

LEARNING FROM EXPERIENCE – PESTICIDE MONITORING IN ENGLISH GROUNDWATER

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Hypothesis

- The occurrence of pesticides in groundwater can be predicted by generalised linear modelling

$$G = f(\text{site}) + f(\text{compound}) + f(\text{site} \times \text{compound}) + \text{error}$$

What endpoint
to predict to?

The vulnerability

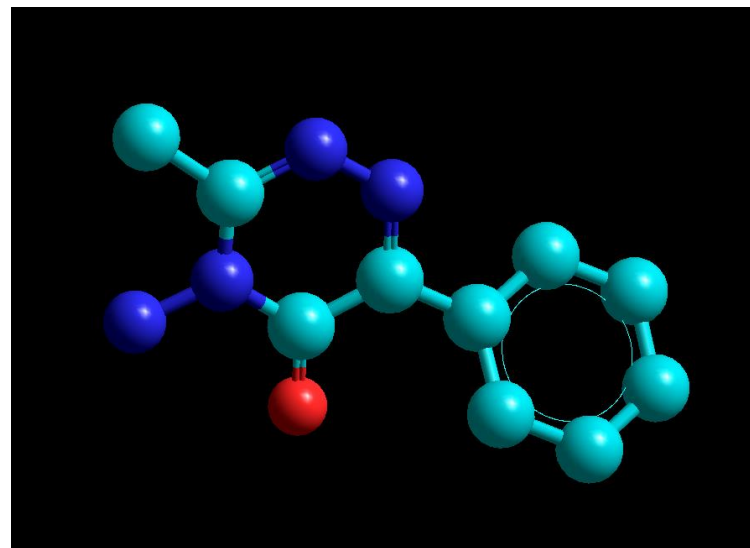
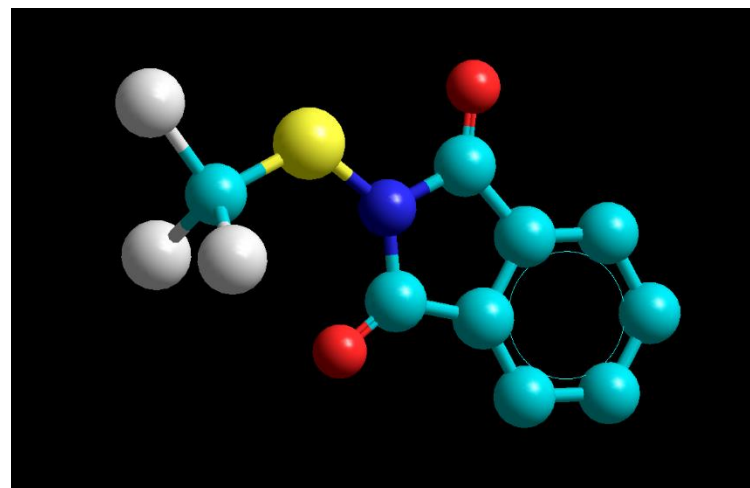
The hazard

Interaction – some sites
are less vulnerable to some
compounds than others

What distribution?

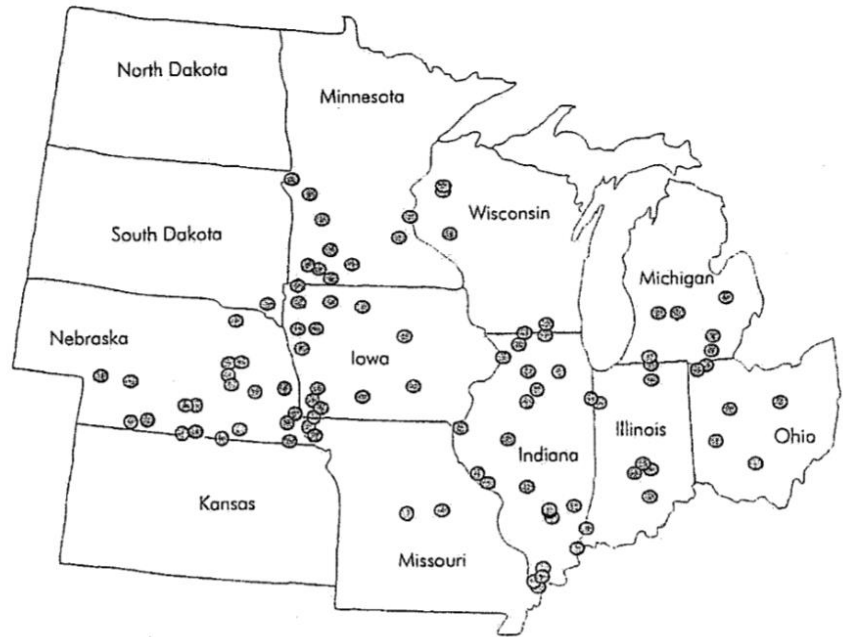
The Hazard – f(compound)

- Use properties only dependent upon the molecule
 - K_{oc} and $t_{1/2}$ are no good
- Model using molecular descriptors
 - Connectivity
 - Atom and group counts
 - Molecular orbitals
 - Hydration energy
 - Dipole moment
 - Refractivity

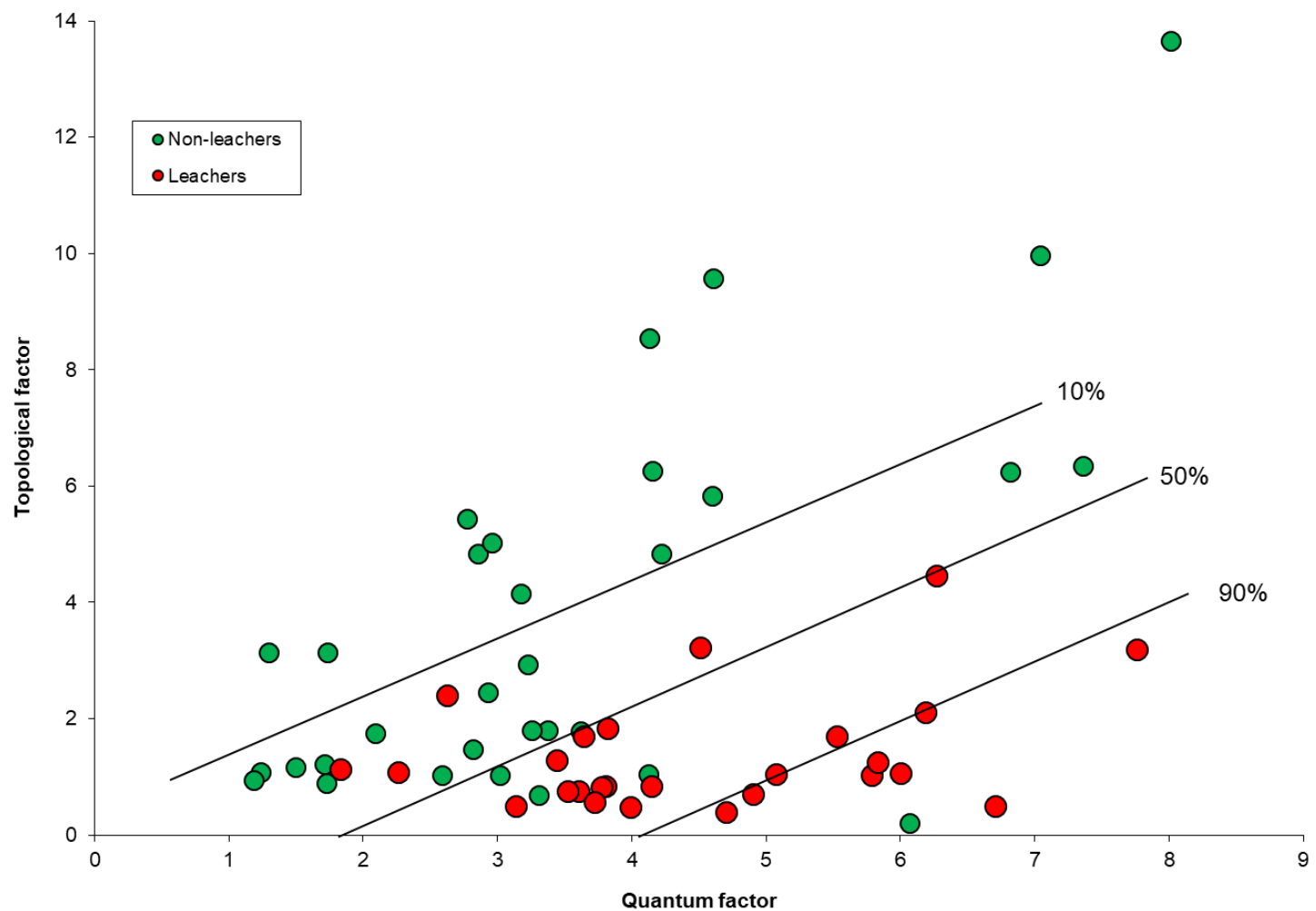


The Hazard – f(compound) – US data

- Great Valley, California
 - 40 compounds
- Kolpin et al. (1997)
 - 56 compounds
 - 303 boreholes
 - 12 US states
- 61 compounds, 27 compounds in common
- Logistic model
 - Bernoulli trial
 - Modelling detected vs. non-detected



