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1 **Buffer strip width and agricultural pesticide contamination in Danish lowland**
2 **streams: Implications for stream and riparian management**

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11

12 **Abstract**

13 According to the European Water Framework Directive member states are obliged to ensure that all
14 surface water bodies achieve at least good ecological status and to identify major anthropogenic
15 stressors. Non-point source contamination of agricultural pesticides is widely acknowledged as one
16 of the most important anthropogenic stressors in stream ecosystems.

17 We surveyed the occurrence of 31 pesticides and evaluated their potential toxicity for benthic
18 macroinvertebrates using Toxic Units (TU) in 14 Danish 1st and 2nd order streams in bed sediments
19 and stream water during storm flow and base flow. Total pesticide concentrations and toxic
20 potential were highest during storm flow events with maximum TU ranging from -6.63 to -1.72. We
21 found that minimum buffer strip width in the near upstream area was the most important parameter
22 governing TU. Furthermore, adding a function for minimum buffer strip width to the Runoff
23 Potential (RP) model increased its power to predict measured TUs from 46% to 64%. However,
24 including a function for tile drainage capacity is probably equally important and should be
25 considered in future research in order to further optimise the RP model. Our results clearly
26 emphasise the importance of considering buffer strips as risk mitigation tools in terms of non-point
27 source pesticide contamination. We furthermore apply our results for discussing the minimum
28 dimensions that vegetated buffer strips should have in order to sufficiently protect stream
29 ecosystems from pesticide contamination and maintain good ecological status.

30

31 Key words: Buffer strip, pesticides, runoff, Water Framework Directive, non-point sources

32 **1. Introduction**

33 Non-point source contamination of streams with pesticides applied in agricultural production is
34 widely acknowledged as one of the greatest stressors to stream ecosystems, and various routes for
35 pesticide transport from the field to stream recipients have been identified (Neumann et al., 2002;
36 Schulz, 2004). There is a clear consensus in the existing literature verifying surface runoff and flow
37 through tile-drains as the most important pathways for non-point pesticide losses in agricultural
38 catchments (Kreuger, 1998; Kronvang et al., 2004; Neumann et al., 2002; Wauchope, 1978). As a
39 consequence, the highest pesticide concentrations occur during heavy precipitation events, and the
40 footprint of pesticides is proposed to be more distinct in small streams due to a closer connectivity
41 between land and stream (Kreuger & Brink, 1988; Probst et al., 2005; Schulz, 2004).

42 According to the European Water Framework Directive (WFD), member states are obliged to
43 measure and ensure that all surface water bodies achieve at least good ecological status within a
44 defined timetable (European Commission, 2000). Requirements are not only to assess the overall
45 ecological quality of surface waters, but also to identify the major environmental and/or
46 anthropogenic drivers of ecological degradation and the extent of impairment. Several biotic indices
47 and multi-metric procedures have been developed attempting to robustly characterise the impact of
48 selected stressors that result in the deviation from good ecological status (Furse et al., 2006).

49 Non-point source pesticide contamination of rivers potentially poses a threat to all stream
50 dwelling organisms (Liess et al., 2005), and there is a growing interest to develop and provide field-
51 based models to assist in characterising the non-point source pesticide contamination that originates
52 from agricultural practices (Friberg et al., 2003; Schäfer et al., 2007, 2011a; Schulz, 2004).

53 However, there is still a need for additional studies that investigate the loss, occurrence and fate of
54 agricultural pesticides in streams and their impact on stream biota. Establishing causal relationships
55 between pesticides and their impact on flora and fauna is difficult due to natural variability in

56 stream ecosystem communities and the co-existing pressures from several other anthropogenic
57 stressors (Liess et al., 2005; Rasmussen et al., 2011). However, Liess & von der Ohe (2005)
58 introduced the SPEcies At Risk indicator for pesticides (SPEAR_{pesticides}), which has been validated
59 as a selective indicator that successfully separates the effects of pesticides from those of other
60 anthropogenic stressors (Schäfer et al., 2007, 2011a). Furthermore, Schriever et al. (2007b) found
61 that SPEAR_{pesticides} was the biological parameter best describing stream macroinvertebrate
62 community responses to a modelled indicator of pesticide surface runoff (RP). In contrast,
63 Rasmussen et al. (2011) were unable to link RP with SPEAR_{pesticides} using a large dataset of small
64 Danish streams, which could be due to the presence of wider buffer strips along Danish streams
65 compared to German streams. Since buffer strip information is not integrated into the RP model but
66 is known to significantly influence pesticide runoff, different buffer strip characteristics between the
67 two sets of study streams can plausibly explain the different results. Implementing a function for
68 buffer strip width (representing a simplified measure for pesticide runoff retaining capacity) might,
69 therefore, significantly improve the predictive power of the RP model.

70 In this study we screened 14 Danish 1st and 2nd order streams for pesticides that are frequently
71 applied in normal agricultural practices in their respective catchments. The study aims were to 1)
72 characterise pesticide occurrence and potential toxicity for benthic macroinvertebrates in Danish
73 streams, 2) identify the environmental parameters that most strongly govern pesticide occurrence
74 and toxicity, and 3) improve the predictive power of the RP model by using detailed environmental
75 data and by adding a function for buffer strip width.

