## Downwind deposits of spray drift: a probabilistic approach

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## overview

#### Introduction

- Probabilistic spray drift modelling in fruit growing
- Probabilistic spray drift modelling, other cases
- Discussion





## Spray drift studies



## Spray drift studies



## overview

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## **Current situation**

- Fruit growing in NL: 19,000 ha
  - Pome fruit (apple, pear): 85%
- Current Dutch regulations: drift deposits on standardized ditch:
  - *Dormant*: drift deposits 17% of applied dose (<May 1)
  - *In full leaf*: 8.6% of applied dose (≥May 1)







## Exposure risk in fruit growing

Main objective of the project:

Development of *higher-tier assessment tool* for authorization of pesticides in *fruit* growing regarding the risk of *exposure* of *aquatic organisms* to *pesticides* 

- Considerations/limitations:
  - Scale = The Netherlands
  - Edge-of-field watercourses only
  - Spray drift is major entry route
  - As realistic as possible







## project set-up: `multi-stage rocket'



implementation into **DRAINBOW**: risk assessment model for exposure of aquatic organisms to ppp due to drift and drainage considering hydrology, time development etc.

+TOXSWA

selection of representative scenarios corresponding to a 90% risk level

exposure assessment model:

risk analysis for whole NL: simulation of ppp deposits onto all watercourses next to all orchards in NL

GIS maps

**spray drift model**: compute deposits of

plant protection product (ppp) onto edge-of-field watercourse next to an orchard one orchard, one ditch



### `stage 1': modelling spray drift in fruit growing

Spray drift model SPEXUS:

(**sp**ray drift **ex**posure model for **u**pward and **s**ideways applications)

- empirical, based on 20 years of field trials
- apple tree orchards (most important fruit crop in NL)
- regression analysis reveals most important factors



*spray deposits measured next to fruit orchard,* 0 – 25 m downwind, *in duplicate* 

#### Spray drift model basics

 $y = q_1 e^{-q_2 x^{c}}$  y = spray drift deposits; x = distance downwind

#### $q_1$ , $q_2$ , c : positive constants, depending on:

- wind speed
- wind direction
- ambient temperature
- canopy density (→ growth stage; BBCH)
- orchard size
- sprayer settings

Fitting the model to experimental data yields the optimal relations for  $q_1$ ,  $q_2$  and c

Details are presented in paper: Biosystems Eng. 154(2017):46-61 http://dx.doi.org/10.1016/j.biosystemseng.2016.08.016





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Special Issue: Spray Drift Reduction

**Research Paper** 

An empirical model based on phenological growth stage for predicting pesticide spray drift in pome fruit orchards



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#### ARTICLE INFO

ide history: blished online 16 Sept	
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An innovative spary drift model is developed to describe downwind deposits of perciclase applied an an other and of pome furth tere (paple, pear). The empirical model is based on 20 years of experimental data of downwind deposits of grazy drift for conventional cross-flow grazy applications. The model research the major factors ratio facting downwind deposits wind speed, wind direction, air temperature and density of the tree cancey. Modelling the cancey density of the trees as a continuous function of time is an innovative approach. Concey density is uniquely related to growth stage through the phenological BBGH index. Observed effects of the mentioned factors on deposits and discussed. Model results and meanned deposits also a con-kickino coefficient of 27%, while covering a range of almost three orient of magnitude. The model forms the basis for risk assessment for exposure of aquade cognitame concerning all alge of-field water bodies in the relatental, implementation of daff mitightion techniques is attrajightforward when appropriate experimental dats on relations of downwind gray deposits are sublish.

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#### Introduction

1.

Downwind off-target deposits of spray drift from pesticide applications have been investigated for many years. In the Netherlands, contamination of degl-of-field water bodies with pesticides sprayed is a major area of concern. Downwind spny deposits have been studied both experimentally and by simulation models. For field crops, where pesticides are commonly applied using boom sprayers, various spray drift models have been developed. Some of these models are based on particle tracking (Miller & Hadfield, 1999; Hokerman, Van de Zande, Porskamp, & Huijsmans, 1997; Butler Elfs & Miller, 2016), ohres use CFD techniques (Baetens et al., 2009) or a plume model (Lebeau, Verstraete, Stainler, & Destain, 2011), For spray applications in fruit crops, downwind deposits of spray drift are significantly higher than those for field crops, mainly caused by the sideways horizontally-driftered application of aprays using common orchard sprayers (Van de Zande, Porskamp, Michieken, Holterman, & Huijsmans, 2000). Therefore, in risk assessments to binportant to

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E-mail address: henkjan.holterman@wur.nl (H.J. Holterman). http://dx.doi.org/10.1016/j.biosystemseng.2016.08.016 1537-5110/© 2016 IAgrE. Published by Elsevier Ltd. All rights reserved 'stage 2': scaling up to countrywide approach

Implementation of the SPEXUS model into countrywide exposure assessment model (whole NL)

Finding a 90<sup>th</sup> percentile risk of exposure to ppp for all edge-of-field watercourses next to all fruit orchards in NL



### A. Spatial configuration

- Location, orientation and geometry of edge-of-field watercourses next to fruit orchards
- Spatial variables:
  - location of orchards (per districts)
  - watercourse types
  - water levels
  - orchard orientations
  - orchard side where watercourse is located





## Countrywide risk assessment model

#### simulation procedure: *spatial* variables



#### B. Temporal configuration



modified from: KNMI klimaatatlas

## Countrywide risk assessment model

simulation of meteorological years: *temporal* variables

Choose a spray application scenario (*number of applications, dates, application techniques, ...*)

dates determine canopy density (growth stage)
 → amount of spray drift



## `stage 3': representative configuration

*It is impossible to carry out full-scale risk assessment combining exposure and fate for all situations countrywide* 

#### Workaround:

- Can we select a single spatial configuration as a representative of all possible configurations?
- So that studying this *single* configuration allows us to do a *countrywide* risk assessment?







