

# Using R For Flexible Modelling Of Pre-Clinical Combination Studies

Chris Harbron

Discovery Statistics

AstraZeneca





# Modelling Drug Combinations

- Why?
- The theory
- An example
- The practicalities in R

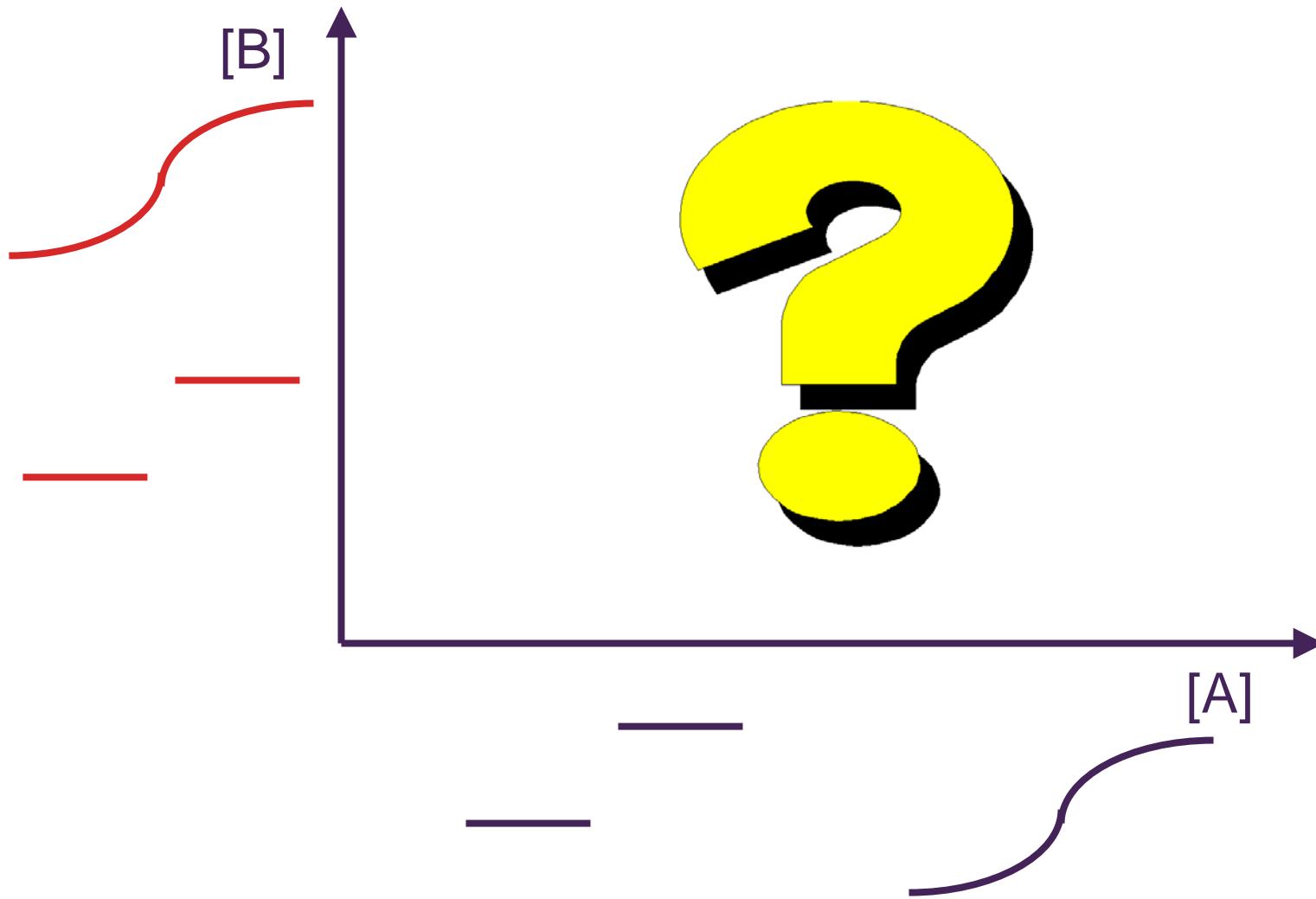


# Why Drug Combinations?

- Making better use of our assets
- Some marketed compounds are combinations e.g. Symbicort
- In some disease areas, e.g oncology, HIV, polypharmacy is the norm
- Compounds licensed only for use in combination with a specific other agent
  - Lapatinib (GSK – Breast cancer) is approved for use in combination with capecitabine
- Increased molecular & pathway level understanding
  - Hypothesise and understanding synergistic actions
  - Link with systems biology

