¹ Supplementary material

- ² Optimized timing of parasitoid release: a mathematical model
- ³ for biological control of Drosophila suzukii

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8 This document contains supplementary material for the manuscript (Pfab et al, 2018).

⁹ Here we describe the model equations in detail and derive the parameters used.

10 1 Framework

¹¹ Our model bases on the general approach by Nisbet and Gurney (1983). This approach al-

12 lows to model populations of insects with dynamically varying instar duration, which in our

13 case depend on the temperature of the environment. The method has proved useful in sev-

eral applications (McCauley et al, 2008; Nelson et al, 2013; Ewing et al, 2016). We extend

the basic model in a straightforward fashion to couple the dynamics of our two species, D.

¹⁶ suzukii and its parasitoid T. drosophilae. Our basic modeling parts are similar to the Ap-

¹⁷ pendix in (McCauley et al, 2008). The modeling parts are:

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D. suzukii densities (host)				
Ε	eggs			
L	larvae			
U	pupae			
Α	adults			
<i>T. drosophilae</i> densities (parasitoid)				
J	juveniles			
Р	adults			
Environment				
С	temperature			
F	fruit availability			

Functions and parameters (for stage i = E, L, U, A, J, P)

Notation	Explanation	Reference		
$\phi_i(C)$	background mortality rate	(11)-(12)-(14)-(15)		
δ_i	total mortality rate	(2)		
g(C)	speed of maturation	(10)		
$ au_i$	time spent in the stage <i>i</i>	(6)		
$f_A(C,F)$	rate of D. suzukii eggs deposited (rate per A	(17)		
	and F)			
$f_P(C,U)$	infestation rate of T. drosophilae (rate per P	(19)		
	and U)			
$\psi(F)$	competition coefficient for D. suzukii larva	(13)		

²⁰ Fecundity, mortality and the duration of the different juvenile stages depend on the tem-

²¹ perature, fruit availability and the population densities. The parameters and functions are

22 taken from different sources, see Section 3. All together, the model is formulated as

$$\frac{dE(t)}{dt} = R_E(t) - M_E(t) - \delta_E(t)E(t)$$

$$\frac{dL(t)}{dt} = M_E(t) - M_L(t) - \delta_L(t)L(t)$$

$$\frac{dU(t)}{dt} = M_L(t) - M_U(t) - \delta_U(t)U(t)$$

$$\frac{dA(t)}{dt} = M_U(t) - \delta_A(t)A(t)$$

$$\frac{dJ(t)}{dt} = R_J(t) - M_J(t) - \delta_J(t)J(t)$$

$$\frac{dP(t)}{dt} = M_J(t) - \delta_P(t)P(t).$$
(1)

The building parts are as follows. Mortality rates δ_i are composed of the temperature de-

²⁴ pendent background mortality rates ϕ_i , and competition and parasitism related terms (for *D*.

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25 suzukii larvae and pupae). That is

$$\begin{split} \delta_{E}(t) &= \phi_{E}(C(t)) \\ \delta_{L}(t) &= \phi_{L}(C(t)) + \psi(F(t))L(t) \\ \delta_{U}(t) &= \phi_{U}(C(t)) + f(C(t), U(t))P(t) \\ \delta_{A}(t) &= \phi_{A}(C(t)) \\ \delta_{J}(t) &= \phi_{J}(C(t)) \\ \delta_{P}(t) &= \phi_{P}(C(t)) \end{split}$$

$$\end{split}$$

$$(2)$$

²⁶ The renewal rates R_i and the maturation rates M_i are

$$R_{E}(t) = \beta_{A}(t)A(t)$$

$$M_{E}(t) = R_{E}(t - \tau_{E}(t))S_{E}(t)\frac{g(C(t))}{g(C(t - \tau_{E}(t)))}$$

$$M_{L}(t) = M_{E}(t - \tau_{L}(t))S_{L}(t)\frac{g(C(t))}{g(C(t - \tau_{L}(t)))}$$

$$M_{U}(t) = M_{L}(t - \tau_{U}(t))S_{U}(t)\frac{g(C(t))}{g(C(t - \tau_{U}(t)))}$$

$$R_{J}(t) = \beta_{P}(t)P(t)$$

$$M_{J}(t) = R_{J}(t - \tau_{J}(t))S_{J}(t)\frac{g(C(t))}{g(C(t - \tau_{J}(t)))}$$
(3)

