into account could be explained by their fear of being criticized (social credibility), which is greater than the fear of being fined (risk exposure) [10,24]. Taking into account the number of crossing and waiting pedestrians is a “following the majority” rule that illustrates how humans adapt their individual behavior to that of others. Indeed, this rule has been well described in animal species [44,63,64] including strongly gregarious species such as sheep [43]. The lower mimetic coefficients suggest that men are more individualistic than women, but this is only confirmed in France, whilst Japanese men and women behave in the same way.

## 5. Conclusions

This study highlighted that only two human variables, namely intrinsic probability and in particular mimetism, were sufficient and adequate to explain the departure latencies of pedestrians but also the rate of rule breaking. Matching values for each gender and each country led to a significant fitting between simulated and observed data. This leads us to conclude that when crossing the road, at least at signalized crossings, human beings behave like sheep, and that a high rate of accidents with cars might be due to this herding behavior and the misuse of social information. It would be interesting first to assess whether accidents happen more in pedestrian groups or with alone/isolated individuals. In a second point, we need to identify whether these following pedestrians are aware of the risk or not: do they follow thinking that the signal is green or do they follow knowing that the light is red? More research is necessary on these behaviors. With the part 1 [46], we identified two types of pedestrians crossing illegally: the first ones to step off the kerb and those following (this current study). Part 1 [55] gave solutions to avoid intentional illegal crossings, but we need also to prevent pedestrians followings their conspecifics but not being aware of the red light and, as a consequence, of the risks. A sound signal produced when pedestrians cross at the red light could be a solution to stop the first pedestrian from crossing but also to warn other pedestrians that are ready to follow that this is not the right time. The signal could be only a sound but China experimented with some water spray at illegal crossings [68]. These signs need to be tested in order to assess their efficiency but also to determine whether they do not lead to unexpected risky behaviors (i.e., surprise and stop in the middle of the road). To conclude, these studies about the decision-making processes of pedestrians during road-crossing are useful tools to conceive new safety and public education solutions in transport research.

**Supplementary Materials:** The following are available online at [www.mdpi.com/xxx/s1](http://www.mdpi.com/xxx/s1), calculation details and “Results” for the statistical values of parabolic curve estimation tests. Models are also given.

**Author Contributions:** Conceptualization, M.P., J.-L.D. and C.S.; Methodology, M.P., J.-L.D. and C.S.; Formal Analysis, C.S.; Investigation, M.P. and C.S.; Writing—Original Draft Preparation, C.S.; Writing—Review and Editing, M.P., J.-L.D. and C.S.; Funding Acquisition, C.S.

**Funding:** This research was funded by the French Homeland Ministry (Ref.CNRS190529). CS is member of the French Academic Institute (IUF).

**Acknowledgments:** We thank Caroline Bellut, Elise Debergue, Charlotte Gauvin, Anne Jeanneret, Thibault Leclere, Lucie Nicolas, Florence Pontier and Diorme Zausa for their help collecting data. We are grateful to Kunio Watanabe and Hanya Goro (Primate Research Institute, Kyoto University) for their help in authorizing data collection.