76

77 **2. Materials and methods**

78 2.1 Study area

79 The field campaign was conducted in 2009 in a set of study streams that is located on Funen,
80 Denmark (Fig. 1), where catchments are characterised by low elevation and loamy soils with
81 medium to low infiltration capacity. Agriculture and forest are the dominant types of land use.
82 Climatic conditions are temperate and the average regional precipitation is 700 mm year⁻¹.
83 Dominating crop types in the studied catchments were rye, wheat, barley, grass and oilseed rape
84 (Appendix A).

85

86 2.2 Stream characteristics

87 Fourteen 1st or 2nd order streams were selected based on the following selection criteria: year-round
88 water flow, no maintenance activities conducted during the sampling period (dredging and weed-
89 cutting) and no sources of pollution other than from agricultural non-point sources. The streams
90 represent a gradient of potential pesticide contamination predicted from the proportion of adjacent
91 agricultural land. In order to optimise the selection of streams, the pesticide runoff was predicted by
92 applying the runoff potential (RP) model (see also Schriever et al., 2007a, b). The RP-model is a
93 generic indicator that was developed to quantify the risk of pesticide runoff contamination to
94 streams from agricultural land (Schriever et al., 2007a). Calculated RP for site selection support was
95 based on the assumption that any runoff-triggering precipitation event would be evenly distributed
96 among the studied streams. Data input for grown crops and pesticide application was based on 2008
97 data (Danish EPA, 2009).

98 Using aerial photographs, buffer strip dimensions (minimum and average buffer strip width)
99 were determined for each stream by digitalising buffer strips in 500, 1,000 and 2,000 metres
100 sections upstream of the sampling sites in ArcGis 9.2. Average buffer strip width was calculated by

101 simple mathematical integration of the digitalised buffer strip area. The outer boundaries of buffer
102 strips were characteristically visible using summer photos, since buffer strips are relatively
103 unmaintained compared to conventional agricultural fields and fallow land. Consequently, the
104 different types of vegetation found in the buffer strips clearly defined their outer boundaries.

105

106 2.3 Quantification of pesticide contamination

107 The selection of analysed pesticides was based on application frequency and total applied amounts
108 in 2008 (Danish EPA, 2009). This list was augmented with a series of banned pesticides that are
109 commonly found in drinking water wells. In total, 19 herbicides, 6 fungicides and 6 insecticides
110 were included in the sampling program (Appendix B). The sampling campaign was conducted in
111 2009.

112 We used event-triggered samplers to characterise pesticide contamination during heavy
113 precipitation events (Liess and von der Ohe, 2005). The sampling system consisted of two 1 L glass
114 bottles that were deployed in the flowing part of the stream channel. Bottles were filled passively
115 through small (0.5 cm in diameter) glass tubes when the water level increased above the glass tube
116 opening. The two bottles were positioned 5 cm and 10 cm above base flow water level,
117 respectively. Filled water samples were retrieved within 24 hours after each heavy precipitation
118 event. During the sampling period, two precipitation episodes triggered the sampling system. The
119 first episode occurred on the 28th of May and was characterised by a precipitation depth ranging
120 from 7 mm to 10 mm depending on the site. This episode triggered samplers in only six streams.
121 The second episode occurred on the 12th of June and was characterised by a precipitation depth
122 ranging from 19 to 47 mm. The latter triggered the sampling system in all streams.

123 Bed sediment was sampled (stratified sampling) on the 20th of July using a kajak corer (8 cm
124 diameter). All sediment samples were collected within a 50 m stream section extending upstream

125 from the event triggered samplers. One sample consisted of a minimum of 30 sub-samples from the
126 top layer (1-2 cm) of newly deposited sediment at in order to obtain sediment samples that generally
127 were representative for the respective reaches (see also Friberg et al., 2003).

128 Water samples were collected manually in August during low flow conditions in order to
129 characterise the potential ‘background input’ of pesticides originating from groundwater inflow.
130 Banned pesticides were detected in all streams indicating the importance of groundwater input as a
131 source of pesticides. However, in our study, pesticides in the August samples were characterised by
132 a combination of low concentrations and low toxicity to benthic macroinvertebrates. Consequently,
133 we assumed that pesticides originating from groundwater input were of minor importance in the
134 studied streams.

135 The pesticide analyses (including solid phase extraction) were conducted by OMEGAM
136 laboratories in Amsterdam; unfiltered samples were sent to the laboratories in coolers immediately
137 after collection. The final extract of each sample was used in different analysis programs. Analysis
138 programs were based on gas-chromatography mass-spectrometry (GC-MS) or liquid-
139 chromatography mass-spectrometry (LC-MS). The limit of quantification for each compound was
140 determined as the lowest concentration that can be reliably quantified (95% confidence interval)
141 (Appendix B). Detection limits were 0.01-0.1 $\mu\text{g L}^{-1}$ for water samples and 0.01-0.1 mg kg^{-1} (dry
142 weight) for sediment samples. Results were corrected for recovery, which was determined by
143 spiked samples. For all compounds, recovery was reported to be within 85% - 110% of actual
144 concentrations.

