

FORMULATION TOXICITY

ITT5

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PROBLEM RECAP

- Agrochemical formulations are mixtures of active ingredients and various coformulants.
- Toxicity of substance is measured by animal testing.
- If we can model toxicity of formulations/ingredients, we may be able to reduce animal testing.



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- Other ingredients have toxicity in a given range, or may be discrete (e.g. “irritant”/“non-irritant”/“severely irritant”).
- Inverse toxicity of a mix is often assumed to be sum of ingredient inverse toxicities (additivity). How can we check for non-additive interactions?

PROBLEM RECAP



STEP 1: Derive Acute Toxicity Estimates (ATE) of individual ingredients



STEP 2: Calculation of ATE of a mixture and use classification criteria in the left table.

Exposure routes	Classification category or experimentally obtained acute toxicity range estimate (see Note 1)	Converted acute toxicity point estimate (see Note 2)
Oral (mg/kg bodyweight)	0 < Category 1 ≤ 5	0.5
	5 < Category 2 ≤ 50	5
	50 < Category 3 ≤ 300	100
	300 < Category 4 ≤ 2000	500
	2000 < Category 5 ≤ 5000	2500
Dermal (mg/kg bodyweight)	0 < Category 1 ≤ 50	5
	50 < Category 2 ≤ 200	50
	200 < Category 3 ≤ 1000	300
	1000 < Category 4 ≤ 2000	1100
	2000 < Category 5 ≤ 5000	2500
Gases (ppmV)	0 < Category 1 ≤ 100	10
	100 < Category 2 ≤ 500	100
	500 < Category 3 ≤ 2500	700
	2500 < Category 4 ≤ 20000	4500
	Category 5 - See footnote to 3.1.2.5.	
Vapours (mg/l)	0 < Category 1 ≤ 0.5	0.05
	0.5 < Category 2 ≤ 2.0	0.5
	2.0 < Category 3 ≤ 10.0	3
	10.0 < Category 4 ≤ 20.0	11
Category 5 - See footnote to 3.1.2.5.		
Dust/mist (mg/l)	0 < Category 1 ≤ 0.05	0.005
	0.05 < Category 2 ≤ 0.5	0.05
	0.5 < Category 3 ≤ 1.0	0.5
	1.0 < Category 4 ≤ 5.0	1.5
Category 5 - See footnote to 3.1.2.5.		

Ingredient(s) with unknown toxicity is ≤ 10 %;

$$\frac{100}{ATE_{mix}} = \sum_n \frac{C_i}{ATE_i}$$

Ingredient(s) with unknown toxicity is > 10 %;

$$\frac{100 - (\sum C_{unknown \text{ if } > 10\%})}{ATE_{mix}} = \sum_n \frac{C_i}{ATE_i}$$

C_i = concentration of ingredient i ;

n ingredients and i is running from 1 to n ;

ATE_i = Acute toxicity estimate of ingredient i ;

Additivity motivates the following

$$\mathbf{y} = \mathbf{A}\mathbf{x} + \epsilon$$

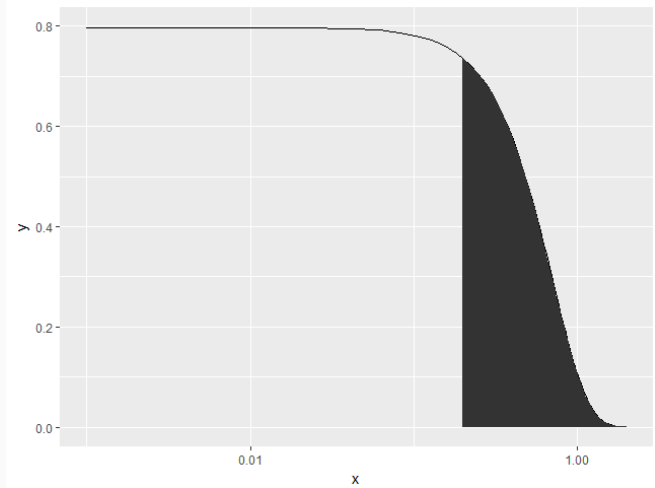
- \mathbf{y} - Inverted formulation toxicities
- \mathbf{A} - Matrix with rows containing ingredient proportions
- \mathbf{x} - Inverted ingredient toxicities
- ϵ - Measurement error

Bayesian inference of \mathbf{x} incorporates prior knowledge and gives confidence in the result.

PREDICTING CLASSIFICATION

E.g. How confident can we be that an ingredient is Category 1 (< 5 mg/kg) in acute toxicity?

Simply integrate posterior over category interval:



- What's the source of uncertainty in results?





- What's the source of uncertainty in results?
- If we can deduce this, could reduce animal trials by testing chemicals that are responsible for uncertainty.



WORKING WITH DISCRETE DATA

E.g. skin toxicity may simply be “Severely irritant” / “Irritant” / “Non-irritant”.

How can we incorporate this discrete data into the linear model?

Class/Category	Serious Eye Damage - Category 1	Eye Irritation - Category 2A	Eye Irritation - Category 2B
Pictogram			(no pictogram)
Signal word	Danger	Warning	Warning
Hazard statement	Causes serious eye damage.	Causes serious eye irritation.	Causes eye irritation.

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One possibility is a hidden / latent numerical toxicity in which categories are clustered.

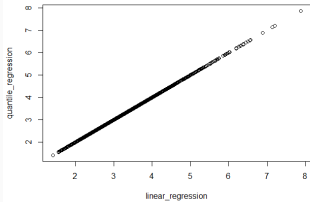
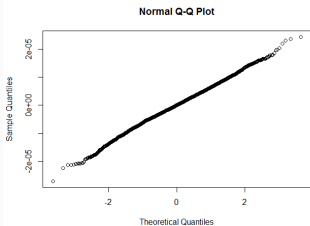
$$x_i \sim N(\mu_{C_i}, \sigma_{C_i}^2)$$

i.e. supervised learning but without direct observations of x_i .

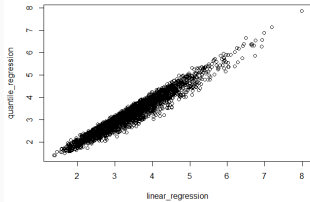
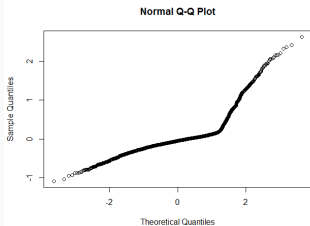
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DETECTING NON-ADDITIVITY

Linear Dataset



Nonlinear Dataset



The problem requires methodology for the following:

1. A robust way to detect non-linear effects and estimate which combinations of chemicals will interact in a non-linear way.
2. Identify sources of uncertainty in results to inform what data needs to be collected.
3. Reliably predict from a mix of discrete/continuous/interval data.

Over to Tsoogii