The Hazard – f(compound) – molecular model



The Hazard – f(compound) – molecular model

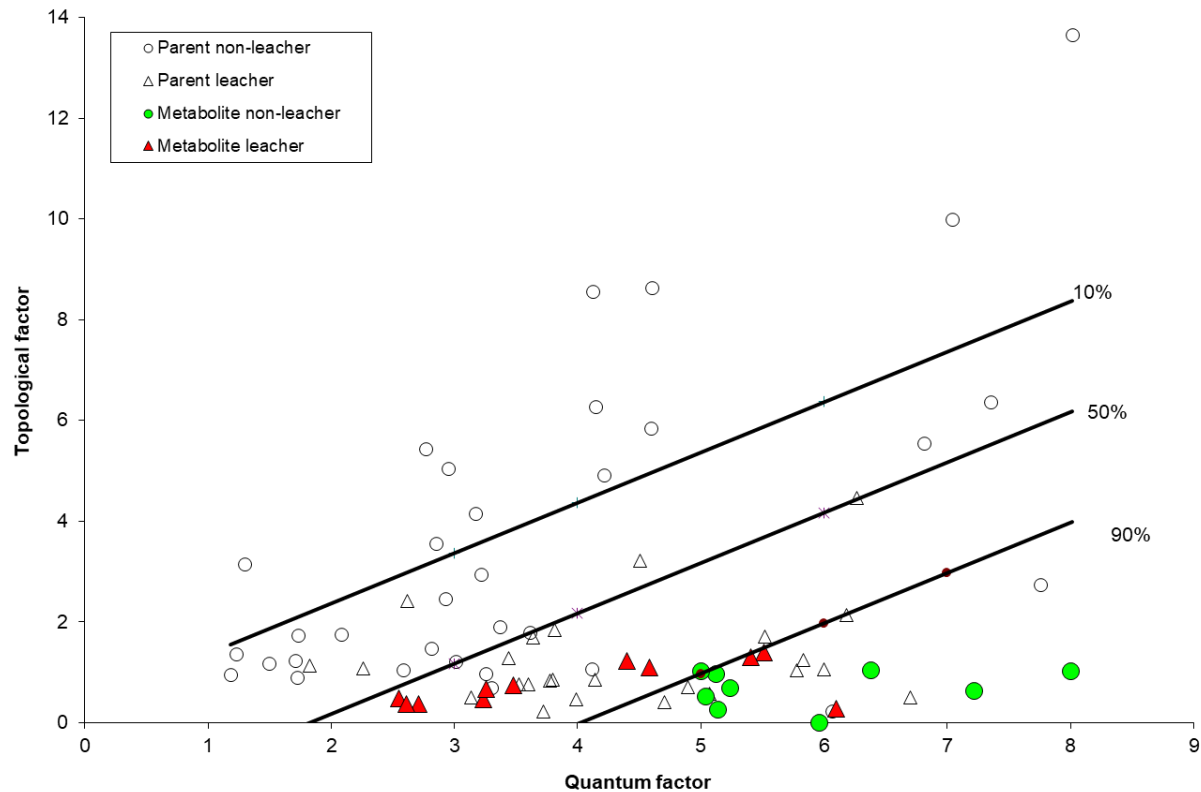
- The probability of a compound being detected (θ)

$$\log\left(\frac{\theta}{1-\theta}\right) = \underbrace{0.77\mu - 0.185 \Delta H_{\text{hyd}}}_{\text{Quantum factor - Solubility}} - \underbrace{2.215 {}^6\chi_p^v - 53 {}^7\chi_{pc}^v}_{\text{Topological factor - degradation?}} - 1.27$$

- Correctly classifies 91% of compounds
- 50% probability of being detected if:

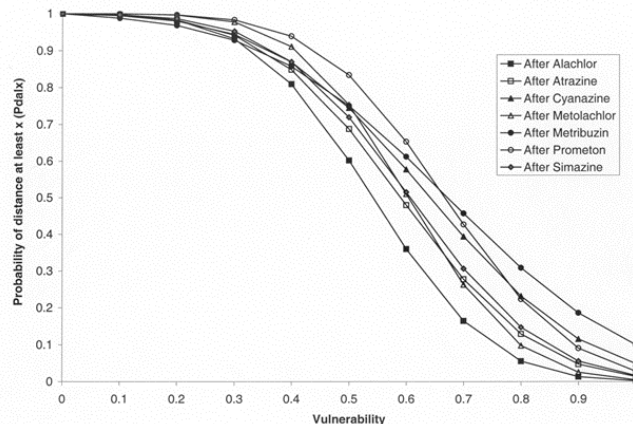
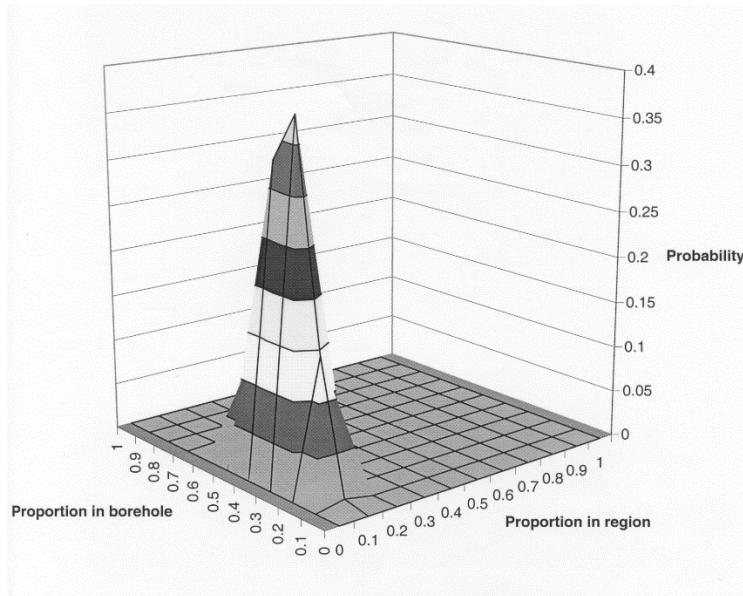
$$0.28 \mu < {}^6\chi_p$$

The Hazard – f(compound) – metabolites



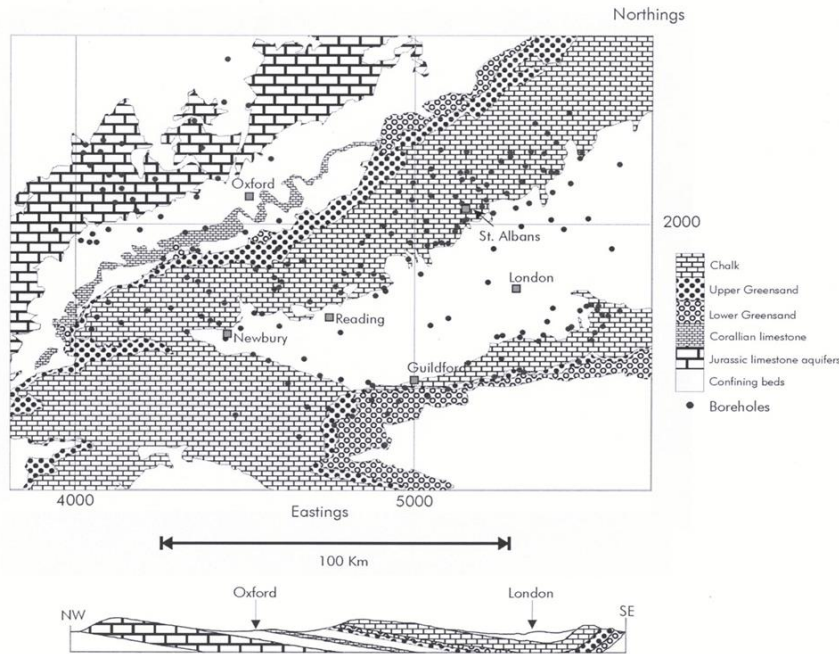
- The mobile daughter of immobile parent or the immobile daughter of a mobile parent