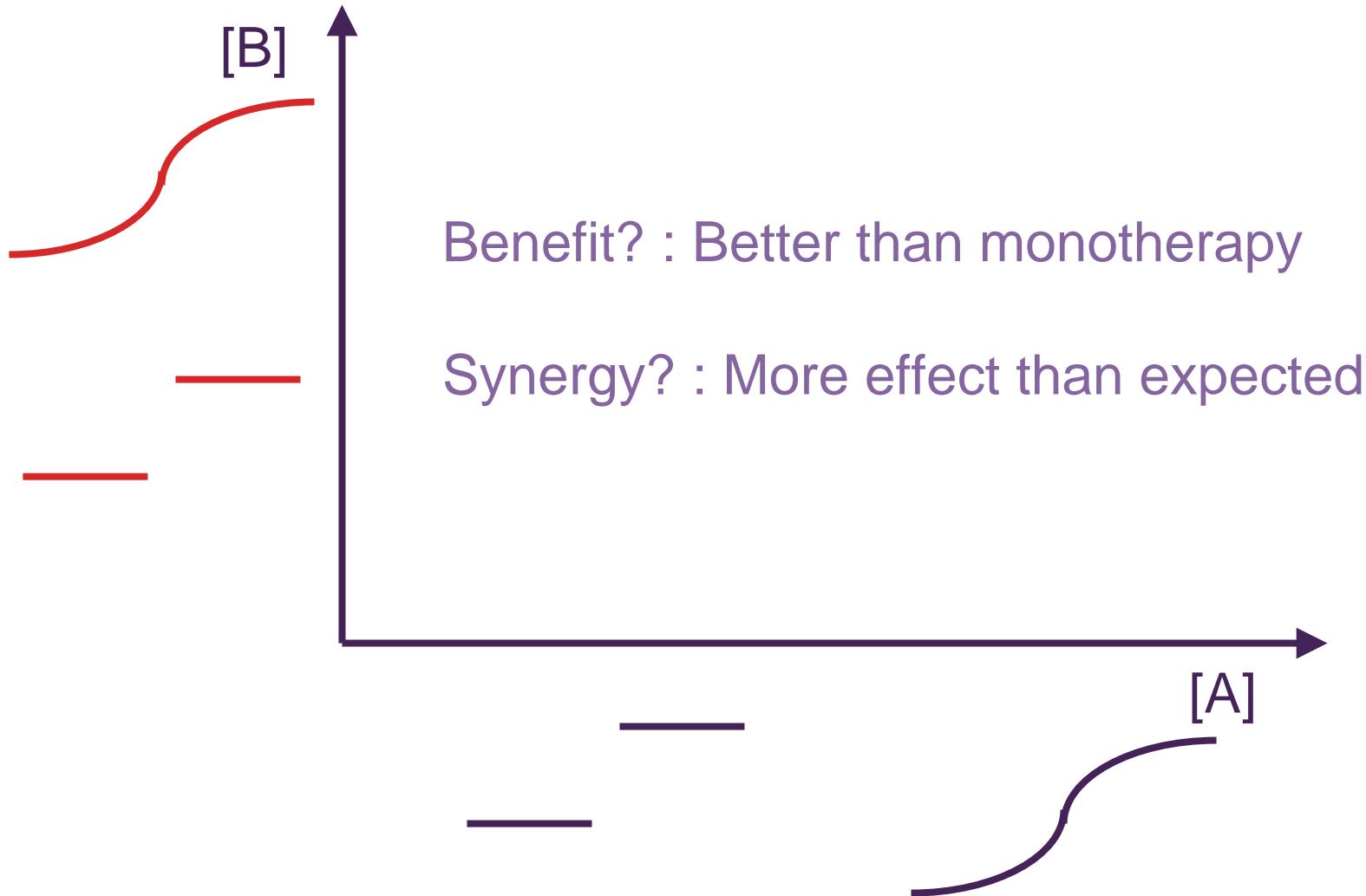


# Combination Studies



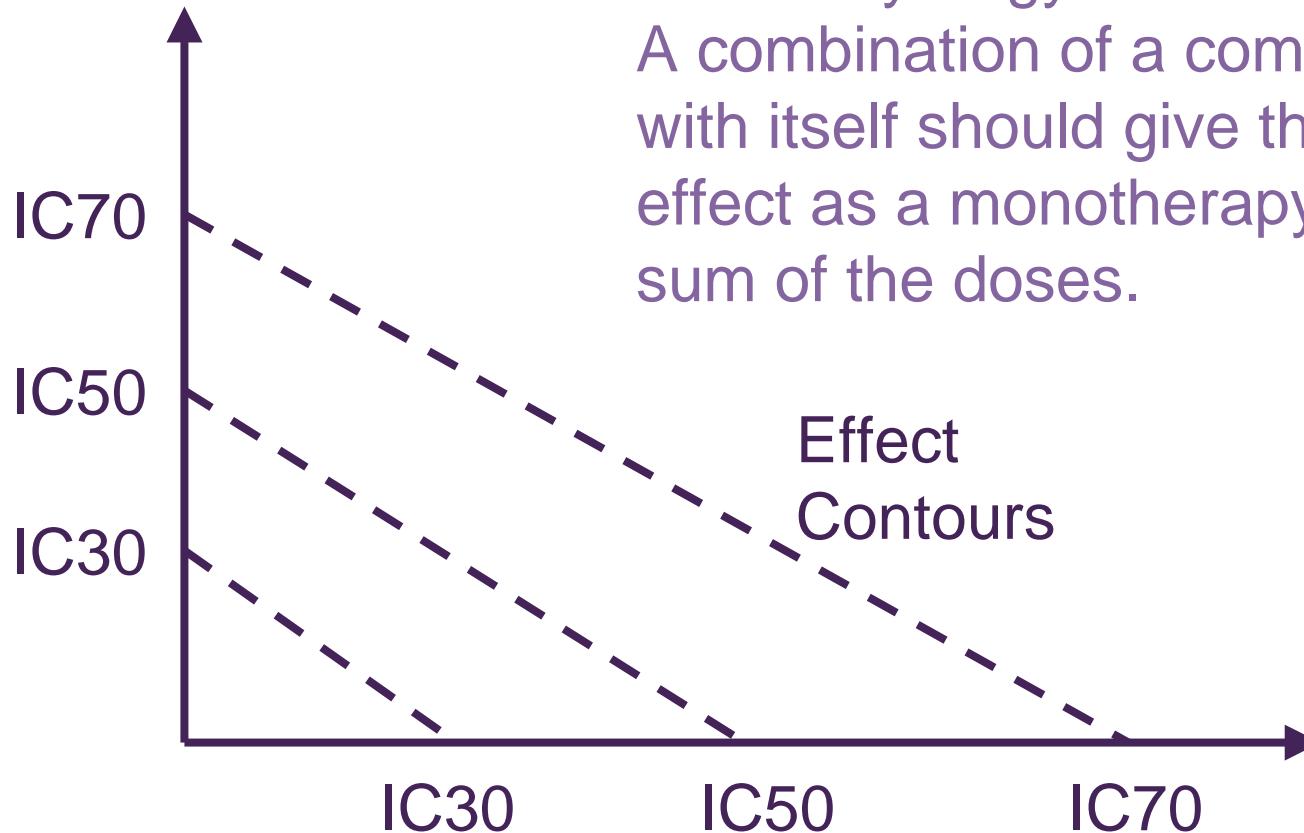


# Combination Studies





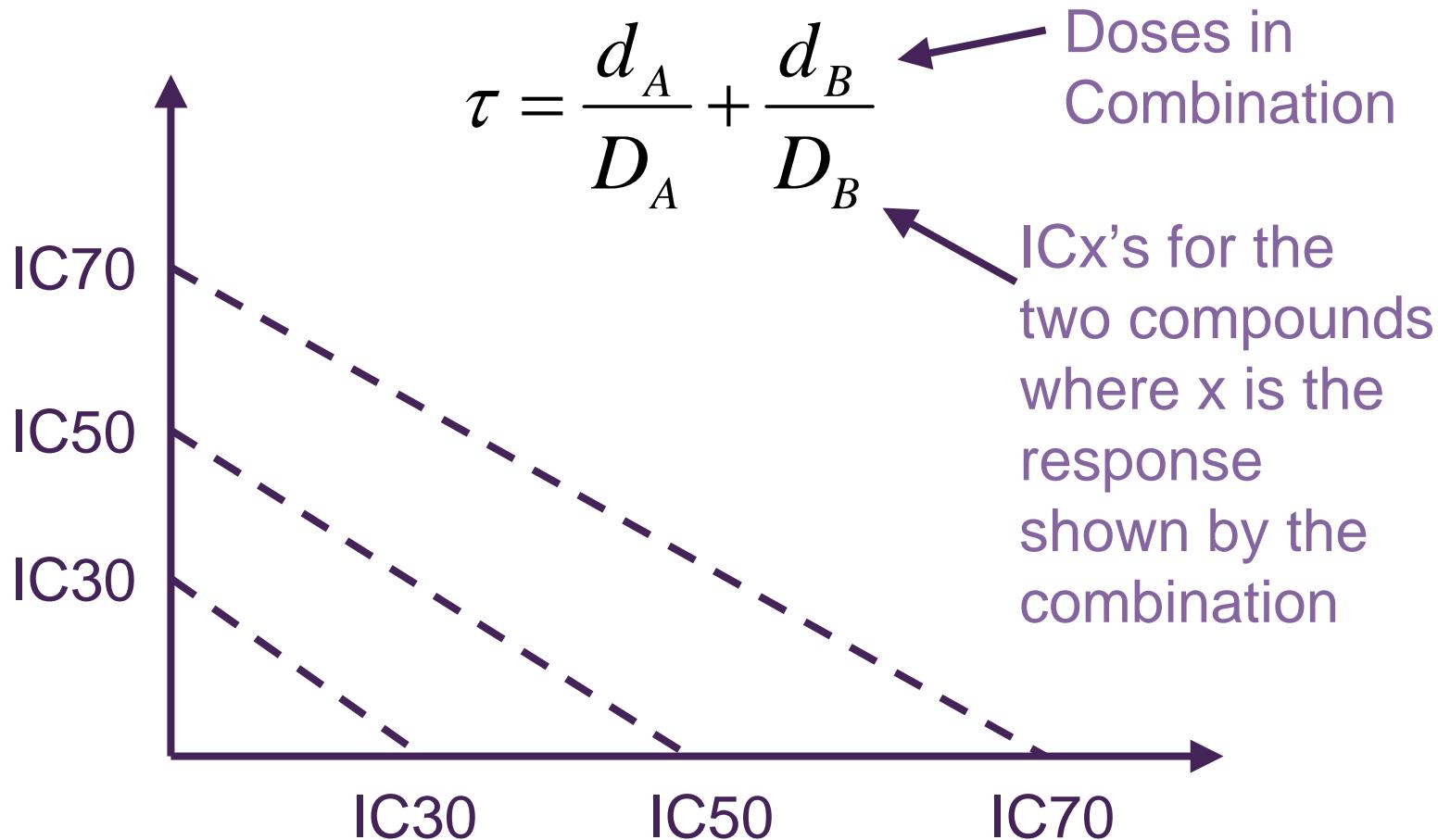
# Assessing Synergy Loewe Additivity



Based around “sham synergy”  
or “self synergy”  
A combination of a compound  
with itself should give the same  
effect as a monotherapy at the  
sum of the doses.

