

Complex networks and simulation strategies: an application to olive fruit fly dispersion

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Abstract— In this work a study for the role of different environmental factors to the evolution of olive fruit fly, via an appropriate network of population traps is given. More explicitly, the olive fruit fly is a parasitic insect that infests olive groves in many countries. Through the use of a network of traps a simulation model was developed and used to simulate the dispersion of olive fruit fly inside a real olive grove for different environmental factors, such as different starting areas of olive fruit fly presence, different temperature sets as well as different drifting distances. Results showed that the level of infestation of the grove was not dependent on the limited areas the olive fruit fly emerged but on the drifting distance a fly could travel per day.

Keywords—trap network, simulation, modeling, olive fruit fly, dispersion

I. INTRODUCTION

Robust modeling of spatiotemporal environmental data has gain a lot of attention nowadays, because of their complex nature due to stochastic effects that emerge in real fields. To this end the use of a dense network of appropriate sensors or population traps is used in order to monitor and understand the underlying dynamic of environmental systems. Studies of such networks are focused either to the analyses of network data in order to attract knowledge and build analytical models or to the development of sophisticated simulation codes in order to explore dynamic properties of complex systems. In the following, an application of the aforementioned framework to the olive fruit fly is presented.

The parasitic insect pest *Bactrocera Oleae* commonly known as olive fruit fly, is a great menace of olive production in many olive oil producing countries. It causes significant damage in crop quality (aesthetic damage of fruit) as well as quantity (crop yield reduction) [1, 2].

During its biological cycle the olive fruit fly goes through four stages: egg, larva, pupa and adult. Olive fruits are the only growth habitat of the larva stage, thus adult female flies oviposit their eggs inside the olive fruits by stinging them. Once the larva emerges it feeds on the fruit which causes fruit damage and may result in a premature drop. The pupation stage

can happen either inside the fruit or in the soil. Finally, when the adult flies emerge a new cycle of mating and oviposition begins [3].

The population dynamics of the olive fruit fly depends heavily upon environmental conditions. Temperature is an important, if not the primary, environmental factor that affects the development of the immature stages of the olive fruit fly [4, 5]. Many models have been designed and used to relate temperature with insect development. The most common model that is being used is the degree-day model. With this model the heat that an organism accumulates during the day, between lower and upper thresholds, is calculated in degree-day units. Also, the total amount of degree-days required for an organism to develop from one life stage to the other can be determined. The most common methods for calculating degree-days are the single sine curve and mean temperature methods [6]. Other methods used are the max-min, “saw-tooth” and double sine curve methods [7].

Another environmental factor that can affect the development of the olive fruit fly is Relative Humidity [8, 9]. Both Temperature and Humidity are affected by temporal and spatial variations of solar radiation, such as time of day and period of the year, sky conditions and the presence of shading. Furthermore, proximity to the sea, or to mountains and exposure to the wind among others, can also impact on environmental conditions, due to the creation of a micro-climate. An olive grove field can all exist in the same micro-climate, or depending on the size of it and the specific topographic conditions present, numerous micro-climates could exist. This paper adopts a simplified approach and assumes the same micro-climate for all of the olive grove site used.

In olive groves a network of traps is used to monitor the population of the olive fruit fly and data obtained by those traps can indicate the level of infestation. The main reason for the use of traps is because of the olive fruit fly’s dispersion in the olive grove. Based on these data the following simulation model uses a network of traps to monitor the population of the olive fruit fly and can be used as a prediction tool. Other methods to monitor population outbreaks have been proposed [10, 11, 12], where stochastic algorithms were used to predict population outbreaks of users watching a video. The simulation

model used in this paper is based on work conducted by Voulgaris et al [13], where the robustness of the model was proved.

II. METHODOLOGY

A. Simulation model

Initially, the field where the evolution of the olive fruit fly's population would take place was constructed. The 2D grid that was constructed, simulated a real olive grove in St. George, Corfu, Greece with a total area of 9000m x 7400m, was divided into 100m x 100m trap cells. As a result the total grove is assumed that is covered by a grid network of 90 x 74 trap cells which were used to monitor the population of the olive fruit fly. Each monitoring cell was subsequently divided into 10 x 10 smaller cells, where each cell corresponded to a 10m x 10m area. Fig. 1 displays the produced grid where black colored cells represent the olive grove area while white colored cells represent areas where there are no olive trees (settlements, roads, etc.). The overlaying grey cells represent the grid network of trap cells.