145

146 2.4 Predicted pesticide exposure

147 The runoff potential model was produced to predict runoff contamination of a generic compound
148 instead of predicting actual runoff losses for a specific compound. However, due to the high

149 resolution and quality input data (field block-specific crop data) we were able to meet data
 150 requirements for a more detailed version of the model in terms of grown crops (Eq. (1)). Due to the
 151 high resolution of crop data, we could additionally improve our estimates for pesticide application
 152 rates using the average compound-specific application rate for each crop type in 2009 (Danish EPA,
 153 2010). Thus, we could calculate the runoff potential for the compounds associated with each crop
 154 type instead of just predicting runoff for a generic compound. For further details on the original RP
 155 model, consult Schriever et al. (2007a). We calculated RP for all sites applying a two-sided corridor
 156 of 100 metres extending 500 metres upstream of the sampling location. Modification of the
 157 considered catchment size, i.e. implementation of other corridor lengths (1,000 or 2,000 m) or
 158 utilisation of the total catchment had no significant effect on the results. For convenience the two-
 159 sided 100 metres corridor extending 500 metres upstream will be referred to as the stream corridor.
 160 We calculated pesticide runoff by first applying the runoff model underlying the RP (modified after
 161 Schriever et al. (2007a):

$$162 \quad \text{gLOAD} = \sum_{i=1}^n \sum_{j=1}^m \sum_{l=1}^k A_{i,j} \cdot D_l \cdot \left(1 - \frac{I_j}{100}\right) \cdot \frac{1}{1 + \frac{\text{Koc}_l \cdot \text{OC}_i}{100}} \cdot f(s_i) \cdot \frac{f(P_i, T_i)}{P_i} \quad (1)$$

163 where index i refers to the respective field blocks, index j refers to different crop types present on
 164 the fields, and index l refers to specific pesticides. $A_{i,j}$ is the size of agricultural land (ha), D_l is the
 165 application rate of the pesticide compound, I_j is the crop- and growth phase-specific plant
 166 interception of the substance at the time of the precipitation event (%), Koc_l is the organic carbon
 167 sorption coefficient of the pesticide compound, OC_i is the soil organic carbon content of a field
 168 patch (%), s_i is the mean slope of a field (%), $f(s_i)$ describes the influence of the field slope. P_i is the
 169 precipitation depth (mm) of the considered event, T_i refers to the soil texture of a field
 170 (sandy/loamy), $f(P_i, T_i)$ is a function describing the surface runoff volume for vegetated soils in the
 171 middle or late period for vegetation growth. RP (Eq. (2)) is then calculated as:

172
$$RP = \log\left(\max_{i=1}^n(\text{gLOAD}_i)\right)$$
 (2)

173 The runoff potential model was parameterised as follows: field-specific crop types for each field
174 block in the stream corridor were extracted from a national Danish database (LOOP) (Grant et al.,
175 2006). Soil slope in the stream corridor was estimated using a Digitalised Elevation Map (DEM)
176 with 1.6 metres resolution in ArcGis 9.2. Soil texture composition (including humus content) within
177 the stream corridor was extracted from the Hair database (Greve et al., 2007). According to Thomas
178 & Goudie (2000), sandy soil was defined as soils containing < 10% clay and > 85% sand. The
179 relative organic carbon content of soils was calculated as 57% of the humus content (Thomas &
180 Goudie, 2000). The average crop-specific application rate for each pesticide compound potentially
181 applied in 2009 was extracted from national pesticide statistics (Danish EPA, 2010). Precipitation
182 data was provided by the Danish Meteorological Institute (<http://www.dmi.dk>) (100 km²
183 resolution). The daily recorded precipitation was assumed to result from a single precipitation
184 event. Plant interception values (I_j) were assigned to all crop types that were present during the
185 considered precipitation event according to Linders et al. (2000).

186

187 2.5 Data analysis

188 We applied toxic units (TU) as a measure for pesticide toxicity, calculating TU for all pesticides
189 detected in each sample. TU values are based on the acute 48h LC50 value for *Daphnia magna*, as
190 given in Tomlin (2001) (eq. (3)).

191
$$TU_{(D. magna)} = \log(C_i/LC50_i)$$
 (3)

192 where $TU_{(D. magna)}$ is the toxic unit for pesticide i , C_i is the measured concentration of pesticide i and
193 $LC50_i$ is the corresponding 48h LC50 value for *D. magna* exposed to pesticide i . We identified the
194 maximum TU for each water sample, and additionally calculated the summed TU for all pesticides
195 in each water sample. The summation of all TUs is based on the assumption that all compounds act

196 under the principle of toxic additivity. As the number of components in a toxic mixture increases,
197 the range of deviation from toxic additivity is proposed to decrease (the Funnel hypothesis) (Warne
198 & Hawker, 1995).

199 All environmental parameters considered (minimum and average buffer strip width,
200 proportion of agriculture in the stream corridor, crop types, estimated pesticide application, field
201 slopes and soil texture) were then correlated to the summed TU, maximum TU, number of
202 pesticides and sum concentration of pesticides using Spearman rank order (r) correlations ($P < 0.05$).
203 All tests were performed using the software SAS enterprise guide 4.2. Leverage and Cook's
204 Distance were calculated for all fitted regressions in order to evaluate the contributed weight of
205 each data point. No values for Cook's Distance exceeded 0.1 and no leverage values were greater
206 than $2*(p/n)$, where p is the number of parameters in the model including the intercept, and n is the
207 total number of observations. R^2 values are given for all presented regressions.

208 In addition, we attempted to improve the RP model by implementing various functions of
209 minimum and average buffer strip width in the stream corridor. A fitted regression of the modified
210 RP model as a function of calculated TUs was compared to that of the original RP model using
211 Analysis of Covariance (ANCOVA) ($p < 0.05$) in SAS 9.2.