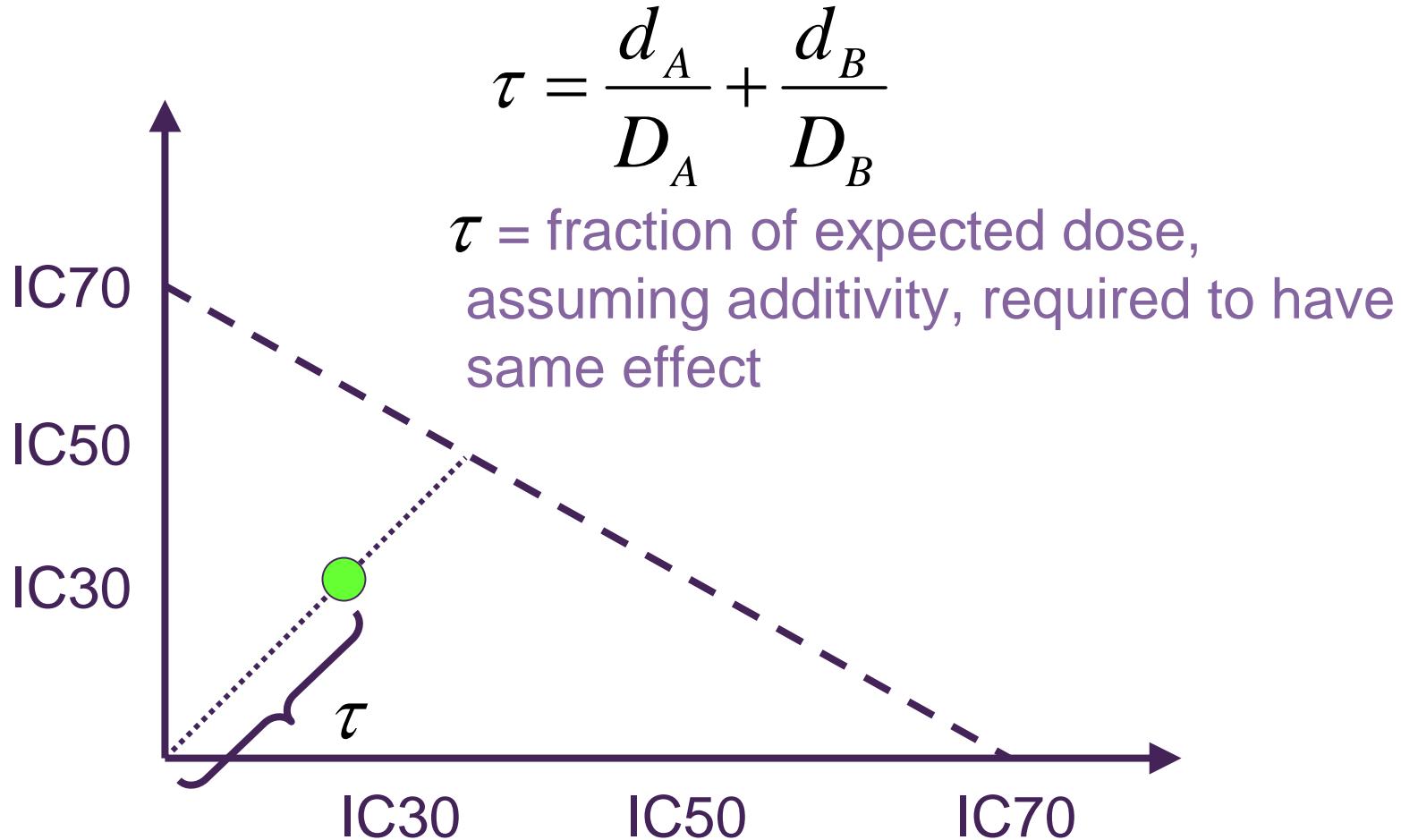


# Interaction Index – Berenbaum Combination Index – Chou & Talalay



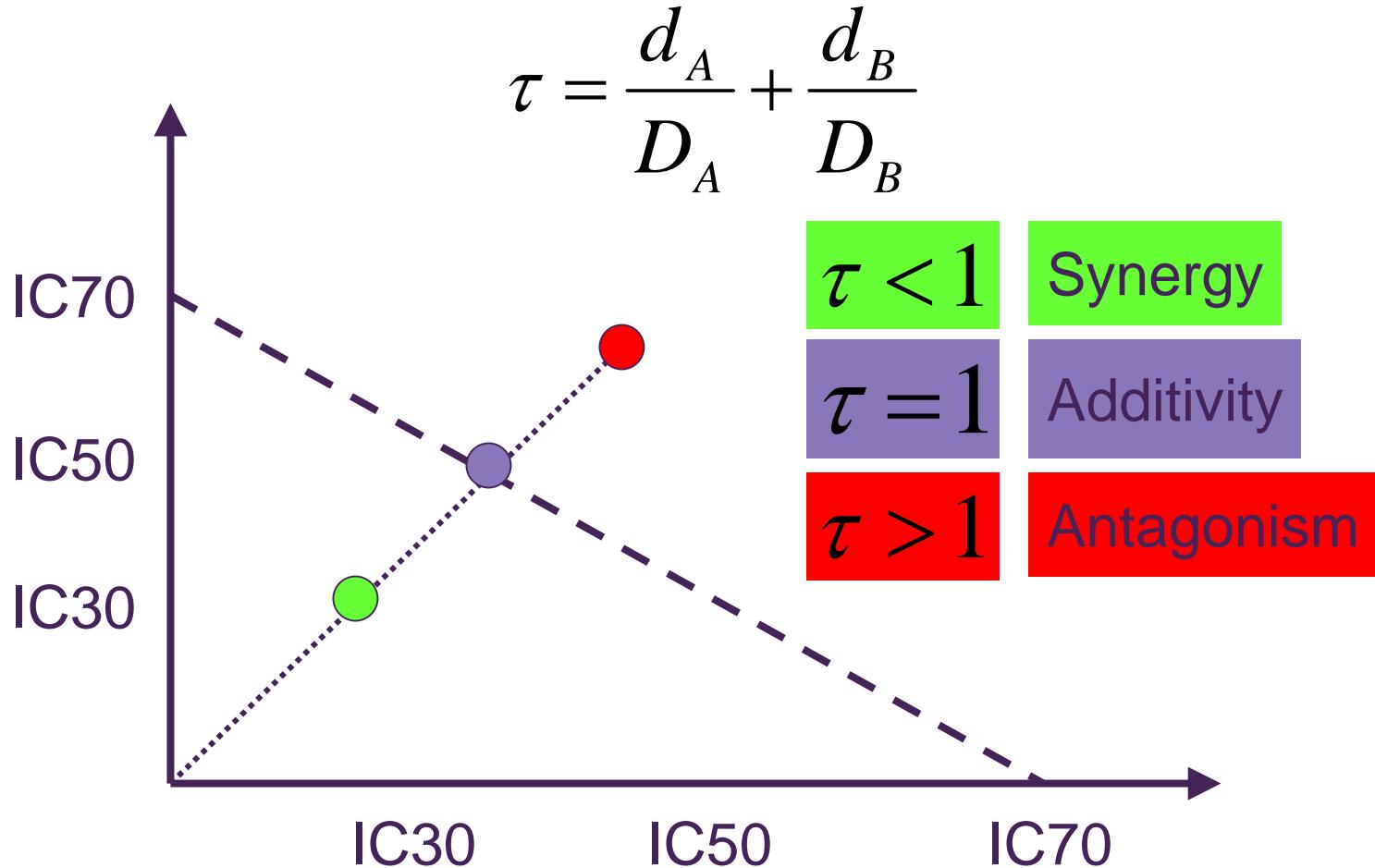


# Interaction Index – Berenbaum Combination Index – Chou & Talalay





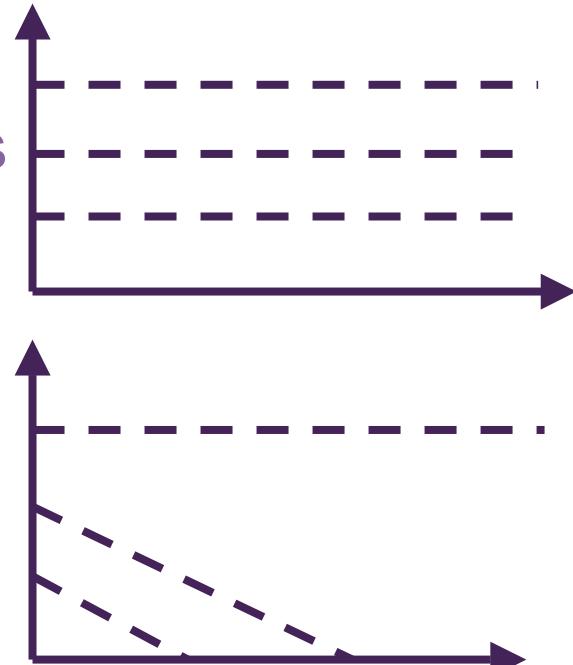
# Interaction Index – Berenbaum Combination Index – Chou & Talalay





# Interaction Indices

- Wish to calculate these:
  - With standard errors / confidence intervals
    - Statements of confidence – significance tests
- Use more flexibly and powerfully
  - Combining combination doses together
  - Overall assessments of synergy
- Covering a wide variety of situations
  - Inactive agent
  - Partial Response Agent
  - Multiple Plates / Experiments





# Unified Tau

$$1 = \begin{cases} \frac{d_A}{D_A} + \frac{d_B}{D_B} & d_A \text{ or } d_B = 0 \quad \text{Monotherapies} \\ \frac{d_A}{D_A} / \tau_{(i)} + \frac{d_B}{D_B} / \tau_{(i)} & d_A \text{ and } d_B > 0 \quad \text{Combinations} \end{cases}$$