The population of the olive fruit fly was divided into two categories: the immobile, which corresponded to the egg, larva and pupal development stage, and the mobile, which corresponded to the perfect adult insect both sexually mature and immature, population.

The development of the immobile population, throughout the simulation, was achieved by using the degree-day model. For each insect that entered one of the three stages belonging in the immobile population the total amount of degree-day units, needed to evolve to the next stage is computed. The following function is used to compute the accumulated heat:

$$(t_i - T_L) * (1 - (1 / (1 + \exp(-10 * (t_i - T_U)))))) \quad (1)$$

where t_i is the temperature of the i -th simulation step and T_L and T_U are the lower and upper developmental thresholds of the insect, respectively.

The adults of the mobile population in each simulation step

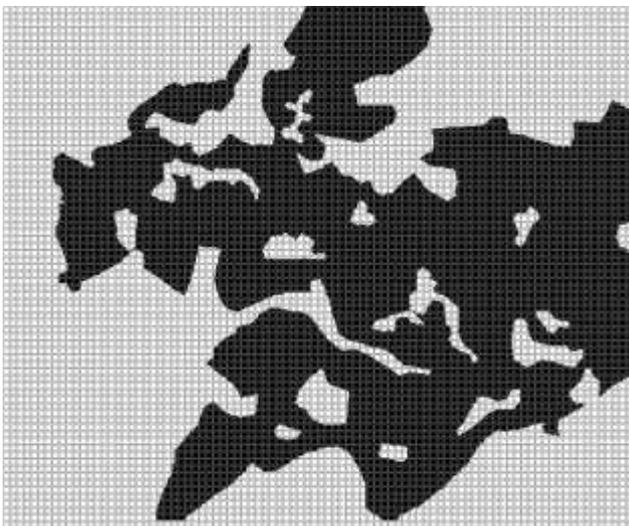


Fig 1. 2D construction of St. George, Corfu olive grove.

either stay in the cell that they were positioned in the previous step or drift in another randomly selected cell, which represents an olive grove area, which is inside a randomly selected radius of cells. Adult dispersion inside the olive grove is temperature dependent and currently is affected by high temperatures. The maximum distance (D) that the olive fruit fly can travel per simulation step is calculated by the following equation:

$$D = d_i * (1 - (1 / (1 + \exp(-10 * (t_i - T_M)))))) \quad (2)$$

where d_i is the randomly computed maximum distance the olive fruit fly can travel in the i -th simulation step, t_i is the temperature of the i -th simulation step and T_M is the temperature motion threshold, which is set to 35 °C [14] (as it is referenced in [15]). Fig. 2 displays the graph of (2), with d_i set to 60m. All temperatures referred to, are ambient environmental temperatures.

The simulation starts with the emergence of the overwintered generation. For the first 40 steps of the simulation random numbers of adult flies, in the range of 100 and 250, are positioned randomly inside the 2D grid. In an attempt to simulate the ripeness of the olive fruit, when oviposition can commence, the adult flies of the overwintered generation can start to reproduce 30 steps after the simulation starts.

B. Experiments

In order to examine the dynamic behavior of the olive fruit fly in space and time we ran the simulation model for two scenarios. In the first scenario the overwintered generation could emerge only in 25% of the olive grove, while in the second scenario the emergence could occur only in 50% of the olive grove. We ran each simulation scenario under two different sets of temperatures (Fig. 3) and under three maximum distance constraints an adult fly can travel in each simulation step.

Temperatures were obtained from the Hellenic National Meteorological Service as they were documented by their meteorological station in Corfu, Greece. The distance between the station (39°36'07''N, 19°54'42''E) and the olive grove site (39°42'51.3''N, 19°41'39.1''E) is 22.41 km, which can account for some variation due to differences in micro-climate. As these were the only reliable and available data, they were used in the simulation model. For both sets, temperatures from the time period of 21st of June to 31st of December were used.