212 3. Results

213 3.1 Pesticides and TU

214 The results of the field campaign disclosed a total of 13 herbicides, 5 fungicides and 2 insecticide
215 that were actually detected in water samples from the 14 study streams (Table 1). Summed
216 concentrations ranged from 0.01 to 3.17 $\mu\text{g L}^{-1}$, the number of detected pesticides per sample
217 ranged from 1 to 13, maximum TU ranged from -6.63 to -1.72, and summed TU ranged from -6.63
218 to -1.57. In total, five of the nine streams at risk for receiving pesticide runoff (proportion of
219 agricultural land $\geq 50\%$) were characterised by at least one sample with summed and maximum
220 TUs ≥ -3 . The carbamate insecticide Pirimicarb and the Strubilurine fungicide Azoxystrobin were
221 the pesticides primarily responsible for the high TU values due to corresponding low $\text{LC50}_{(D. magna)}$
222 values. No pesticides were detected in the sediment samples.

223 Minimum buffer strip width was the environmental parameter most strongly correlated with
224 summed TU and maximum TU ($r = 0.80$, $P < 0.0001$, Fig. 2a), followed by the proportion of
225 agricultural land in the stream corridor ($r = 0.48$, $P < 0.05$, Fig. 2d). Applying the maximum TU
226 generated a comparable significant correlation with minimum buffer strip width ($r = 0.78$,
227 $P < 0.0001$, Fig. 2b) and a slightly stronger significant correlation with the proportion of agriculture
228 in the stream corridor ($r = 0.66$, $P < 0.01$, Fig. 2e). Applying the average buffer strip width generated
229 a significant but weaker correlation with summed TU and maximum TU ($r = 0.61$, $P < 0.01$ and $r =$
230 0.65 , $P < 0.01$, respectively) (data not shown). Furthermore, the number of pesticide compounds was
231 significantly correlated to the minimum buffer strip width ($r = 0.72$, $P < 0.001$, Fig. 2c) and the
232 proportion of agricultural land in the stream corridor ($r = 0.49$, $P < 0.05$, Fig. 2f). Autocorrelations
233 were found between the summed TU and the number of pesticides ($r = 0.82$, $P < 0.0001$), as well as
234 total pesticide concentration ($r = 0.71$, $P < 0.001$) (data not shown). Furthermore, total pesticide
235 concentration was autocorrelated with the number of pesticides ($r = 0.90$, $P < 0.0001$) (data not

236 shown). The proportion of agricultural land was significantly correlated to minimum and average
237 buffer strip width in the stream corridor ($r = 0.66$, $P < 0.01$ and $r = 0.73$, $P < 0.001$, respectively), as
238 shown in Fig. 3. No correlation was found between estimated compound-specific applied amounts
239 of pesticides in the stream corridor and in-stream concentrations of the respective compounds.

240

241 3.2 Predicted pesticide exposure

242 The runoff potential model (RP) was significantly correlated with the summed TU ($r = 0.70$,
243 $P < 0.001$, Fig. 4a) and the maximum TU ($r = 0.63$, $P < 0.01$) (data not shown). Adding the inverse
244 function for minimum buffer strip width (within a 2×100 m stream corridor extending 500 m
245 upstream from a sampling point) to the runoff model (underlying RP) by simple multiplication
246 improved the significance of the correlation found between the RP and the summed TU ($r = 0.83$,
247 $P < 0.0001$, Fig. 4b) and the maximum TU ($r = 0.70$, $P < 0.001$) (data not shown), reflected by reduced
248 data variability around the fitted regression. In other words, the explanatory power of the model
249 increased from 46% to 64% by adding the inverse function for minimum buffer strip width to the
250 RP model. Slope and intercept were not significantly different between the two regression lines
251 ($P < 0.05$).

252

253 **4. Discussion**

254 4.1 The influence of buffer strips on the occurrence of pesticides in streams

255 Minimum buffer strip width within a two-sided 100 m stream corridor extending 500 m upstream
256 from the pesticide sampling point was the environmental parameter most strongly correlated with
257 summed and maximum TUs for pesticides in stream water during storm flow. Decreasing summed
258 and maximum TUs with increasing minimum buffer strip width probably reflects runoff reduction,
259 due especially to infiltration and pesticide adsorption to organic matter within the buffer strip
260 (Anbumozhi et al., 2005; Lacas et al., 2005; Vidon et al., 2010). Minimum buffer strip width was
261 autocorrelated with the proportion of agricultural land in the stream corridor and hence buffer strip
262 width may act as a surrogate for the proportion of agricultural land. However, numerous site-
263 specific studies document clear effects of buffer strips as a useful tool for reducing pesticide
264 transport from fields to stream recipients. For example, both Lacas et al. (2005) and Schriever et al.
265 (2007a) found that precipitation intensity and local field characteristics (field slopes and crop
266 types/growth phases) were more sensitive parameters than the proportion of agricultural land in the
267 sub-catchments when predicting pesticide runoff. The strong correlations between minimum buffer
268 strip width and TU measures (and pesticide concentrations) that were observed in this study,
269 additionally suggest that the site properties only affected TU measures marginally. This probably
270 reflects comparable site- and climatic- and agricultural (e.g. crop types and growth phases at the
271 time of the storm events) properties in the region.