27 with the birth rates

$$\beta_A(t) = \lambda_A h(C(t), F(t))F(t)$$

$$\beta_P(t) = \lambda_P f(C(t), U(t))U(t)$$
(4)

where λ_A and λ_P are the sex ratios of *D. suzukii* and *T. drosophilae*, to take into account

²⁹ that only females lay eggs. The functions f and h include the temperature dependence of the

fecundities and their leveling-off when there are many occasions for ovipositing. Further S_i are the stage survival probabilities for i = E, L, U and J,

$$S_i(t) = e^{-\int_{t-\tau_i(t)}^t \delta_i(\sigma) d\sigma}.$$
(5)

The maturation delays τ_i are given by the implicit relations

$$\Omega_i = \int_{t-\tau_i(t)}^t g(C(\sigma)) d\sigma \tag{6}$$

where g(C) is the speed of maturation in dependence of the temperature. The speed of maturation is normalized to be 1 at its maximum, making Ω_i the minimum duration of the stage.

³⁶ Under this assumption, we can apply a time-change to render the maturation delays ³⁷ constant. This technique is described in (McCauley et al, 2008; Nelson et al, 2013). Without ³⁸ going into detail, the transformation is

$$\tilde{t} = \int_0^t g(\sigma) d\sigma.$$
(7)

After this transformation the delays of the different stages become constant Ω_i , so that the system can be easily solved using readily available software (e.g. the MATLAB package

- dde23 (Shampine et al, 2000)). We chose to use Wolfram Mathematica (Wolfram Research,
- ⁴² 2016) to solve the system of delay differential equations (DDEs). After simulating the model
- ⁴³ under the transformation, we apply the inverse transformation to recover the dynamics in
- ⁴⁴ normal time. The *Mathematica* code is freely available on request.
- ⁴⁵ The idea of transforming the time to render the delays constant can be seen as a continuous
- 46 counterpart for discrete degree day models as the models for the dynamics of *D. suzukii*
- ⁴⁷ by (Wiman et al, 2014, 2016). Note that it would also be possible expressing the delays
- themselves through DDEs (Nisbet and Gurney, 1983; Johnson et al, 2015; Ewing et al, 2016) and use software which can solve DDEs with variable delays (e.g. the Fortran pack-
- ⁴⁹ 2016) and use software which can solve DDEs with variable delays (e.g. the Fortran pack-⁵⁰ age DKLAG6 (Thompson and Shampine, 2006), the Python package pyDDE, the MAT-
- LAB package ddesd (Shampine, 2005) and the R package PBSddesolve (Couture-Beil et al,
- ⁵² 2013)).

53 2 Starting conditions

- 54 We start the system with only adult *D. suzukii*, assuming that no eggs are laid before. That
- ⁵⁵ is for t < 0, $A(t) = A_0$, while E(t), L(t), U(t), $R_E(t)$, $M_E(t)$, $M_L(t)$ and $M_U(t)$ are all zero. ⁵⁶ To reduce the influence of the starting conditions, we run the simulations for an initial year
- ⁵⁷ which is disregarded.
- 58 Parasitoid adults are added in the same fashion after starting the simulation with D. suzukii
- ⁵⁹ alone. Given the time of the parasitoid release t_{intro} , we set for $t < t_{intro}$ the variables J(t),
- ⁶⁰ P(t), $R_J(t)$ and $M_J(t)$ to zero, and shift the adult parasitoid density at $t = t_{intro}$ to $P(t) = P_0$.

61 **3 Parameters and functions**

62 3.1 Maturation delays

The stage durations at various constant temperatures are taken for *D. suzukii* from (Tochen et al, 2014) (using the delays for females in Table 1 and 2 there) and for *T. drosophilae* from (Rossi Stacconi et al, 2017) (using the delays for females in Table 2 there). The data corresponds to the egg-larva stage and the pupal stage of *D. suzukii*, and the entire juveniles stage of *T. drosophilae*. We use the data separately to define the speed of maturation $g_i(C)$ for stage *i* as the inverse of the stage duration, and we fit its temperature dependence by a