Selection criteria for single config:

- Overall risk level ~90%
- Important spatial configuration
- Important fruit growing region
- Common water body type
- Common orchard orientation
- Common summer water level



## Countrywide scenarios procedure

- Selection of *limited* set of spray application scenarios
- Countrywide simulations for risk assessment, for these scenarios → compute overall 90<sup>th</sup> % PEC
- Selection of a single spatial configuration \_\_\_\_\_\_ as a representative of all possible configurations
- Simulations for the single configuration during many stochastic years (e.g. 10,000 y)
- Determination of temporal percentile of single configuration corresponding to the overall 90<sup>th</sup> % PEC

*PEC* = *predicted environmental concentration* 





a relatively common watercourse with a common water depth, in a district with lot of fruit growers, next to an orchard with common orientation

# Typical scheme for spray applications in apple tree orchards

#### Week number







#### Selected basic scenarios

## Five scenarios selected to represent most types of spray treatment and fate of pesticides

	scenario					
code	E1	L1	E3	L3	S15	
application date	<b>E</b> arly	Late	Early	Late	Season	
# spray applications	1	1		3		

early = canopy starts developing (May)
late = in full leaf (August)
season = during summer season
multiple applications: 1 week interval;
full dissipation is assumed within one week





## How to determine appropriate temporal percentile $T_{90}$ example: basic scenario E1



- 100 years
- 74,000 spatial configurations
   → 7,400,000 PECs

to conclude for E1: in a multi-year study for the **single** configuration the 58<sup>th</sup> percentile corresponds to the **countrywide** 90<sup>th</sup> percentile risk

- 10,000 years
- 1 selected spatial configuration (monitoring)

# How to determine appropriate temporal percentile $T_{90}$ example: basic scenario E1







- Determine overall PEC<sub>90</sub> (countrywide)
- Lookup this value in local cpdf
- Find corresponding temporal percentile  $T_{90}$

#### Apply this procedure to

- 5 basic scenarios: E1, L1, E3, L3, S15
- 7 pesticide application techniques: conventional + 6 drift reducing techs
- 10 crop-free buffer zones: 0-9m
- $\rightarrow$  table of 350 T<sub>90</sub> values !

real situations (not covered by the basic scenarios) can be approximated by 'smart interpolation' using these 350 cases

#### Scenario E1 20 yrs distributions, *selected* spatial config



![](_page_21_Picture_3.jpeg)

#### Scenario E1 20 yrs distributions, selected spatial config

![](_page_22_Figure_1.jpeg)

#### Selecting the year with PEC close to the given percentile

![](_page_22_Picture_3.jpeg)

![](_page_22_Picture_4.jpeg)

#### Scenario E1 20 yrs distributions, *selected* spatial config

![](_page_23_Figure_1.jpeg)

![](_page_23_Picture_2.jpeg)

![](_page_23_Picture_3.jpeg)

#### All basic scenarios

countrywide simulation, N=100yrs

![](_page_24_Figure_2.jpeg)

#### Scenario S15: 15 spray applications 20 yrs distributions, *selected* spatial config

Assumption: risk is governed by *maximum* PEC of 15 applications

![](_page_25_Figure_2.jpeg)

![](_page_25_Picture_3.jpeg)

![](_page_25_Picture_4.jpeg)

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![](_page_26_Picture_5.jpeg)

![](_page_26_Picture_6.jpeg)

Probabilistic modelling, other crops/situations

- Under development...
- Field crops: same set-up seems possible:
  - Spray drift model *IDEFICS*
  - Combining NL maps of crops, edge-of-field watercourses
  - How to implement *crop rotation*? Is it relevant?
- Non-target arthropods & plants
- Exposure risk for workers, bystanders, residents

![](_page_27_Picture_8.jpeg)

## overview

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- Probabilistic spray drift modelling in fruit growing
- Probabilistic spray drift modelling for field crops

### Discussion

![](_page_28_Picture_5.jpeg)

![](_page_28_Picture_6.jpeg)

#### Discussion

- Highly drift-reducing techniques: input from *drains* may be significant! How does this affect the present results?
- Using selected ditch with limited number of basic scenarios is not 'the real thing'
- Scenarios with *slow dissipation* of pesticides in ditch are not parameterised yet: *challenge!*
- Other countries/climates: same procedure should be possible, provided that all relevant data is available
- Regulatory implementation: combined exposure & fate; higher tiers: might be a `long and winding road' ...

![](_page_29_Picture_6.jpeg)

![](_page_29_Picture_7.jpeg)

![](_page_30_Picture_0.jpeg)

![](_page_30_Picture_1.jpeg)

![](_page_30_Picture_2.jpeg)