272

273 4.2 Improving pesticide runoff predictability by adding buffer strip information

274 Applying high-resolution data, the runoff potential (RP) model successfully predicted the toxicity of
275 agricultural pesticides occurring in stream water during storm events. We found, however, that
276 adding a function for the minimum buffer strip width – within a two-sided 100 m corridor

277 extending 500 m upstream – to the RP model markedly improved the power of the model to predict
278 summed TUs from 46% to 64% by reducing the data variability around the regression line. The
279 slope and intercept of the regression line did not significantly change by adding the function for
280 minimum buffer strip width to the RP model, which reflects that the overall correlation between the
281 RP and summed TUs remains constant with or without buffer strip information. However, our
282 results clearly emphasise that minimum buffer strip width should be added to the model whenever
283 data is available, and furthermore underline the importance of considering buffer strip width in
284 upstream environments of stream sites potentially at risk of being impacted by agricultural
285 pesticides. Moreover, these findings lend support to Rasmussen et al. (2011) who were unable to
286 confirm the correlation between the RP and $SPEAR_{pesticides}$ that was found by Schriever et al.
287 (2007b) in German streams without buffer strips. Rasmussen et al. (2011) suggested that their
288 results were probably confounded by the presence of buffer strips surrounding the study streams.

289 No data was available in terms of tile drainage intensity for the fields surrounding the streams
290 that were examined in this study. However, loamy and clayey agricultural soils are often intensively
291 tile drained, and the tile drains serve as a direct route for pesticides from field to surface waters
292 underneath the buffer strip. Such sites have been found to be extremely vulnerable to pesticide loss,
293 especially if macropores have developed in the soil (Kronvang et al., 2004; Lewan et al., 2009;
294 Renaud and Brown, 2008). We therefore infer that incorporating information about tile drainage
295 conditions in the considered (sub-) catchment would further improve the predictive power of the RP
296 model.

297

298 4.3 Pesticide characteristics and their potential ecological impact

299 In this study, the summed toxic units (TU) based on storm flow water samples ranged from -6.63 to
300 -1.57. Applying the maximum TU for single pesticides did not significantly change this spectrum.

301 No pesticides were detected in any of the stream bed sediment samples taken in this study, which
302 could reflect too high detection limits and/or an inappropriate sampling technique. More strategic
303 sampling using a stationary suspended sediment sampler is proposed to further optimise the
304 detection success of adsorbed pesticides (Liess et al., 1996). However, Friberg et al. (2003) detected
305 several lipophilic pesticides adsorbed to bed sediments in Danish streams applying a technique
306 similar to the one used in the present study. An additional factor that potentially explains the
307 absence of pesticides in newly deposited bed sediments was the occurrence of several heavy
308 precipitation events during July, which could have reduced the residence time for the pesticides that
309 were adsorbed to fine particulate organic matter.

310 Nevertheless, the range of TUs measured in this study does have the potential to impair stream
311 ecosystems. Benthic macroinvertebrates have been shown to respond strongly to pesticide
312 contamination (Norum et al., 2010; Rasmussen et al., 2008; Schäfer et al., 2007), and they have
313 successfully been applied as indicator organisms for pesticide contamination in the recently
314 developed $SPEAR_{pesticides}$ index (Liess & von der Ohe, 2005). Applying the $SPEAR_{pesticides}$ index,
315 macroinvertebrate community changes have been observed at maximum TUs down to -3 in field
316 studies (Schäfer et al., 2011b). The recommended and currently applied threshold value
317 characterising good ecological status in the online SPEAR calculator (33% SPECies At Risk)
318 corresponds to a maximum TU value of -3 (see also
319 <http://www.systemecology.eu/SPEAR/calculator/index.php?lang=en>).

320 We found that the maximum TU and summed TUs concurrently exceeded the threshold value
321 for ecosystem effects in five streams representing more than 50% of the streams at risk of being
322 contaminated by agricultural pesticides (proportion of agriculture \geq 50% in the stream corridor).
323 Other anthropogenic stressors may be of higher importance than non-point pesticide contamination
324 (Rasmussen et al., 2011), but our results clearly emphasise that non-point source pesticide

325 contamination is a potential problem in small Danish streams. Not surprisingly, the insecticide
326 Pirimicarb represented the primary risk for benthic fauna due to its mode of action, which acts
327 selectively against this group of organisms. Fungicides having a less specific mode of action were,
328 additionally, relevant stressors for the benthic macroinvertebrates. Our findings are congruently
329 supported by a large body of evidence that identifies insecticides and fungicides as the primary
330 pesticide stressors directly impacting benthic macroinvertebrates in streams (see e.g. Liess et al.,
331 2005; Schäfer et al., 2007, 2011a; Schulz, 2004). In addition, we found that the herbicide,
332 Pendimethalin (inhibits mitosis), might also act as a potentially important stressor for benthic
333 macroinvertebrates.

334

335 4.4 Implications for stream management and the protection of stream ecosystems

336 The regression line in Fig. 2b represents the maximum TU as a function for minimum buffer strip
337 width; $Y = -6.586(\pm 0.681) + 6.235(\pm 1.24) * \exp(-0.249(\pm 0.105)x)$. Assuming that the relationship
338 is causative, the minimum buffer strip width necessary for obtaining good ecological status
339 (maximum TU ≤ -3), as required by the European WFD, is 6.6 metres. This is strongly contrasted
340 by present legislative requirements in Denmark where only natural streams or streams with a high
341 ecological objective (approximately 40% of the total stream network) are required to have 2 metres
342 of uncultivated buffer strips. The aim of buffer strips in Denmark is only to protect stream banks
343 from erosion, and pesticide application restrictions are currently enforced only via application
344 guidelines for specific compounds. The vast majority of Danish streams are therefore still
345 unprotected against pesticide contamination. However, considering the large variability in data
346 around the fitted regression and the preceding difficulties in predicting optimal dimensions for the
347 buffer strip retaining capacity, we recommend that the suggested minimum buffer strip width is
348 considered with care. Furthermore, Schäfer et al. (2007) detected very high maximum TUs in small