**Conflicts of Interest:** The authors declare no conflict of interest.

## References

1. Twain, M. *What is Man? And Other Philosophical Writings*; University of California Press, London, UK, 1973; p. 62.
2. Haghani, M.; Sarvi, M. Following the crowd or avoiding it? Empirical investigation of imitative behaviour in emergency escape of human crowds. *Anim. Behav.* **2017**, *124*, 47–56.
3. Helbing, D.; Farkas, I.; Vicsek, T. Simulating dynamical features of escape panic. *Nature* **2000**, *407*, 487–490.
4. Helbing, D.; Schweitzer, F.; Keltsch, J.; Molnár, P. Active walker model for the formation of human and animal trail systems. *Phys. Rev. E* **1997**, *56*, 2527.
5. Mawson, A.R. Understanding Mass Panic and Other Collective Responses to Threat and Disaster. *Psychiatry Interpers. Biol. Process.* **2005**, *68*, 95–113.
6. Dyer, J.R.G.; Johansson, A.; Helbing, D.; Couzin, I.D.; Krause, J. Leadership, consensus decision making and collective behaviour in humans. *Philos. Trans. R. Soc. B Biol. Sci.* **2009**, *364*, 781–789.
7. Kurvers, R.H.J.M.; Wolf, M.; Naguib, M.; Krause, J. Self-organized flexible leadership promotes collective intelligence in human groups. *Open Sci.* **2015**, *2*, 150222.
8. King, A.J.; Johnson, D.D.P.; Van Vugt, M. The Origins and Evolution of Leadership. *Curr. Biol.* **2009**, *19*, R911–R916.
9. Van Vugt, M.; Hogan, R.; Kaiser, R.B. Leadership, followership, and evolution: Some lessons from the past. *Am. Psychol.* **2008**, *63*, 182–196.
10. Faralla, V.; Innocenti, A.; Venturini, E. *Risk Taking and Social Exposure*; Social Science Research Network: Rochester, NY, USA, 2013.
11. Sueur, C.; Class, B.; Hamm, C.; Meyer, X.; Pelé, M. Different risk thresholds in pedestrian road crossing behaviour: A comparison of French and Japanese approaches. *Accid. Anal. Prev.* **2013**, *58*, 59–63.
12. Faria, J.J.; Krause, S.; Krause, J. Collective behavior in road crossing pedestrians: The role of social information. *Behav. Ecol.* **2010**, *21*, 1236–1242.
13. Gould, J.P. Risk, stochastic preference, and the value of information. *J. Econ. Theory* **1974**, *8*, 64–84.
14. Lipovac, K.; Vujanic, M.; Maric, B.; Nestic, M. The influence of a pedestrian countdown display on pedestrian behavior at signalized pedestrian crossings. *Transp. Res. Part F Traffic Psychol. Behav.* **2013**, *20*, 121–134.
15. Pelé, M.; Bellut, C.; Debergue, E.; Gauvin, C.; Jeanneret, A.; Leclere, T.; Nicolas, L.; Pontier, F.; Zausa, D.; Sueur, C. Cultural influence of social information use in pedestrian road-crossing behaviours. *Open Sci.* **2017**, *4*, 160739.
16. Boos, M.; Pritz, J.; Lange, S.; Belz, M. Leadership in Moving Human Groups. *PLoS Comput. Biol.* **2014**, *10*, e1003541.
17. Faria, J.J.; Dyer, J.R.G.; Tosh, C.R.; Krause, J. Leadership and social information use in human crowds. *Anim. Behav.* **2010**, *79*, 895–901.
18. Katsikopoulos, K.V.; King, A.J. Swarm Intelligence in Animal Groups: When Can a Collective Out-Perform an Expert? *PLoS ONE* **2010**, *5*, e15505.
19. Lewis, J.; Wartzok, D.; Heithaus, M. Individuals as information sources: Could followers benefit from leaders' knowledge? *Behaviour* **2013**, *150*, 635–657.
20. Rosenbloom, T. Crossing at a red light: Behaviour of individuals and groups. *Transp. Res. Part F Traffic Psychol. Behav.* **2009**, *12*, 389–394.
21. Camazine, S.; Deneubourg, J.-L.; Franks, N.R.; Sneyd, J.; Theraula, G.; Bonabeau, E. *Self-Organization in Biological Systems*; Princeton University Press, 2003; ISBN 978-0-691-11624-2.
22. Sueur, C.; Deneubourg, J.-L. Self-Organization in Primates: Understanding the Rules Underlying Collective Movements. *Int. J. Primatol.* **2011**, *32*, 1413–1432.
23. Hung, A.A.; Plott, C.R. Information cascades: Replication and an extension to majority rule and conformity-rewarding institutions. *Am. Econ. Rev.* **2001**, *91*, 1508–1520.
24. Osman, L.M. Conformity or compliance? A study of sex differences in pedestrian behaviour. *Br. J. Soc. Psychol.* **1982**, *21*, 19–21.