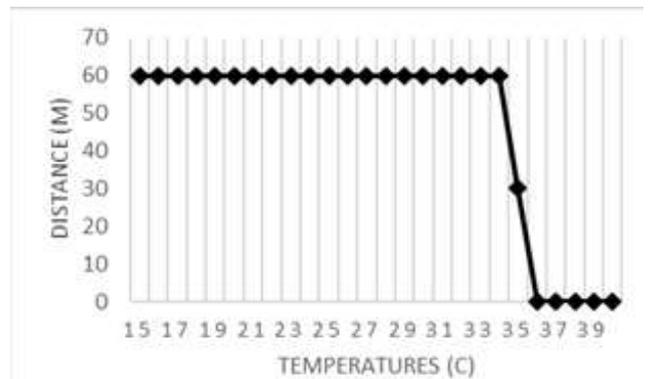


Fig. 2 Equation (2) graph, with d_i set to 60m and T_M set to 35°C.

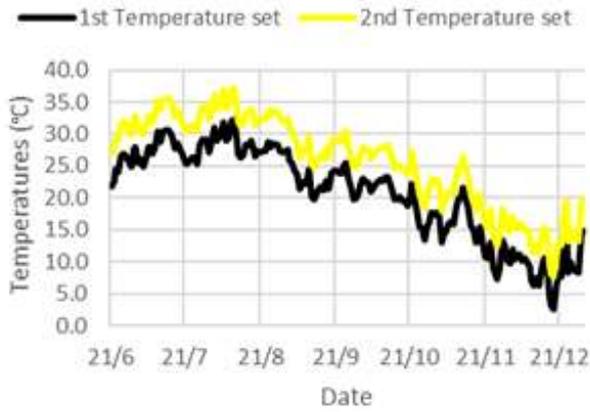


Fig. 3 Temperature sets.

The first set of temperatures were the mean daily temperatures of the year 2001 and the second set was the first temperature set increased by 5 °C uniformly throughout all time steps. This was done, so that a simplified version of a significantly warmer scenario can be tested. This could be in line with a future

climate change scenario that takes into account of global warming, or a warmer climate in another location.

The average distance an olive fruit fly can travel when host availability is low is 400m. On the other hand, when host availability is over 30%, average travel distance is shorter, about 180m. These findings were obtained from measurements done in Corfu, Greece [16] (referenced in [13]). Three distinct maximum drifting host constrains were placed on the adult flies which indicated host availability in the olive grove. For each scenario and for each temperature set, simulation runs were conducted where the olive fruit fly could travel up to a 10m (host availability is high) or 60m or 150m (host availability is low) radius, in each simulation step.

III. RESULTS

Fig. 4 displays the results of the simulation runs for the first scenario, where the overwintered generation could only emerge in 25% of the total olive grove. For each distance constrain it displays the dispersion of the olive fruit fly for the two temperature sets when the population size of the overwintered