349 French streams that were flanked by buffer strips exceeding 11 metres. This could indicate that the
350 correlation between minimum buffer strip width and the TU obtained in this study is not applicable
351 for general extrapolation in time or space. However, the results of Schäfer et al. (2007) could be
352 confounded by intensive tile-drainage, as tile drains introduce an important transport route
353 underneath the vegetated buffer strips. Only few authors have attempted to describe the dimensions
354 that buffer strips should have for optimum performance in terms of pesticide retention (Johnson et
355 al., 2007), probably reflecting the numerous highly variable factors influencing pesticide runoff,
356 including timing and volume of rainfall events occurring subsequent to pesticide application, buffer
357 strip vegetation types and growth phases, soil infiltration capacity, soil moisture and runoff velocity
358 (Klöppel et al., 1997, Lacas et al., 2005; Pot et al., 2005). Depending on the site characteristics,
359 climatic conditions and local pesticide application practices, optimal buffer strip width change. As a
360 consequence, buffer strips wider than 6.6 metres could be necessary for sufficient protection of
361 stream ecosystems from pesticide surface runoff, as it has also been found for different phosphorus
362 forms and other pollutants (Hoffmann et al., 2009; Mander, 2005; Uusi-Kämpö, 2005).

363

364 **5. Conclusions**

365 The minimum width of buffer strips in the near upstream area was found to be the most important
366 environmental parameter governing measured summed and maximum TUs in Danish streams. This
367 suggests that the prevalence and dimensions for buffer strips currently required by Danish
368 legislation is, in general, far from sufficient in protecting stream ecosystems from non-point source
369 pesticides. Despite the fact that small streams with catchment sizes under 10 km² are disregarded
370 within the European WFD (European Commission 2000), we believe it is still essential to protect
371 the upper branches of streams with buffer strips especially since these systems serve as sources for

372 recolonisation to the reaches further downstream (targeted in the WFD). Providing such sources
373 would add some valuable recovery capacity to the stream ecosystems.

374 Adding a function for minimum buffer strip width to the Runoff Potential (RP) model
375 improved its power to predict summed Toxic Units in the study streams from 46% to 64% without
376 changing the slope or intercept of the regression line. This underlines the importance of considering
377 buffer strip dimensions in the near upstream area within the risk assessment procedure. Using high-
378 resolution data (including buffer strip dimensions) the RP model was found to be a useful screening
379 tool for the identification of stream sections at risk for pesticide contamination. However, we
380 suggest that pesticide transport from agricultural catchments to streams via tile drain flow would
381 further improve the predictive power of the model. Future research should address these
382 shortcomings of the model.