Vulnerability – f(site)



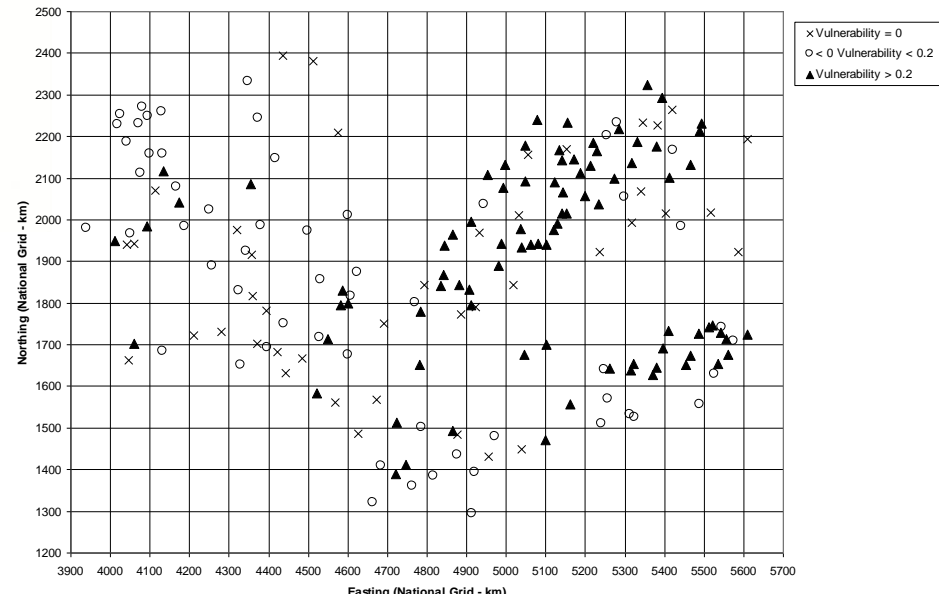
- A vulnerable site is one which shows more detects of pesticides than would be expected
- Bayesian analysis of proportions
 - Prior distribution is the proportion of detects in the region
 - Pesticide analysis seen as a Bernoulli trial
 - Updates with each compound
 - Independent of compound
 - Probability of detection of the next compound
 - Comes with uncertainty

Vulnerability – f(site) – Thames basin

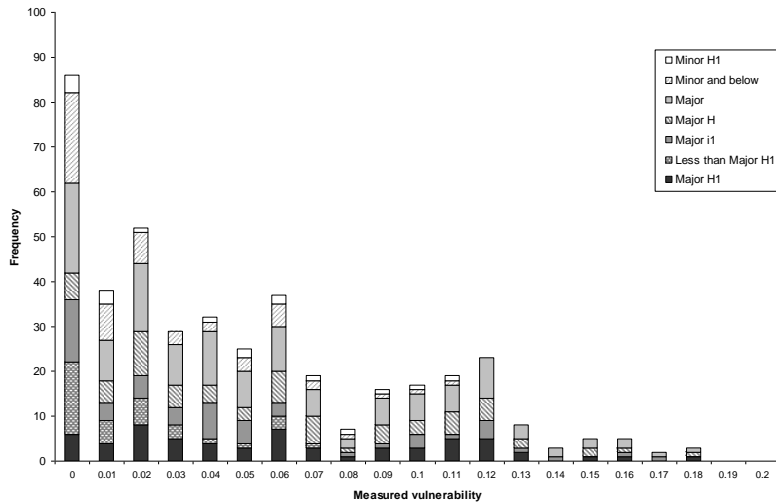


Thames region EA has best pesticide dataset in UK

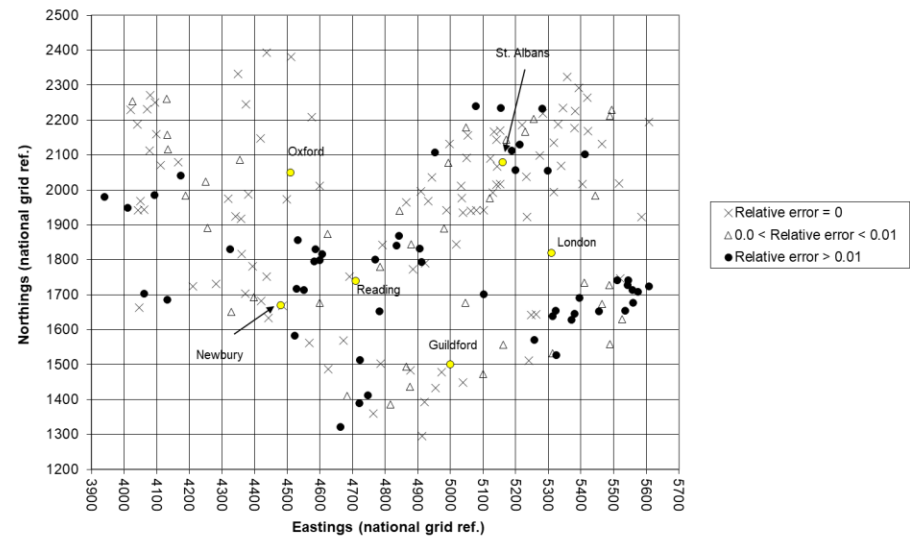
- 359 boreholes
- 27 compounds



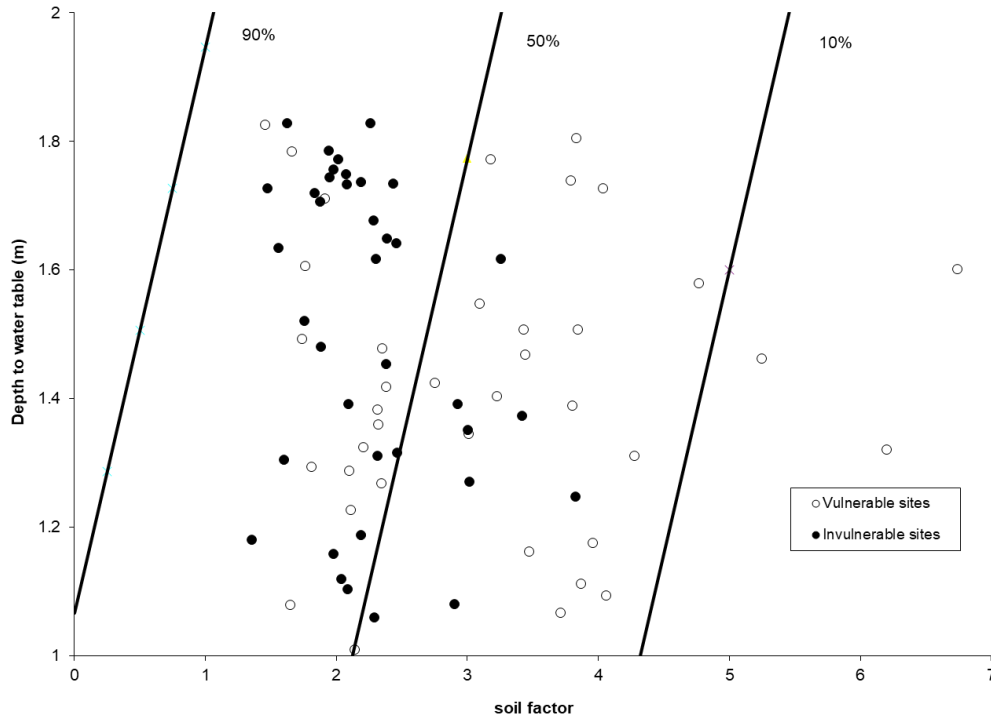
Vulnerability – f(site)



- Can be compared to existing groundwater vulnerability
- They do not match well
- The uncertainty can be plotted



Vulnerability – f(site) – US dataset



■ What controls vulnerability?