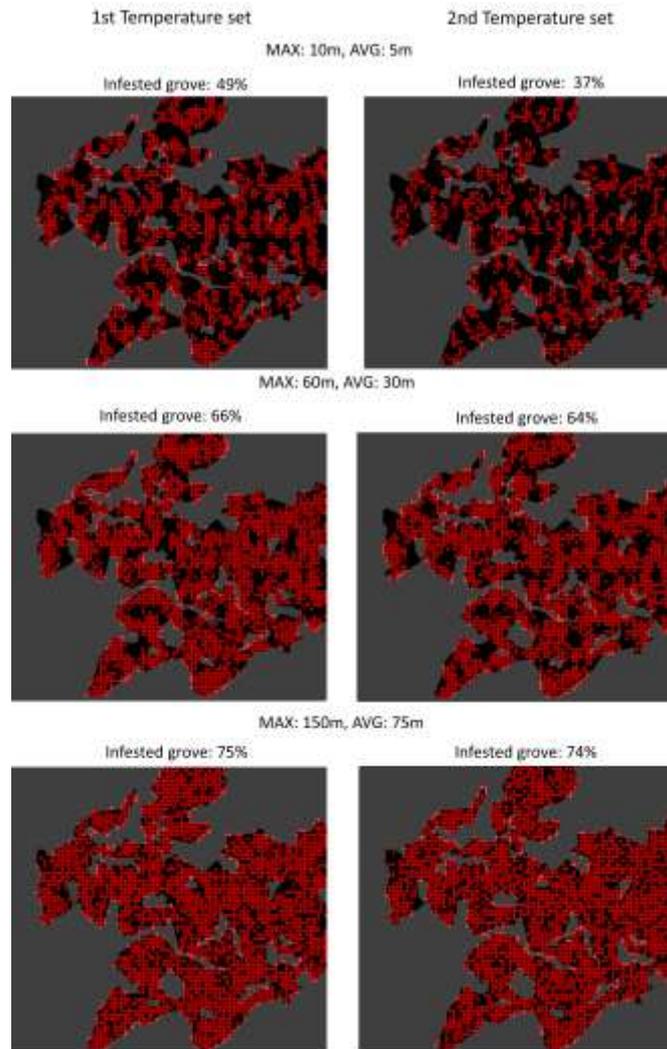


Fig. 4 Olive fruit fly dispersion for first scenario under two temperature sets and drifting distance constraints. Red colored cells indicate infestation, black colored cells indicate no olive fruit fly presence.

generation reaches its peak. Trap cells with red color indicate that the area is infested, while black colored cells indicate no olive fruit fly presence.

When the olive fruit fly can travel only small distances, like in the 10m radius constrain, many cells (black cells) show no presence of the olive fruit fly. For the first temperature set about half of the olive grove is infested, while for the second temperature set the infestation level is significantly lower about 37%. For larger distances, 60m and 150m constrain cases, infestation levels have increased to 65% and 74.5%, on

average, respectively.

The same behavior can be observed in Fig. 5, where the results of the simulation runs for the second scenario are displayed. In this scenario the overwintered generation could emerge only in 50% of the olive grove.

Again, in the 10m radius constrain for the first temperature set 58% of the olive grove is infested. On the other hand, for the second temperature set the infestation level has decreased by 8%. If we observe the results for the 60m radius constrain the