383

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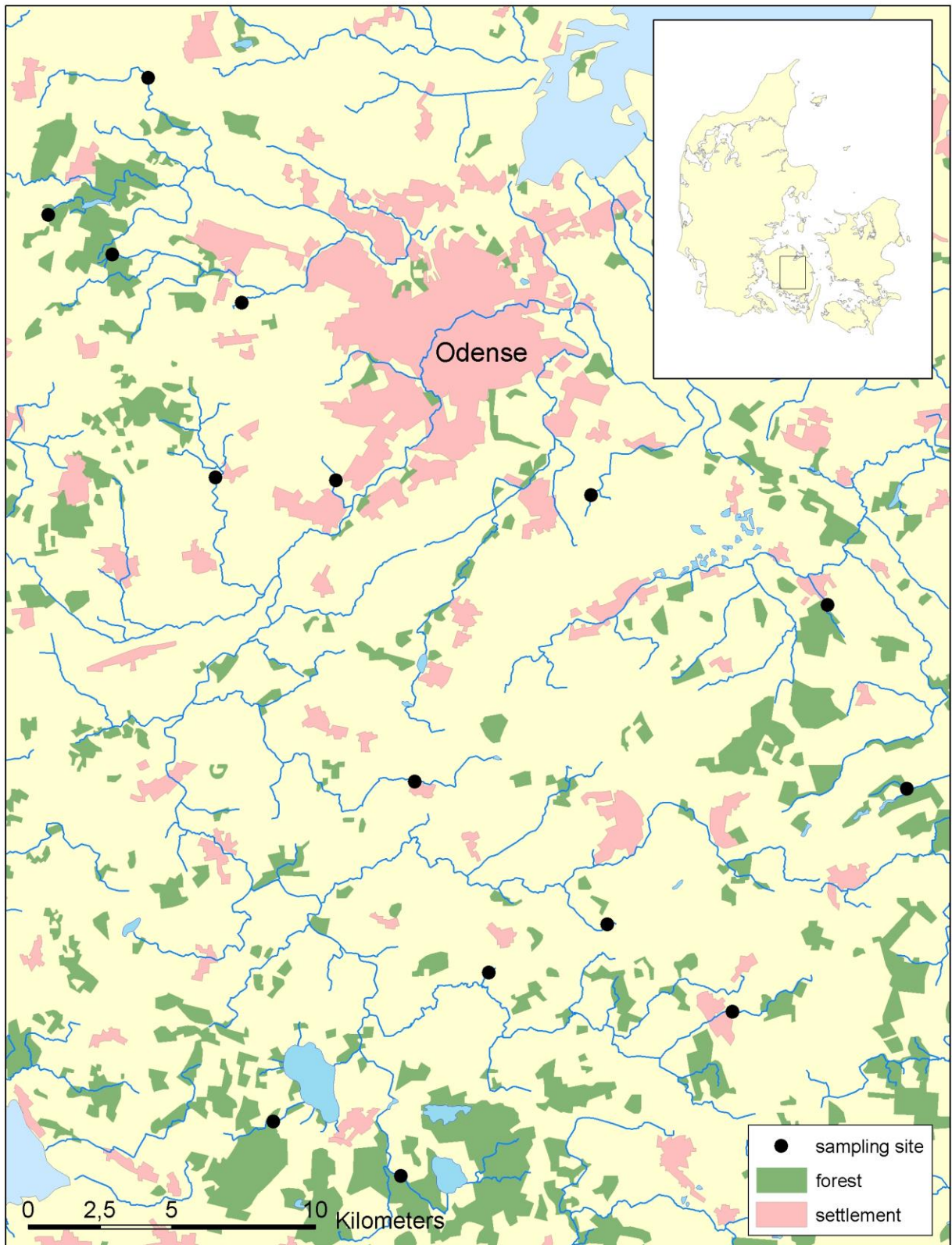
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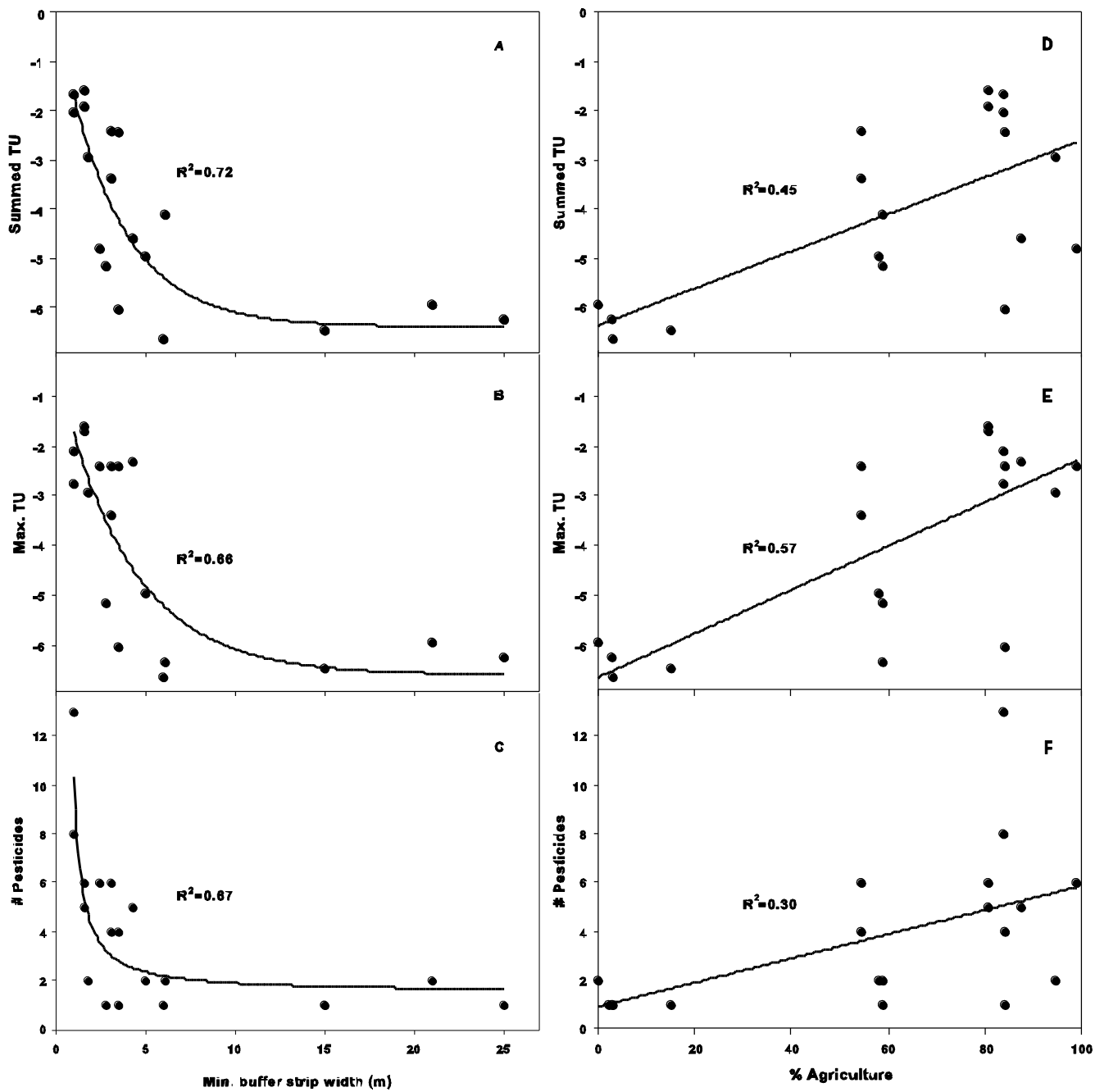
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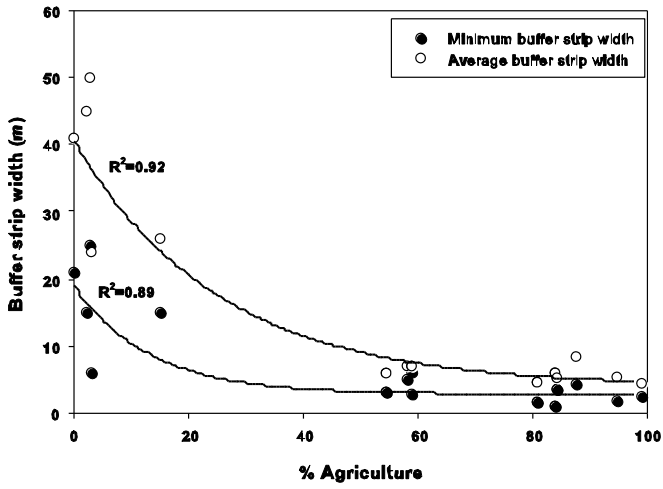
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Fig. 1: Schematic map of the 14 study stream locations.