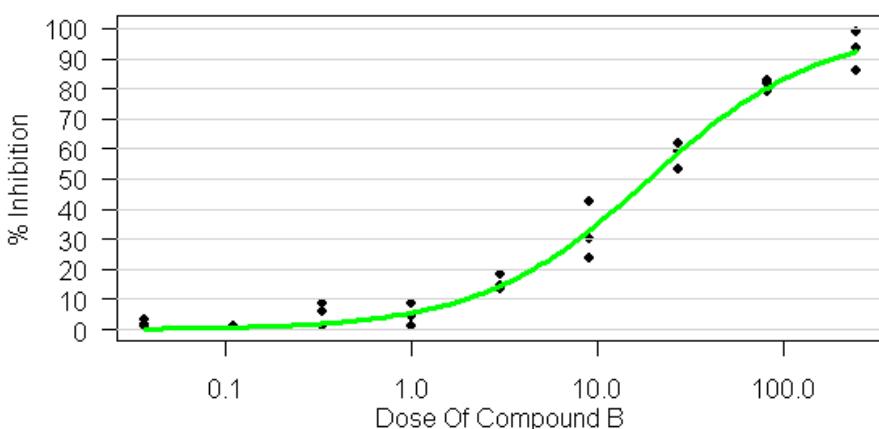
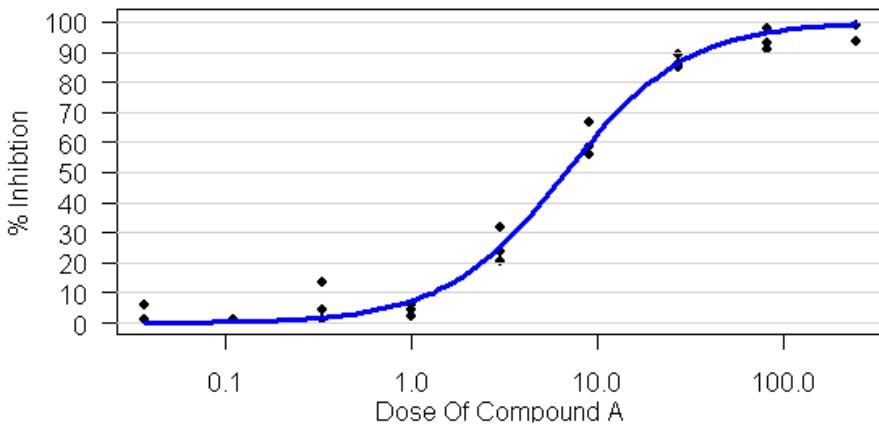
- Where  $\tau_{(i)}$  is either:

- a constant – response surface
  - (with discontinuities at the axes)
- a separate value for each point
  - Berenbaum's interaction index
- a separate value for each ray (segment)
- a separate value for each dose level of a compound
- could fit tau as a continuous function of dose or ray