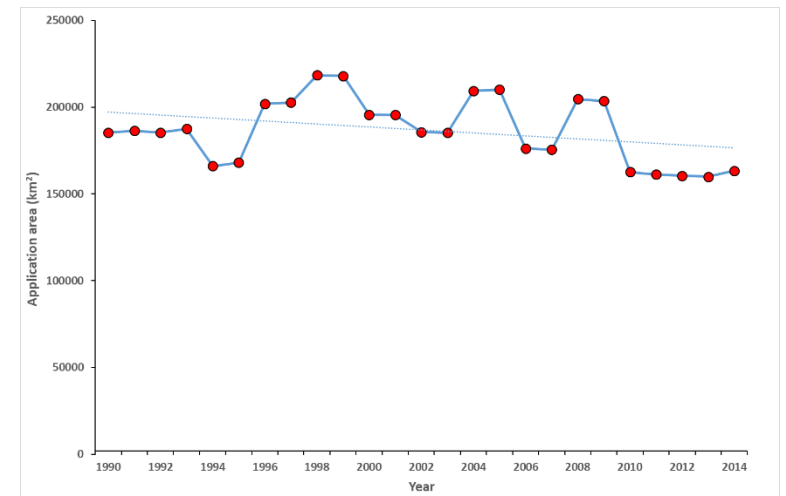
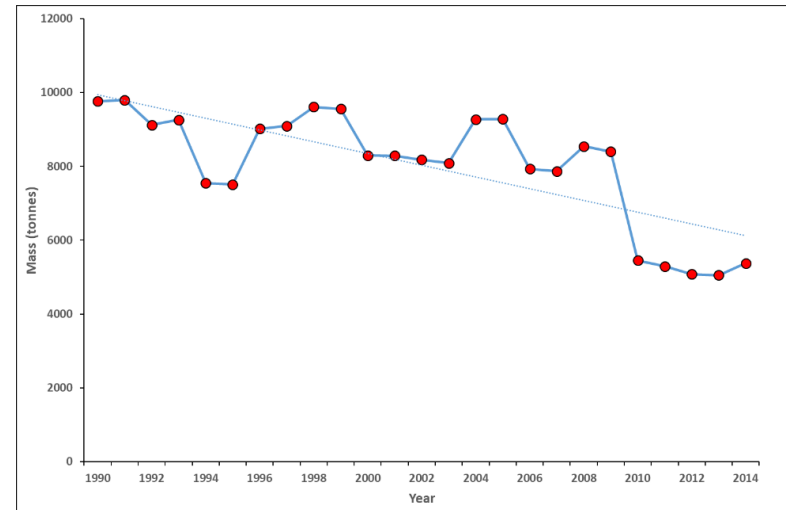
- Compare detection to catchment properties

■ Probability of detecting a compound (θ)

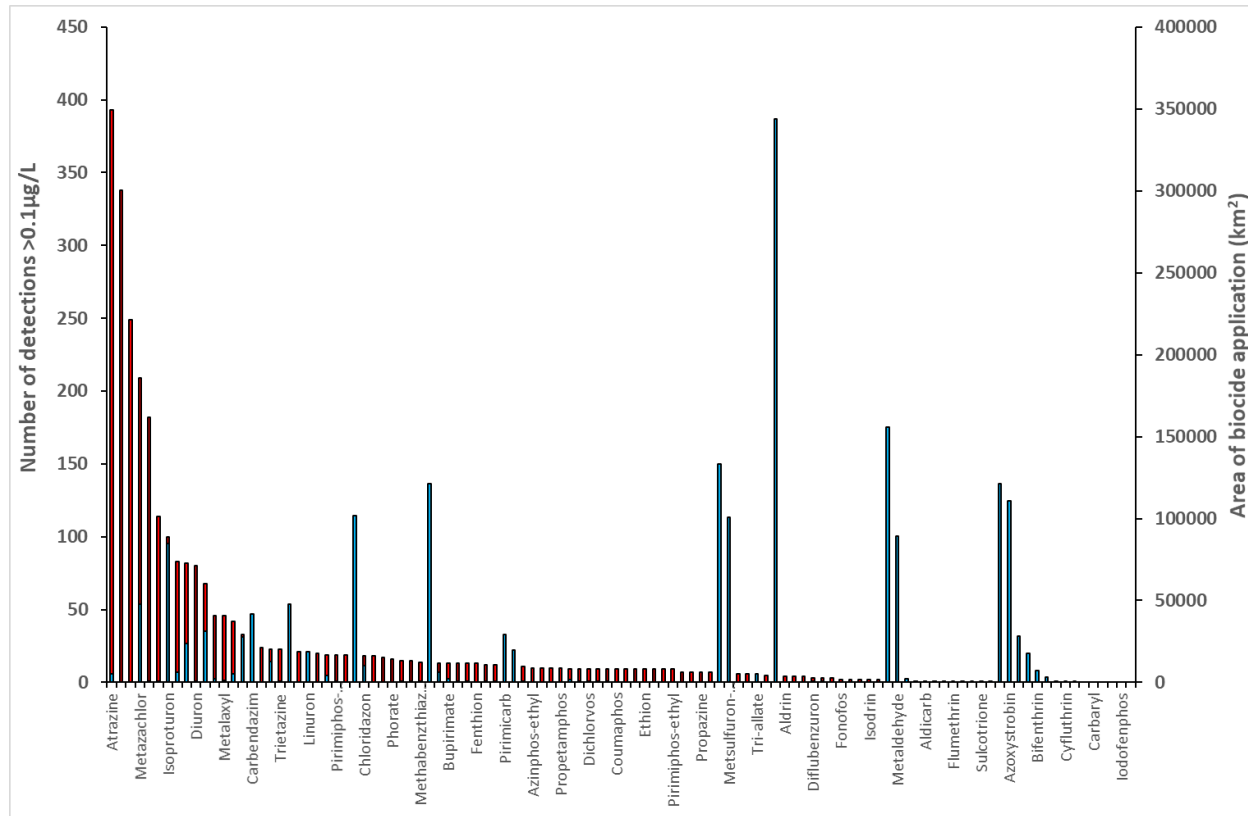
$$\ln\left(\frac{\theta}{1-\theta}\right) = 0.985 - 0.695\%OM - 0.03\%sand + 1.135wtdepth$$

English groundwater

- EA groundwater data for all English groundwater
 - Data from 2005 to 2017
 - 113 compounds
 - 3357 unique borehole locations
 - > 1.5 million observations
- Pesticide usage statistics available
 - Pesticide area and amount
 - 68300 tonnes applied over study period
 - Significant decline in total amount, total area and average application rate