⁶⁹ Gaussian bell curve of the shape

$$g_i(C) = \frac{1}{\Omega_i} e^{-\left(\frac{\mu_{g_i} - C}{\sigma_{g_i}}\right)^2} \tag{8}$$

where Ω_i is the minimum duration of a stage (at constant optimal temperature $C = \mu_{g_i}$). The

- ⁷¹ fit is obtained by the least square method in logarithmic scale. We approximate the speed of
- maturation of the different stages with a general speed function g(C)

$$g_i(C) \approx \frac{1}{\Omega_i} g(C)$$
 (9)

73 where

$$g(C) = e^{-\left(\frac{\mu_g - C}{\sigma_g}\right)^2} \tag{10}$$

⁷⁴ with μ_g and σ_g being the averages of the values we found for μ_{g_i} and σ_{g_i} , see Fig. 3.

⁷⁵ After the fit we divide the egg-larva stage of *D. suzukii* into egg and larva stage using the ⁷⁶ proportions measured by Emiljanowicz et al (2014) at constant temperature (in Table 5

there). The finally obtained parameter values are $\Omega_E = 1.3$, $\Omega_L = 5.3$, $\Omega_U = 4.3$, $\Omega_J = 16.9$,

there). The finally obtained parameter values are $\Omega_E = 1.3$, $\Omega_R = 28.9$ and $\sigma_g = 13.7$.

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80 3.2 Mortality

Mortality rates depend on the temperature. Note that experiments on the mortality of *D. suzukii* differ sometimes considerably in their outcomes (Kinjo et al, 2014; Tochen et al, 2014; Jakobs et al, 2015; Stephens, 2015; Shearer et al, 2016; Ryan et al, 2016; Enriquez

⁸⁴ and Colinet, 2017).

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D. suzukii adults

86 87

D. suzukli adults

We use the data on the adult life span of D. suzukii from (Shearer et al, 2016) (medium 88 life span of females from Figure 3 there). There the life length has been measured for two 89 90 different different phenotypes of D. suzukii: a summer morph and a winter morph, which 91 are induced by high and low temperatures respectively during juvenile development. For the sake of simplicity, we assume that the flies always exhibit the most favorable phenotype 92 and thus we take for each temperature the value of the better adapted phenotype. We fit the 93 average life lengths with a log-transformed Gaussian curve by the least square method in 94 logarithmic scale (see Fig. 5a). This form turned out to fit the data best after testing different 95

⁹⁶ formulas. We then use the inverse of that curve as the mortality rate, i.e.

$$\phi_A(C) = \frac{1}{\varepsilon_{\phi_A}} e^{\left(\frac{\log[C+s] - \log[\mu_{\phi_A} + s]}{\sigma_{\phi_A}}\right)^2}$$
(11)

⁹⁷ with s = 8. The obtained parameter values are $\varepsilon_{\phi_A} = 117.4$, $\mu_{\phi_A} = 2.6$ and $\sigma_{\phi_A} = 0.79$. For ⁹⁸ the simulations we assume that mortality in the wild is twice as large as in the laboratory ⁹⁹ due to predation from species different to *T. drosophilae*, and additional weather influences ¹⁰⁰ as wind, rain, humidity etc. The factor 2 for increased mortality in the wild is presumably a ¹⁰¹ conservative estimate; though we found no information about mortality of *D. suzukii* in the ¹⁰² wild, two recent papers suggest that mosquito mortality in the wild is around 4 times higher ¹⁰³ than under laboratory conditions (Cianci et al, 2013; Marini et al, 2016).

Note that the tolerance curve at lower temperatures is mostly guessed because we could
 not find clear data for long-time survival of winter morph *D. suzukii* adults at very low
 temperatures, although some efforts have been made into this direction (Ryan et al, 2016;
 Enriquez and Colinet, 2017).

¹⁰⁸ Juvenile D. suzukii

The juvenile mortality of *D. suzukii* rate is assumed to depend on temperature and competition when fruit availability is limited. For the temperature dependent background mortality rate, we use the same function $\phi_V(C)$ for all three juvenile stages of *D. suzukii*. The function is derived from the stage survival probabilities and stage durations of males and females from (Tochen et al, 2014) (from Table 2 and Table 3 there; survival can be deduced

¹¹⁰

from the sample size in this table by using that all replicates were started with 50 females and 50 males). This data is fitted by a Gaussian function of the form

$$\phi_V(C) = \frac{1}{\varepsilon_{\phi_V}} e^{\left(\frac{C - \mu_{\phi_V}}{\sigma_{\phi_V}}\right)^2} \tag{12}$$

and we obtain the parameter values $\varepsilon_{\phi_V} = 42.6$, $\mu_{\phi_V} = 17.0$ and $\sigma_{\phi_V} = 13.2$ by the least square method.