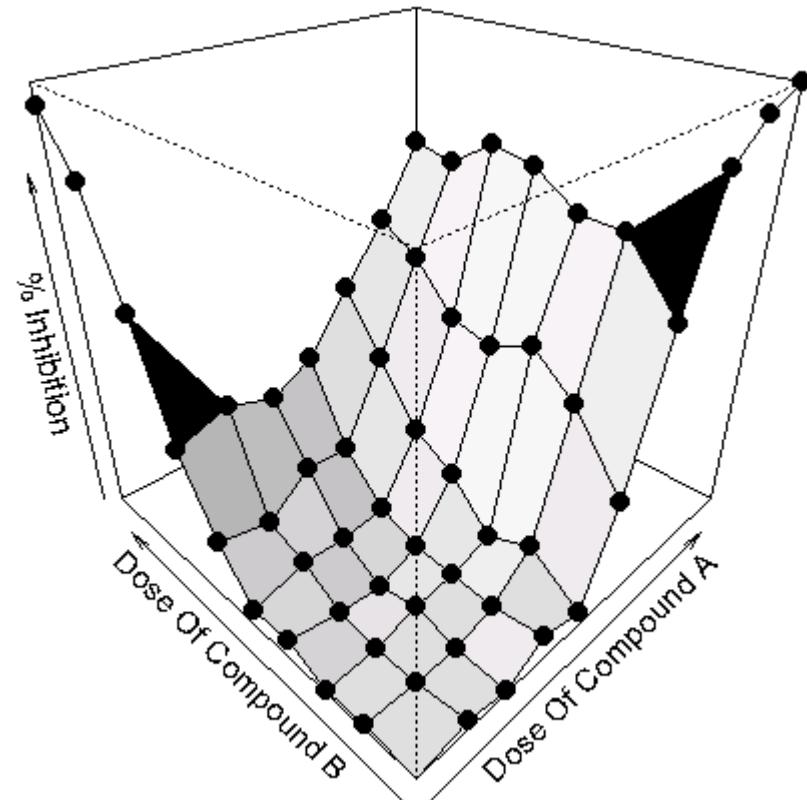


# An Example

## Monotherapies



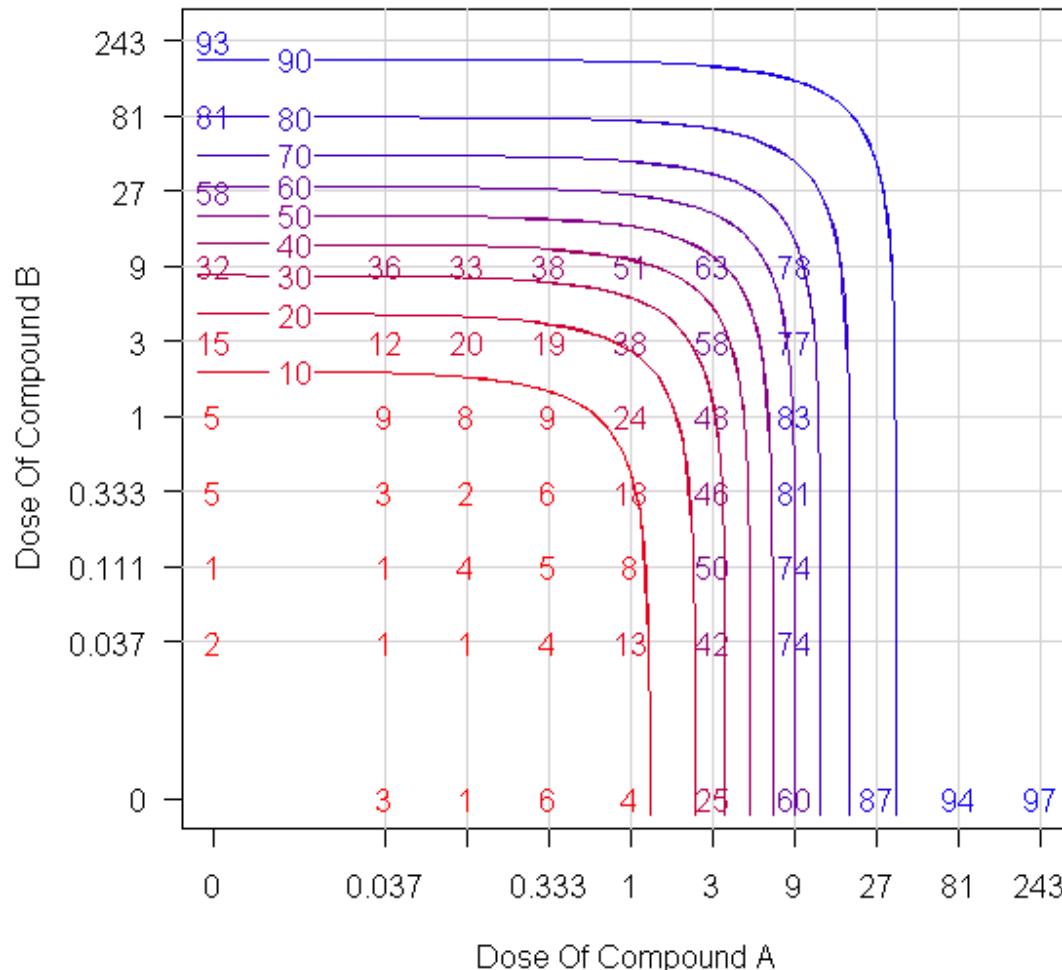
## Combinations





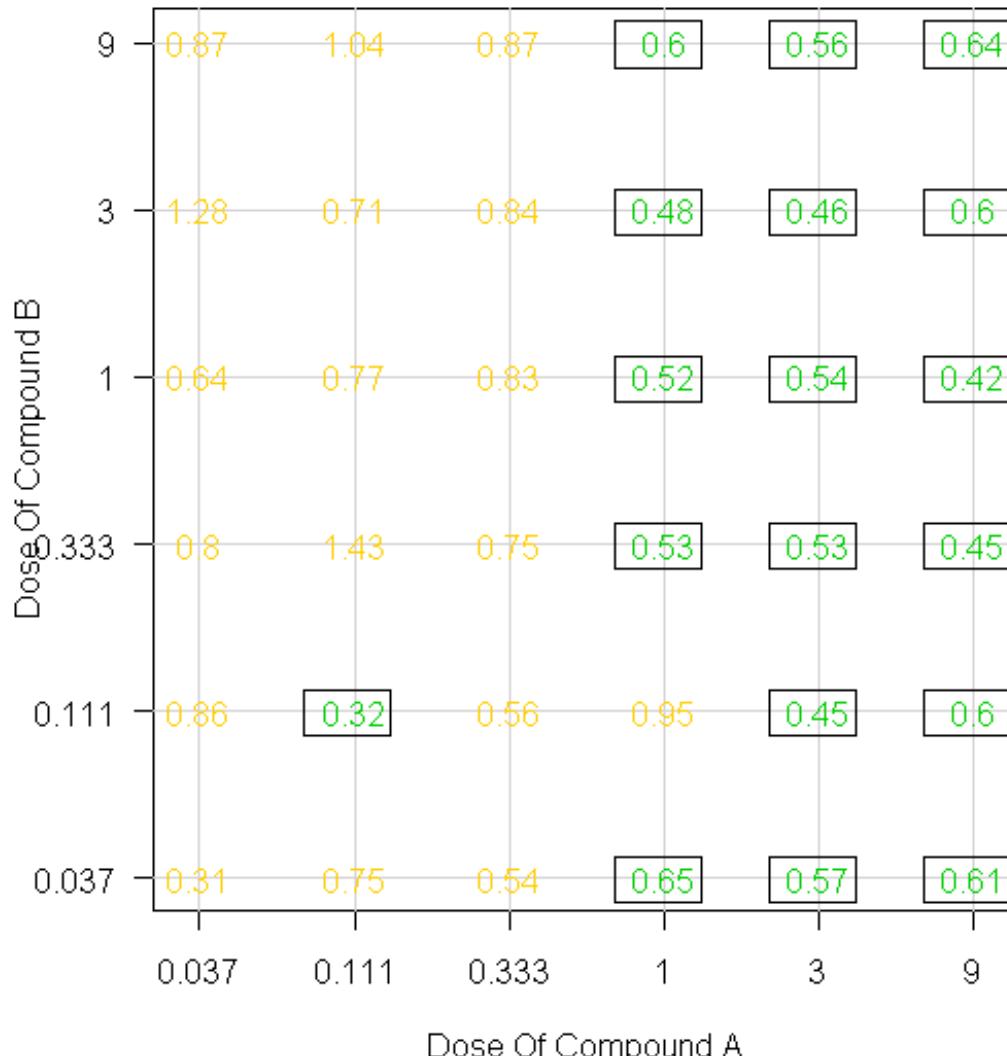
# EDA Suggests Synergy At Higher Doses Of Drug A