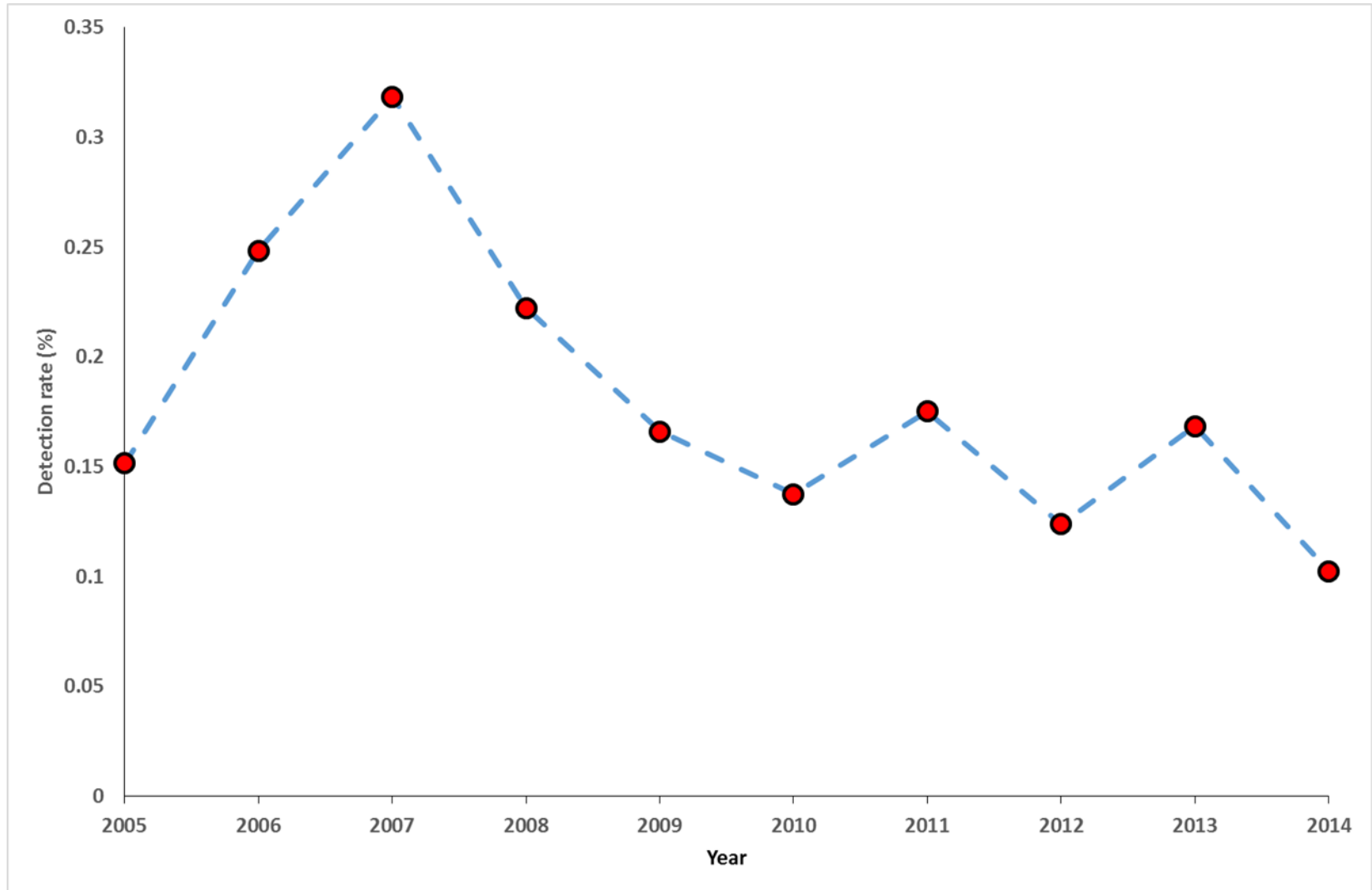


English groundwater – Detection rate

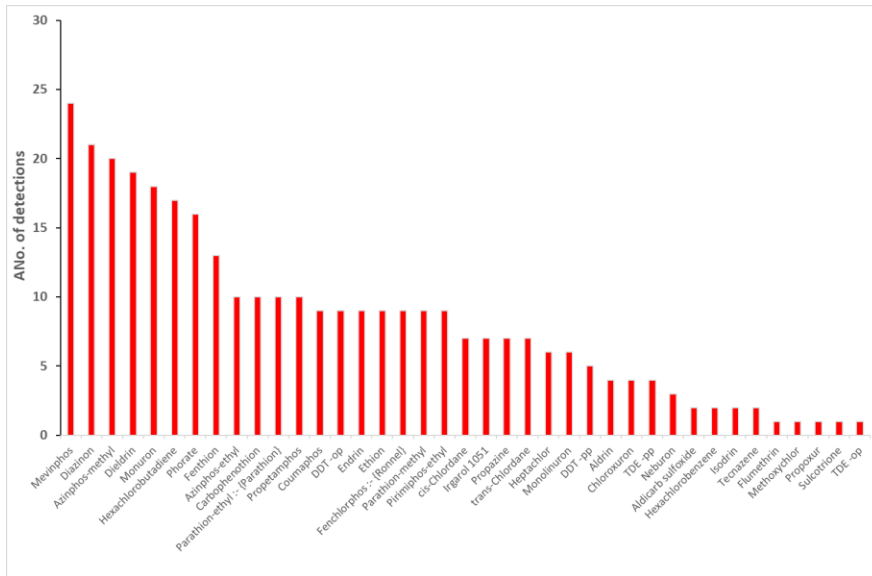


- 98 out 113 compounds detected above 0.1 µg/l
- Average detection rate was 0.19%
- Detection rate declined over the period

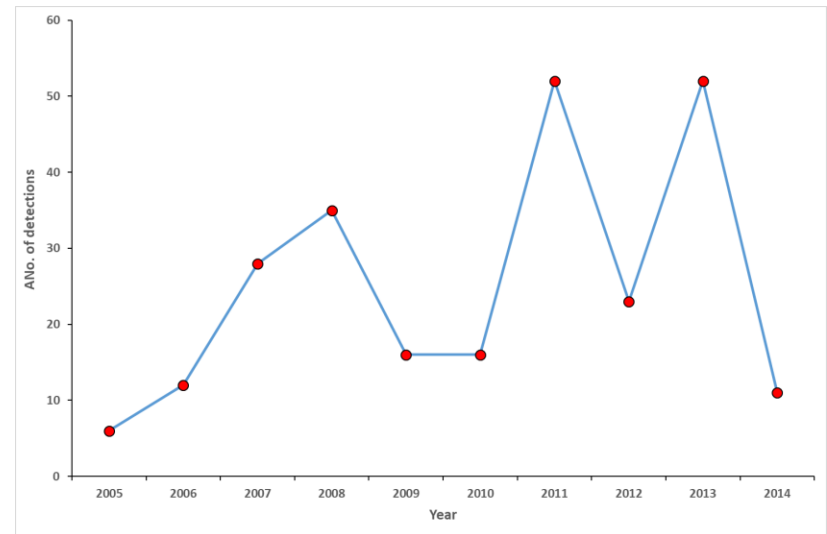
English groundwater – Detection rate



English groundwater – legacy detection rate



- 38 of the 98 detected were never applied during the period
- 11.4% of all detections were these legacy pesticides
- The detection rate for these did not change



English groundwater – Detection rate

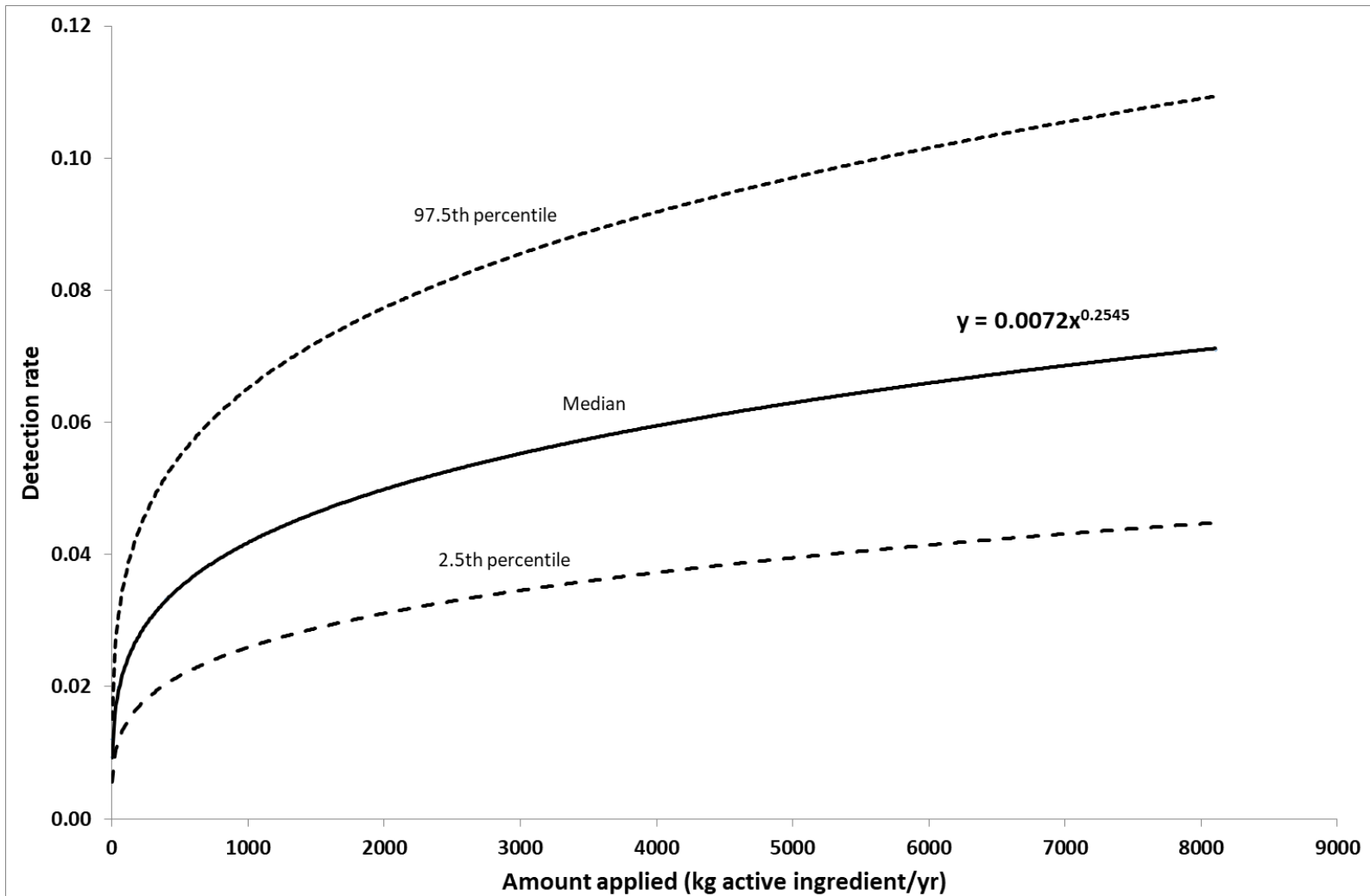
Activity Solubility – detection increases Solubility – detection increases HOMO/LUMO gap – detection increases