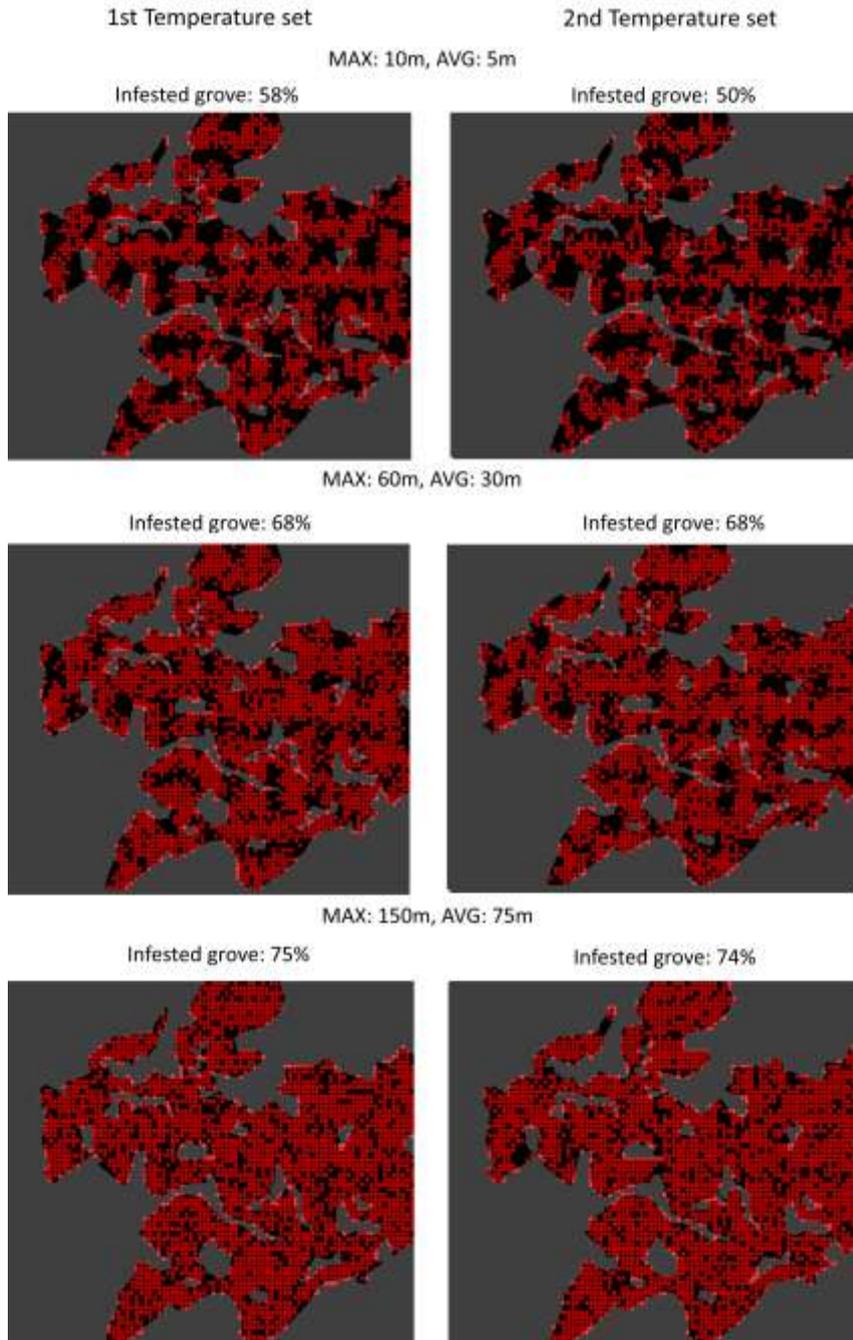


Fig. 5 Olive fruit fly dispersion for second scenario under two temperature sets and drifting distance constraints. Red colored cells indicate infestation, black colored cells indicate no olive fruit fly presence

number for both temperature sets 68% of the olive grove has been infested. Finally in the 150m radius constrain the olive fruit fly has infested three quarters of the olive grove.

When compared with the simulation runs in Fig. 4, one can observe that for the 10m and 60m distance constrains and both temperature sets, the level of infestation is greater. On the contrary, for the 150m distance constrain infestation levels are the same. This difference in infestation can be attributed to the starting points of infestation. Specifically in the first scenario up to 25% of the grove could be infested while in the second scenario up to 50% of the grove could be infested and through drifting the olive fruit fly can more easily infest the rest of the grove.

Fig. 6 displays the olive fruit fly's population growth over time for both temperature sets. A total of four generations emerged including the overwintered generation for the first temperature set, while for the second temperature set three generations emerged. The first generation for the second temperature set has a quite large time gap from the overwintered generation, which is a result of the higher temperatures the olive fruit fly had to develop under.

IV. DISCUSSION

Simulation runs were conducted for two different scenarios under two temperature sets and with different maximum drifting distance constrains on the mobile population of the olive fruit fly.

It was shown that no matter the number of starting areas the olive fruit fly emerges, even when the maximum travelling distance was 10m per day, a considerable percentage of the olive grove was infested although there are large areas that had no presence of the insect. However when the maximum drifting distance increased, 60m and 150m cases, the limited areas the overwintered adult insects emerged didn't seem to hinder the infestation of a considerable percentage of the olive grove.

Finally, for simulation runs where the temperatures were higher, which was the case with the second temperature set, the dispersion of the olive fruit fly inside the olive grove didn't seem to be affected when the olive fruit fly could travel greater distances (60m and 150m), because host availability was low. On the other hand, when host availability was high and thus olive fruit fly traveled a smaller distance (10m), the olive fruit fly didn't dispersed much inside the olive grove.