25. Powell, M.; Ansic, D. Gender differences in risk behaviour in financial decision-making: An experimental analysis. *J. Econ. Psychol.* **1997**, *18*, 605–628.
26. Tom, A.; Granié, M.-A. Gender differences in pedestrian rule compliance and visual search at signalized and unsignalized crossroads. *Accid. Anal. Prev.* **2011**, *43*, 1794–1801.
27. Centola, D.; Baronchelli, A. The spontaneous emergence of conventions: An experimental study of cultural evolution. *Proc. Natl. Acad. Sci. USA* **2015**, *7*, doi:10.1073/pnas.1418838112.
28. Ito, Y. Socio-cultural backgrounds of Japanese interpersonal communication style. *Civilisations* **1989**, *39*, 101–128.
29. Lee, S.M.; Lim, S.; Pathak, R.D. Culture and entrepreneurial orientation: A multi-country study. *Int. Entrep. Manag. J.* **2011**, *7*, 1–15.
30. Mihet, R. *Effects of Culture on Firm Risk-Taking: A Cross-Country and Cross-Industry Analysis*; International Monetary Fund: Washington, DC, USA, 2012.
31. Czaczkes, T.J.; Czaczkes, B.; Iglhaut, C.; Heinze, J. Composite collective decision-making. *Proc. R. Soc. B* **2015**, *282*, doi:10.1098/rspb.2014.2723723.
32. Pelé, M.; Sueur, C. Decision-making theories: Linking the disparate research areas of individual and collective cognition. *Anim. Cogn.* **2013**, *16*, 543–556.
33. Miller, R.G.; Gong, G.; Muñoz, A. *Survival Analysis*; John Wiley & Sons: Hoboken, NJ, USA, 1981.
34. Aoyagi, S.; Hayashi, R.; Nagai, M. *Modeling of Pedestrian Behavior in Crossing Urban Road for Risk Prediction Driving Assistance System*; SAE International: Warrendale, PA, USA, 2011.
35. Papadimitriou, E. Theory and models of pedestrian crossing behaviour along urban trips. *Transp. Res. Part F Traffic Psychol. Behav.* **2012**, *15*, 75–94.
36. Sueur, C.; Perony, N.; Amblard, F.; Guillaume, J.-L. Modélisation des réseaux sociaux. In *Analyse des Réseaux Sociaux Appliquée à l'Ethologie et à l'Ecologie*; Editions Matériologiques: Paris, France, 2015; pp. 423–444.
37. Epstein, J.M. *Why Model?* Princeton University Press: Princeton, NJ, USA, 2008.
38. Sueur, C.; Petit, O.; Deneubourg, J. Selective mimetism at departure in collective movements of Macaca tonkeana: An experimental and theoretical approach. *Anim. Behav.* **2009**, *78*, 1087–1095.
39. Brewer, M.; Fitzpatrick, K.; Whitacre, J.; Lord, D. Exploration of Pedestrian Gap-Acceptance Behavior at Selected Locations. *Transp. Res. Rec. J. Transp. Res. Board* **2006**, *1982*, 132–140.
40. Jiang, N.; Shi, M.; Xiao, Y.; Shi, K.; Watson, B. *Factors Affecting Pedestrian Crossing Behaviors at Signalized Crosswalks in Urban Areas in Beijing and Singapore*; ASCE Library: Reston, VA, USA, 2011; pp. 1090–1097.
41. King, M.J.; Soole, D.; Ghafourian, A. Illegal pedestrian crossing at signalised intersections: Incidence and relative risk. *Accid. Anal. Prev.* **2009**, *41*, 485–490.
42. Sueur, C. Viability of decision-making systems in human and animal groups. *J. Theor. Biol.* **2012**, *306*, 93–103.
43. Pillot, M.-H.; Gautrais, J.; Arrufat, P.; Couzin, I.D.; Bon, R.; Deneubourg, J.-L. Scalable Rules for Coherent Group Motion in a Gregarious Vertebrate. *PLoS ONE* **2011**, *6*, e14487.
44. Pillot, M.-H.; Deneubourg, J.-L. Collective movements, initiation and stops: Diversity of situations and law of parsimony. *Behav. Process.* **2010**, *84*, 657–661.
45. Yannis, G.; Papadimitriou, E.; Theofilatos, A. Pedestrian gap acceptance for mid-block street crossing. *Transp. Plan. Technol.* **2013**, *36*, 450–462.
46. Pelé, M.; Deneubourg, J.-L.; Sueur, C. Decision-Making Processes Underlying Pedestrian Behaviors at Signalized Crossing: Part the First to Step off the Kerb. *Safety* **2019**, *5*, 79.
47. Fernandez, A.A.; Deneubourg, J.L. On following behaviour as a mechanism for collective movement. *J. Theor. Biol.* **2011**, *284*, 7–15.
48. King, A.J.; Cowlshaw, G. Leaders, followers and group decision-making. *Integr. Commun. Biol.* **2009**, *2*, 147–150.
49. Klein, J.P.; Goel, P.K. *Survival Analysis: State of the Art*; Springer: Berlin/Heidelberg, Germany, 1992; ISBN 978-0-7923-1634-3.
50. Gautrais, J.; Michélena, P.; Sibbald, A.; Bon, R.; Deneubourg, J.-L. Allelomimetic synchronization in Merino sheep. *Anim. Behav.* **2007**, *74*, 1443–1454.
51. Amé, J.-M.; Halloy, J.; Rivault, C.; Detrain, C.; Deneubourg, J.L. Collegial decision making based on social amplification leads to optimal group formation. *Proc. Natl. Acad. Sci. USA* **2006**, *103*, 5835–5840.