$$g(\theta) = 0.26Amount + 0.12\Delta H_{hyd} + 0.25\eta + 0.16LUMO - 0.44HOMO$$

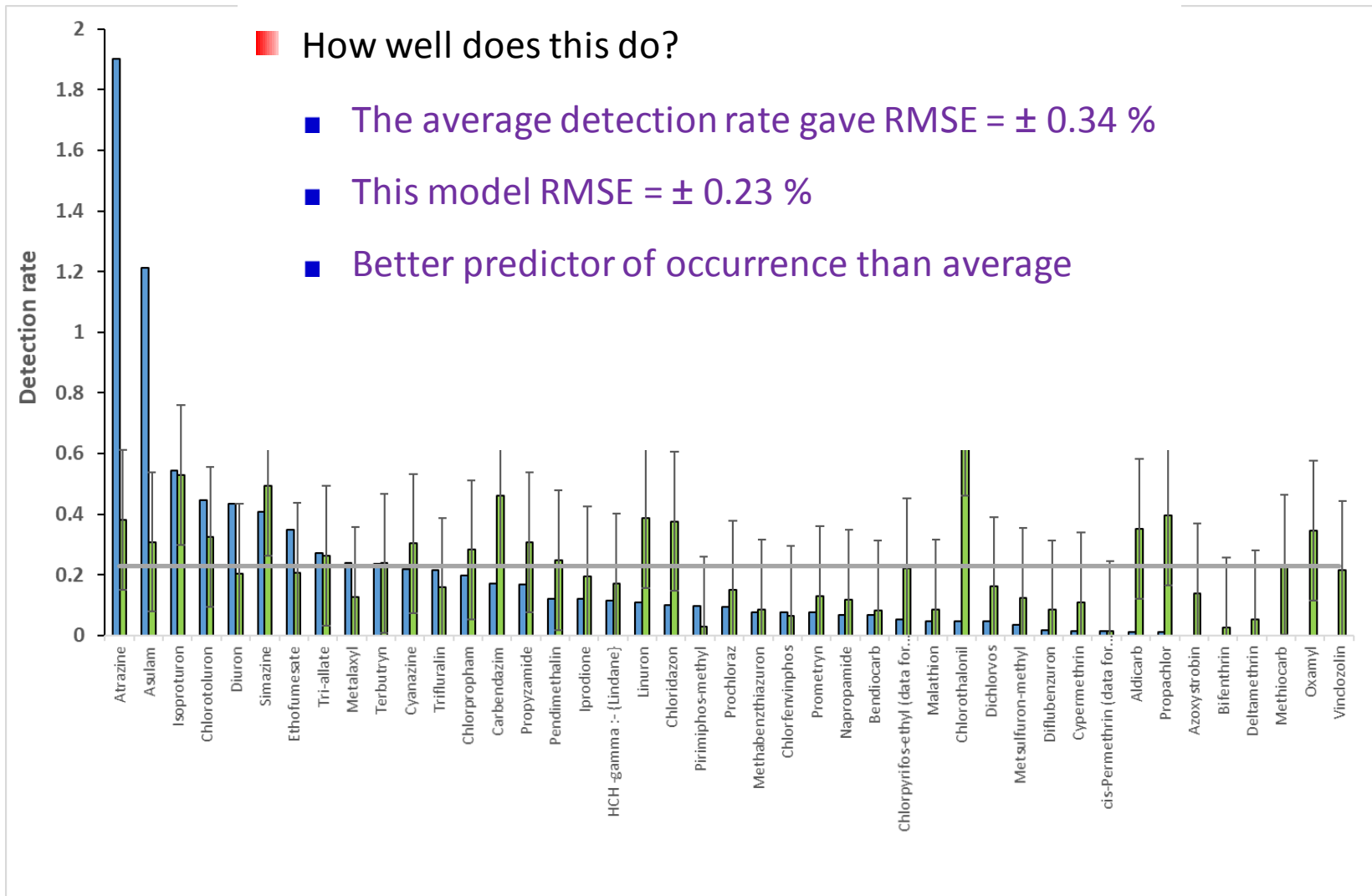
rate of detection = bin(θ , no. of observations)

- What controls detection rate?
 - Detection rate not detection
 - Binomial regression
 - Loading
 - Solubility
 - degradability