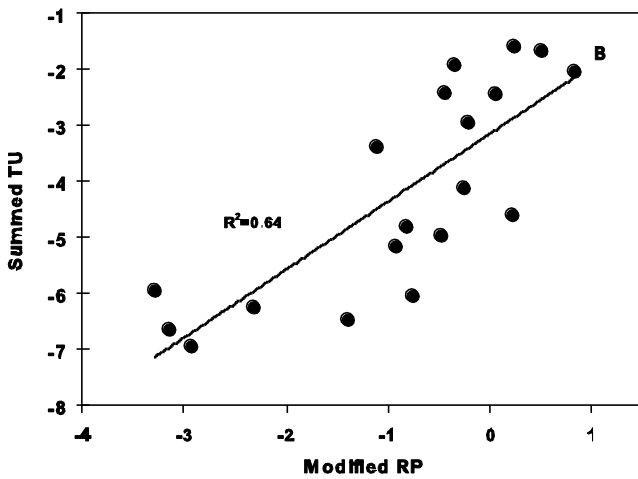
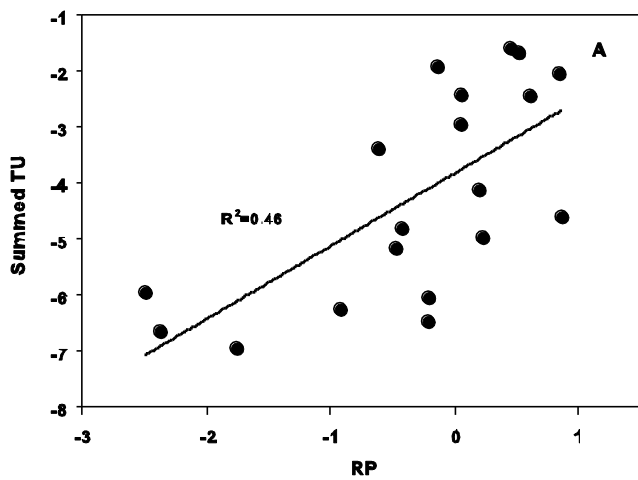


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Fig. 2: The summed TU of all pesticides (A and D), the maximum TU (B and E) and the total number of pesticides (C and F) as a function for minimum buffer strip width and the proportion of agricultural land (D, E and F, respectively). Presented data is based on water samples collected during storm flow conditions (two storm flow events) in 14 Danish streams in spring, 2009.



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 502 Fig. 3: Proportion of agricultural land as a function of minimum (●) and average (○) buffer strip
 503 width. Data represent 14 Danish low-order streams.



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Fig. 4: The RP (A) and a modified version of RP (additionally considering minimum buffer strip width) (B) as a function for summed TU. The RP and modified RP were based on a series of environmental parameters deriving from a 2x100 m stream corridor extending 500 m upstream from the sampling points. Pesticide concentrations were measured during two storm flow events in 14 Danish low-land streams in spring, 2009.

512 Table 1: Pesticides detected in stream water from 14 Danish streams in the period from April to
 513 August, 2009. Three samples were collected in each stream of which two were collected with event-
 514 triggered samplers during May and June (high precipitation events), and one sample was collected
 515 manually during base-flow conditions in August. Pesticide groups are indicated by letters H, F and I
 516 representing herbicides, fungicides and insecticides, respectively.
 517

Compound	Min concentration (ug/L)	Max concentration (ug/L)	Highest TU ^a	Detection frequency (%)
Desethylterbutylazine (H)	0.01	0.11	-4.65	100
Atrazine (H)	0.01	0.02	-6.63	7
Dimethoate (H)	0.01	0.18	-4.05	14
Metachlor (H)	0.01	0.05	-5.82	57
Diflufenican (H)	0.02	0.15	-3.20	29
Metamitron (H)	0.12	0.12	-4.68	7
Pendimethaline (H)	0.02	0.97	-2.46	14
Aclinofen (H)	0.14	0.14	-3.93	7
Propyzamide (H)	0.01	0.43	-4.11	21
Prosulfocarb (H)	0.01	0.07	-3.86	21
Terbutylazine (H)	0.01	0.6	-4.55	57
Hexazinone (H)	0.06	0.06	-6.15	7
Simazine (H)	0.03	0.03	-4.56	7
Boscalid (F)	0.07	0.72	-3.87	36
Azoxystrobin (F)	0.05	0.51	-2.77	43
Propiconazole (F)	0.04	0.27	-4.58	43
Tebuconazole (F)	0.02	0.24	-4.24	50
Dimethomorf (F)	0.01	0.08	-5.12	14
DEET (I)	0.05	0.05	-6.18	7
Pirimicarb (I)	0.01	0.32	-1.72	21

518 ^a Based on LC50 values for 48h acute toxicity tests with *Daphnia magna* (Tomlin, 2001)
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