This function is the only mortality term for the egg stage. For the larva stage we include an additional term, which accounts for competition. As juveniles consume fruit, we assume that this term depends inversely proportional on fruit availability, without loss of generality

$$\psi(F) = \frac{1}{F}.$$
(13)

Any constant in front of the formula (13) can be absorbed in a scaling factor of host densities. For the plots, the population densities obtained in the simulations have been scaled to roughly match observed catch data. For the pupal stage there is an additional mortality term due to parasitism, f(C(t), U(t))P(t), which is described below in Section 3.3.

129 Adult T. drosophilae

Data on life length of *T. drosophilae* adults at different temperatures are presented by Amiresmaeili (2017), see Fig. 5b. To fit a mortality curve to that data, we proceed as for *D. suzukii* adults (dropping the data point for the survival of *T. drosophilae* at the maximum tested temperature because all insects died before the first census and this would conflict with the least square fit in logarithmic scale). We use the function

$$\phi_P(C) = \frac{1}{\varepsilon_{\phi_P}} e^{\left(\frac{\log[C+s] - \log[\mu_{\phi_P} + s]}{\sigma_{\phi_P}}\right)^2}$$
(14)

where we assume that s = 5, since *T. drosophilae* is thought to be more sensitive towards cold temperatures than *D. suzukii* (MVRS - Personal communication). The obtained parameter values are $\varepsilon_{\phi_P} = 79.1$, $\mu_{\phi_A} = 5.5$ and $\sigma_{\phi_A} = 0.88$.

Note that survival at low temperatures is guessed because we could not find data
 measured at cold conditions. For the simulations, we again doubled the mortality to account
 for the difference between laboratory experiments and field survival.

¹⁴³ Juvenile *T. drosophilae*

Juvenile mortality of *T. drosophila* has been measured by Wang et al (2016) at 23 °C (from the average in Fig. 2 with high host abundance there – note that the reported mortality of unexposed hosts has to be added to the values in this figure). Using the stage duration from the same work (from the average value in Fig. 3 there), we obtain the mortality rate $\phi_J(23) = 0.011$ at this temperature. For different temperatures we assume that the juvenile parasitoid mortality $\phi_J(C)$ scales with temperature as the juvenile mortality of *D. suzukii* and obtain

$$\phi_J(C) = \frac{\phi_J(23)}{\phi_V(23)} \phi_V(C).$$
(15)

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Fig. A1: Simulated *D. suzukii* adult population size, mean *D. suzukii* trap catches and mean daily temperature from S.Michele all'Adige, Province of Trento, Italy during 2014-2016. Different courses of the simulation correspond to different values for the *D. suzukii* attack rate $\alpha_A = 1$, 1.5, 2, 4 and 10. The thick curve corresponds to the attack rate we use for the other simulations, $\alpha_A = 2$

152 3.3 Fecundity

153 D. suzukii

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We assume that the daily fecundity of *D. suzukii* depends on two factors: the temperature *C* and the fruit availability *F*. To estimate the fecundity at high fruit availability we use the lifetime fecundity from Tochen et al (2014) (using the data measured on cherry, Table 3 there). From this data we calculate the daily fecundity by using the average life length from the same paper, see Fig. 4 (using the data measured on cherry, Table 1 there). The daily fecundity is then fitted by a skewed Gaussian curve

$$\eta_A(C) = \eta_{A_{max}} e^{-\begin{cases} \kappa_A & \text{if } C \le \mu_{\eta_A} \left(\frac{\mu_{\eta_A} - C}{\sigma_{\eta_A}}\right)^2 \\ 1 & C > \mu_{\eta_A} \end{cases}}.$$
(16)

We choose $\kappa_A = 3$ as a reasonable value for skew parameter, and obtain the parameter values $\eta_{A_{max}} = 4.8$, $\mu_{\eta_A} = 19.6$ and $\sigma_{\eta_A} = 6.4$ by the least square method in logarithmic scale.