English groundwater – Detection rate

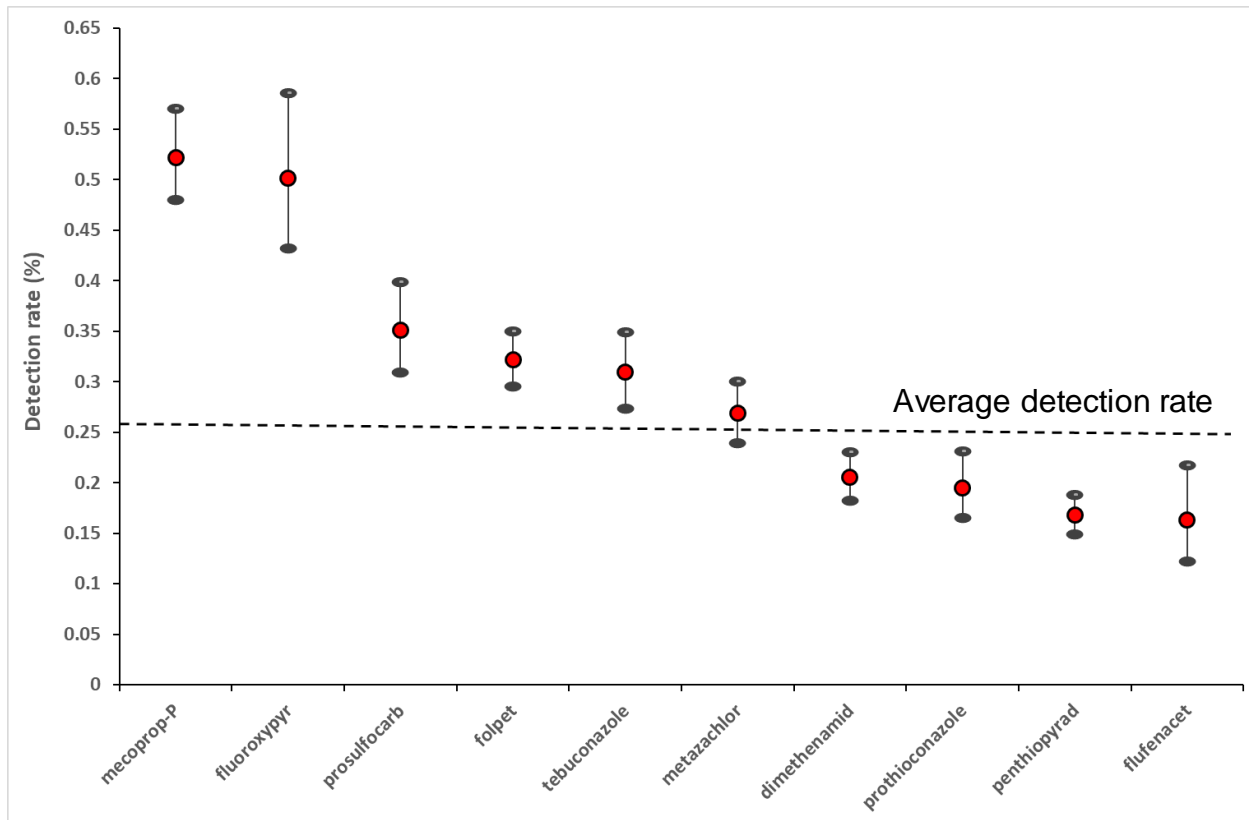


English groundwater – model performance

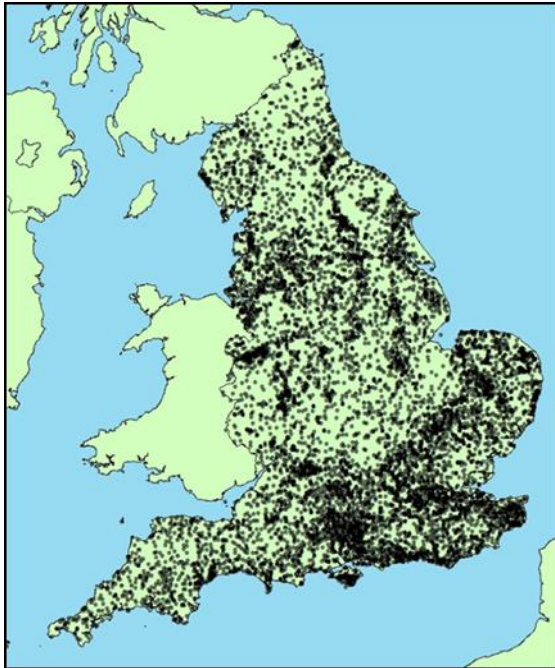


English groundwater – application

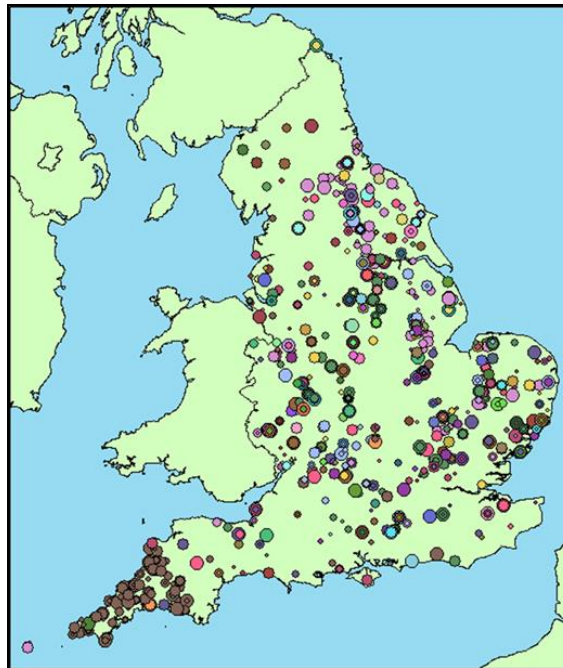
- Apply the model to the most used compounds never analysed for
 - More modern compounds are not necessarily less polluting



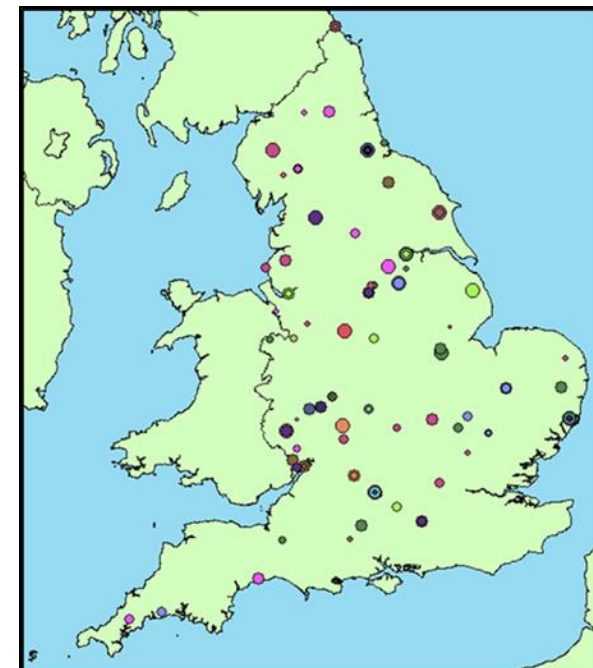
English groundwater - Vulnerability - f(site)



■ Locations of sampled boreholes



■ Locations of boreholes with detections

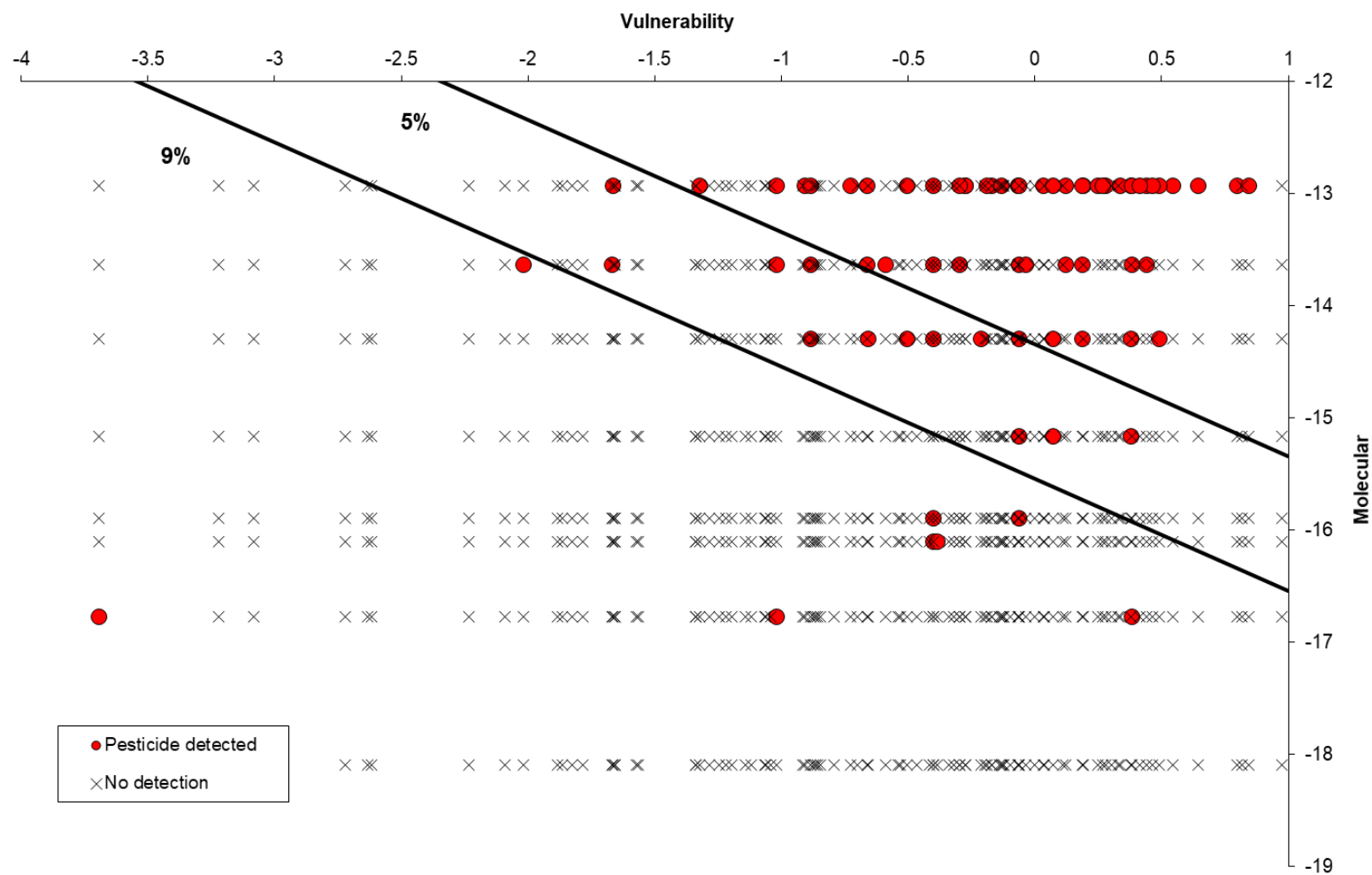


■ Locations of boreholes with legacy detections

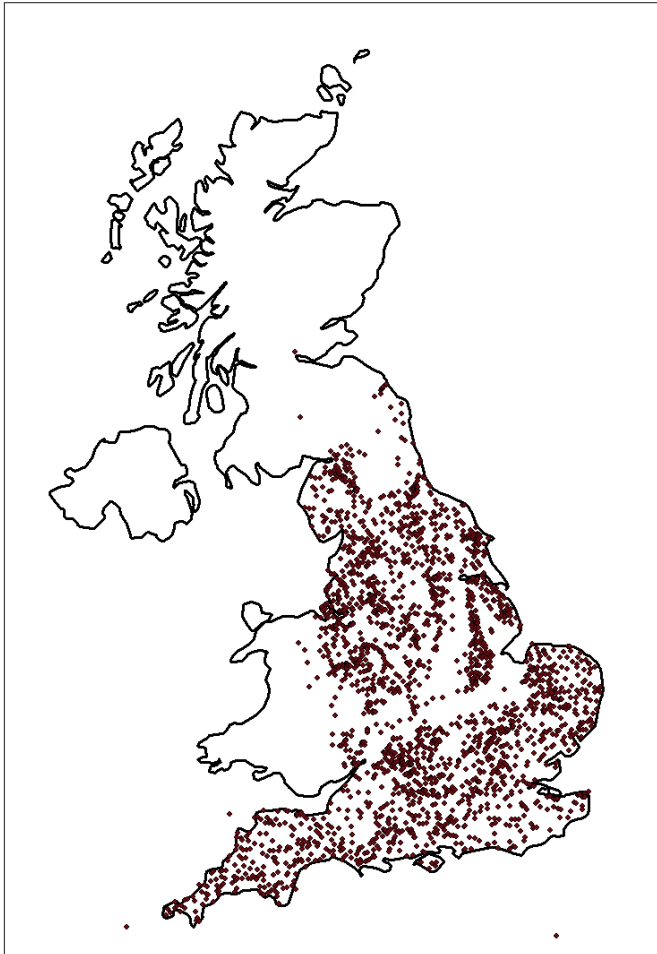
■ Not got to this yet.