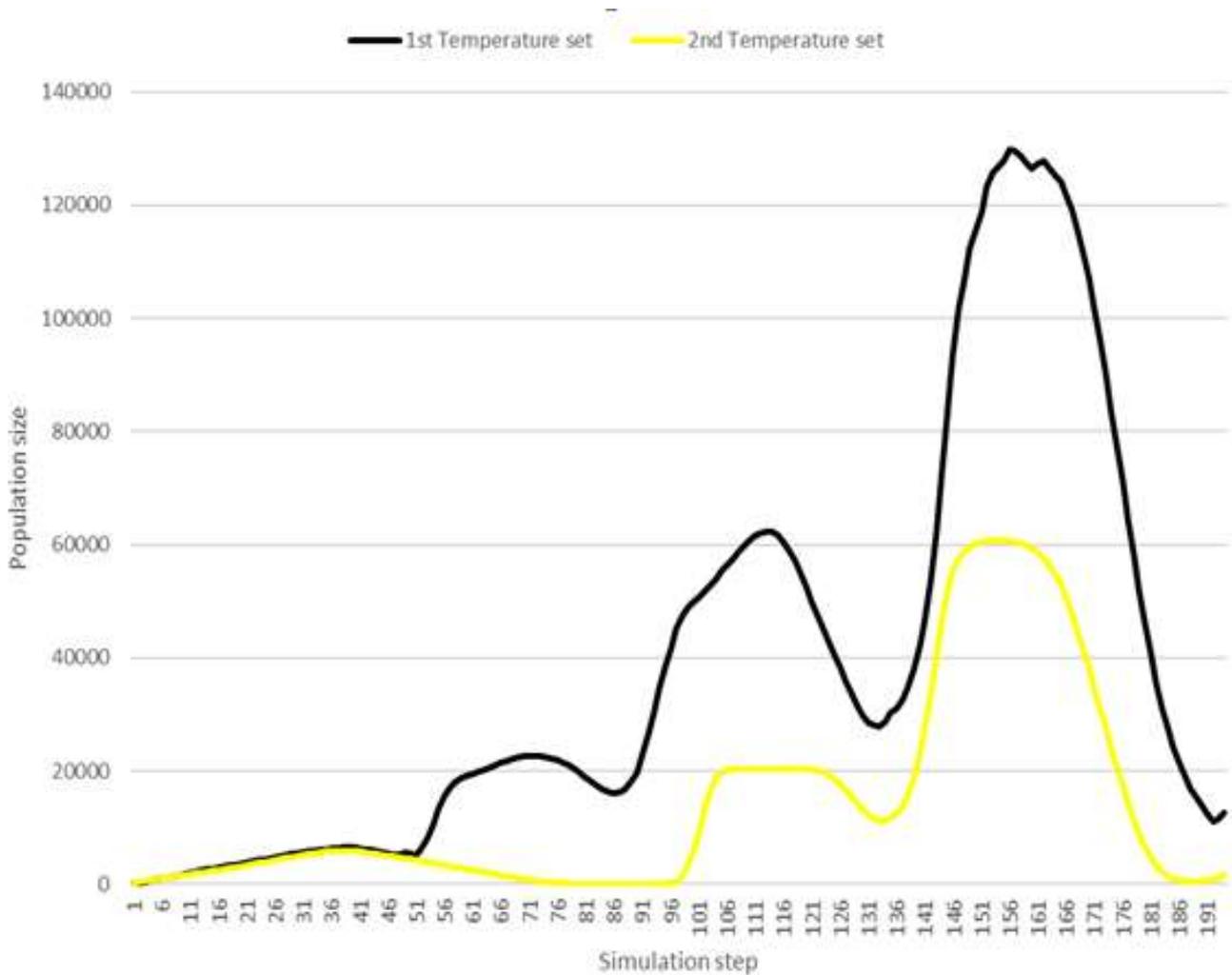


Fig. 6 Olive fruit fly's population dynamics under two temperature sets.

We conclude that the number of infested areas is a function of host availability and temperature.

Through a network of traps our simulation model can monitor the population size of the olive fruit fly during time as well as the dispersion of the flies in space. Our model could be used as a prediction and risk assessment tool of olive fruit fly's outbreaks.

V. FUTURE WORK

Temperature and Relative Humidity data are being collected on another olive grove site, so that temporal and spatial variations on environmental conditions can be captured and thus micro-climatic conditions established. The data will then be used to further test, extend, refine and validate the developed simulation model.

ACKNOWLEDGMENT

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