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For the fruit dependence we assume a type 2 functional response with attack rate α_A . Additionally we include the adult sex ratio $\lambda_A = 1/2$ (Emiljanowicz et al, 2014). Summing ¹⁶⁷ up, the fecundity rate is assumed to be of the form

$$\beta_A = f_A(C, F)F$$
with $f_A(C, F) = \lambda_A \frac{\eta_A(C)\alpha_A}{1 + \alpha_A F}.$
(17)

In order to have an order-of-magnitude estimate for α_A , we computed simulations of the system with only *D. suzukii* for different values of α_A . By comparing those simulations in Fig. A1 with observed catches in the traps, we chose $\alpha_A = 2$ as a reasonable reference value. Note that this is a simplified way to model the fecundity of *D. suzukii*, neglecting possible details as reduced fecundity of the winter morph (Wallingford et al, 2016) and effects of female age and larval diet on fecundity and sex ratio (Emiljanowicz et al, 2014; Tochen et al, 2014).

176 *T. drosophilae*

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¹⁷⁷ We assume that the daily fecundity of *T. drosophilae* depends on temperature *C* and on ¹⁷⁹ the density of host pupae *U*, see Fig. 4. For the daily fecundity with high pupa abundance, ¹⁸⁰ we use the data of (Rossi Stacconi et al, 2017) (from Figure 4 there). This data is fitted by a ¹⁸¹ skewed Gaussian curve

$$\eta_P(C) = \eta_{P_{max}} e^{-\begin{cases} \kappa_P & \text{if } C \le \mu_{\eta_P} \left(\frac{\mu_{\eta_P} - C}{\sigma_{\eta_P}}\right)^2 \\ 1 & C > \mu_{\eta_P} \end{cases}}.$$
(18)

We use $\kappa_P = 3$ as a reasonable value for the skew parameter, and obtain the parameter values $\eta_{P_{max}} = 11.2$, $\mu_{\eta_P} = 22.0$ and $\sigma_{\eta_P} = 8.9$ by the least square method in logarithmic scale.

Using again a type 2 functional response with attack rate α_P we obtain the fecundity rate

$$\beta_P = f_P(C, U)U$$
with $f_P(C, U) = \lambda_P \frac{\eta_P(C)\alpha_P}{1 + \alpha_P U}$
(19)

where the sex ratio $\lambda_P = 0.53$ is obtained from (Rossi Stacconi et al, 2017) (using the total numbers of emerging individuals in Table 2 there).

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Again this functional form neglects several biological details, such as changing sex ra-189 tios, age-dependence of the fecundity and other physiological factors (Rossi Stacconi et al, 190 2017). Note also that different experiments suggest quiet different values for the fecundity. 191 On one hand, the experiments in (Rossi Stacconi et al, 2017) suggest that the daily fecundity 192 is lower over a longer time span and additional decreases with the age of the parasitoid. On 193 the other hand, experiments in (Kaçar et al, 2017) suggest much higher values for the short-194 term daily fecundity. The values we use are between those two extremes, so we believe they 195 are reasonable estimates. 196

¹⁹⁷ The parasitoid attack rate α_P cannot be reliably estimated from laboratory experiments, ¹⁹⁸ so we guessed it roughly by assuming that the parasitoids reach half of their potential ¹⁹⁹ fecundity during the peak season of *D. suzukii* peak season. The obtained value is $\alpha_P = 23$. ²⁰⁰ As this guess could be potentially misleading, we did a sensitivity analysis for its impact. ²⁰¹ For this analysis, we use different values of the parameter and repeat in Fig. A2 a simplified ²⁰² version of Fig. 10 (this figure shows our main objective, the influence of the parasitoid



Fig. A2: *D. suzukii* infestation index in dependence of the time of a single parasitoid introduction. The different lines correspond to different parameter values for the attack rate α_P of *T. drosophilae*. The thick black line corresponds to the value we used for the other simulations, $\alpha_P = 23$. The amount of parasitoids released equals 0.2% of the maximum *D. suzukii* adult population size reached without intervention.

release timing on the success of the intervention). As one can expect, choosing a higher

value for α_P increases the success of the intervention. Also, higher values slightly anticipate

- the optimal release timing, but the beginning of June is a reasonable choice in all scenarios.
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