■ Average 456 observation per borehole (1 to 3030)

Interaction – f(site x compound)



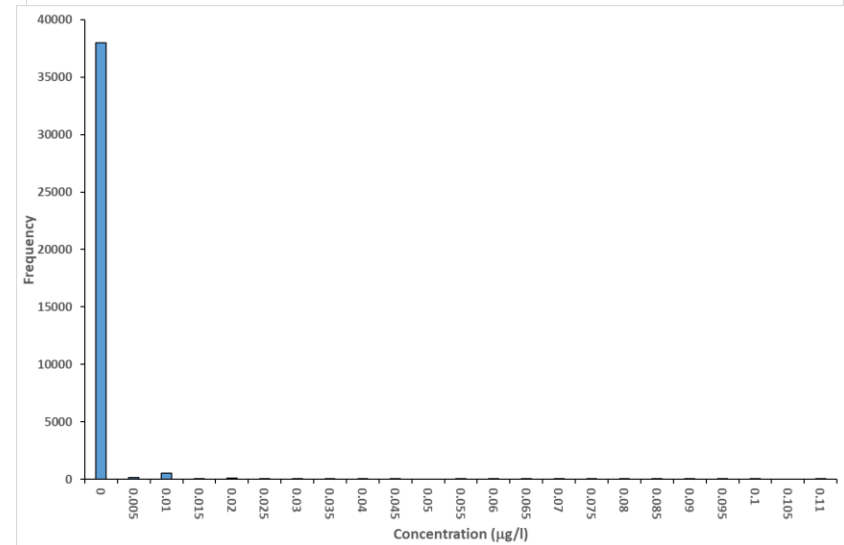
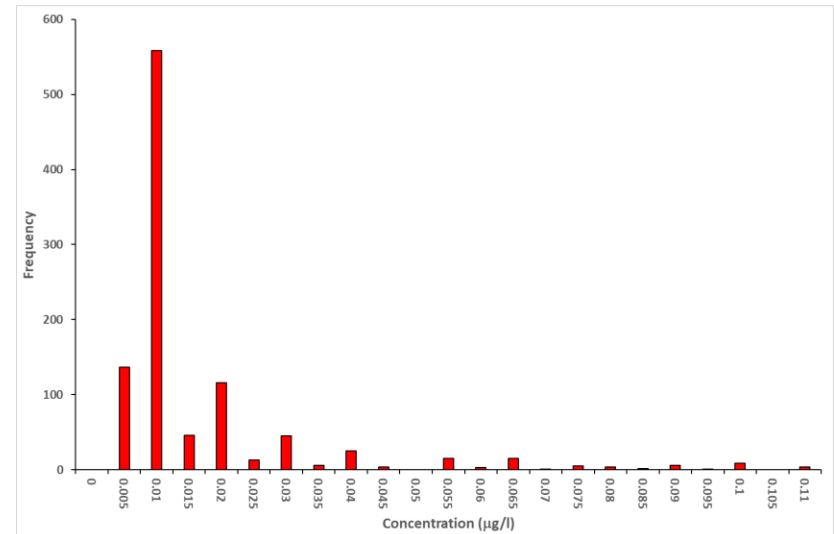
English groundwater – validation and next steps



- Independent dataset
- Very low detection
- Screens for very wide range of compounds
- EA has been using LC-MS on groundwater since 2009
 - 2415 locations
 - 39055 observations

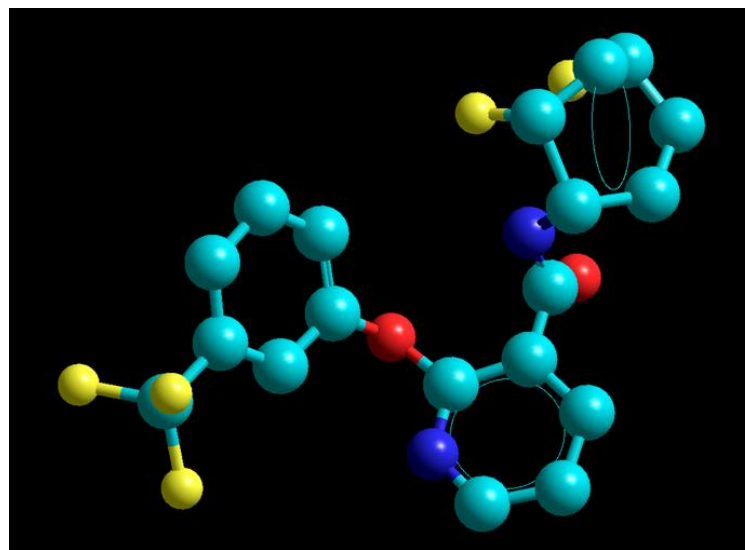
Future directions

- Tracking legacy compounds
 - Real persistence in the environment
- We want to predict concentration
- Distribution is dominated by zeros
 - Even allowing for truncated and censored data
 - But zero is a reasonable result
- Two part model
 - Logistic model
 - Weibull model



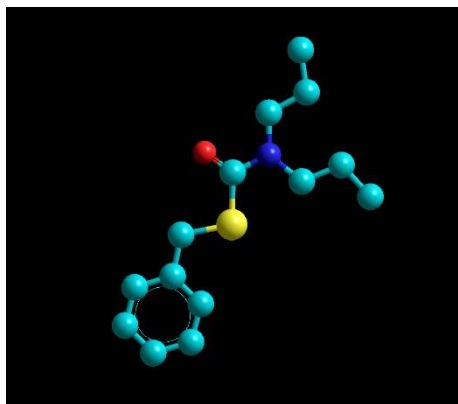
Conclusions

- All achieved on free data
- The developing hierarchy of generalised linear models
 - Logistic models
 - Molecular model
 - Vulnerability model
 - Interaction model
 - Binomial regression
 - Molecular model
- Models have physical interpretation
 - Activity
 - Solubility
 - degradation

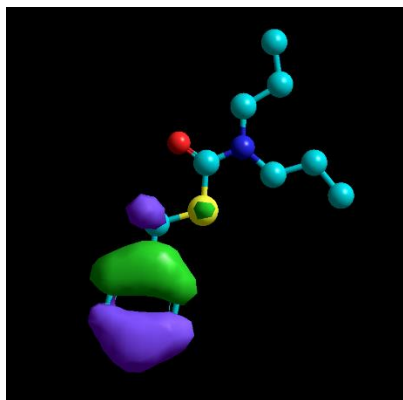


Could we do this?

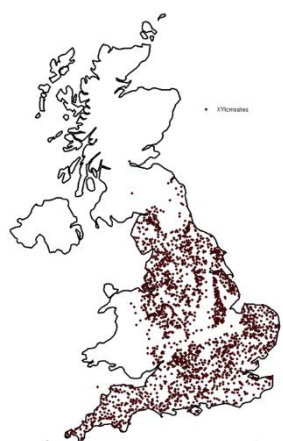
New compound



Molecular model



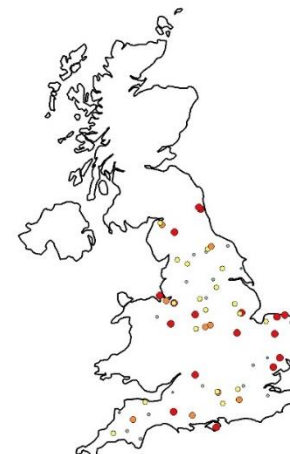
Screened out



Screening locations



Vulnerability model



What concentration? Where?