52. Sober, E. The Principle of Parsimony. *Br. J. Philos. Sci.* **1981**, *32*, 145–156.
53. Sklar, E. NetLogo, a Multi-agent Simulation Environment. *Artif. Life* **2007**, *13*, 303–311.
54. Cioffi-Revilla, C. Invariance and universality in social agent-based simulations. *Proc. Natl. Acad. Sci. USA* **2002**, *99*, 7314–7316.
55. Wilensky, U.; Rand, W. *An Introduction to Agent-Based Modeling: Modeling Natural, Social, and Engineered Complex Systems with NetLogo*; MIT Press: Cambridge, MA, USA, 2015; ISBN 978-0-262-73189-8.
56. Lachapelle, U.; Cloutier, M.-S. On the complexity of finishing a crossing on time: Elderly pedestrians, timing and cycling infrastructure. *Transp. Res. Part Policy Pract.* **2017**, *96*, 54–63.
57. Holland, C.; Hill, R. The effect of age, gender and driver status on pedestrians' intentions to cross the road in risky situations. *Accid. Anal. Prev.* **2007**, *39*, 224–237.
58. Banerjee, A.V. A Simple Model of Herd Behavior. *Q. J. Econ.* **1992**, *107*, 797–817.
59. Bikhchandani, S.; Hirshleifer, D.; Welch, I. A Theory of Fads, Fashion, Custom, and Cultural Change as Informational Cascades. *J. Polit. Econ.* **1992**, *100*, 992–1026.
60. Chen, Y.-F. Herd behavior in purchasing books online. *Comput. Hum. Behav.* **2008**, *24*, 1977–1992.
61. Hirshleifer, D.; Hong Teoh, S. Herd Behaviour and Cascading in Capital Markets: A Review and Synthesis. *Eur. Financ. Manag.* **2003**, *9*, 25–66.
62. Raafat, R.M.; Chater, N.; Frith, C. Herding in humans. *Trends Cogn. Sci.* **2009**, *13*, 420–428.
63. Couzin, I.D.; Krause, J. *Self-Organization and Collective Behavior in Vertebrates*; Academic Press: Cambridge, MA, USA, 2003; Volume 32, pp. 1–75, ISBN 0065-3454.
64. King, A.J.; Sueur, C. Where Next? Group Coordination and Collective Decision Making by Primates. *Int. J. Primatol.* **2011**, *32*, 1245–1267.
65. Krause, J.; Ruxton, G.D. *Living in Groups*; Oxford University Press: Oxford, UK, 2002; ISBN 978-0-19-850818-2.
66. Benedict, R. *The Chrysanthemum and the Sword: Patterns of Japanese Culture*; Houghton Mifflin Harcourt: Boston, MA, USA, 2005; ISBN 978-0-618-61959-7.
67. Lebra, T.S.; Lebra, W.P. *Japanese Culture and Behavior: Selected Readings*; University of Hawaii Press: Honolulu, HI, USA, 1986; ISBN 978-0-8248-1055-9.
68. Chinese "Jaywalker Catcher" Sprays Unruly Pedestrians with WATER. Available online: <http://www.dailymail.co.uk/news/article-5635117/Chinas-latest-jaywalker-catcher-humiliates-pedestrians-spray-WATER-road-crossings.html> (accessed on 5 November 2019).



© 2019 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (<http://creativecommons.org/licenses/by/4.0/>).