Data & Isobologram Assuming Additivity



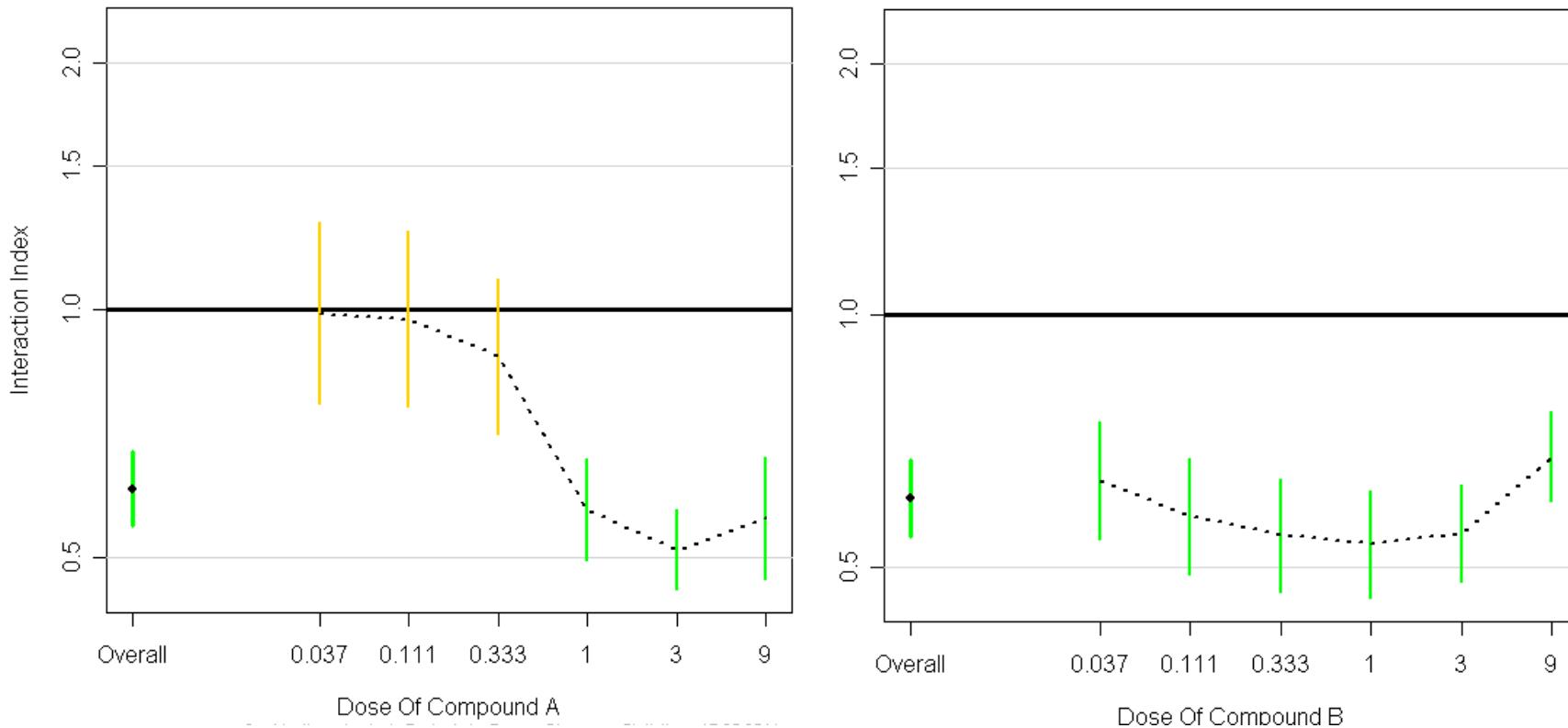


# Identify Individual Combinations Significantly Demonstrating Synergy





# Estimates Of Synergy With 95% CIs Overall & For Different Dose Levels





# Fitting in R

```
fit <- mynls(formula , start=inits)
```

Robust  
version  
of nls()

Selection of  
starting  
parameters

```
response ~ tau.model(.....)
```

Flexibly  
building  
formula

Formula expressed as  
 $1 \sim f(Y, \text{parameters})$   
Not  
 $Y \sim f(\text{parameters})$

```
as.formula(paste(...))
```

Iterative fitting



# Flexibly Building Formula

Varying number of combination parameters to be fit:

```
as.formula(paste("resp ~ tau.model(parameters,  
paste("logtau" , 1:ntaus , sep="" , collapse=","),  
"gp= c(",paste(groupindex,collapse=",") ,  
""))" ))
```

- Build as a text string, then convert to a formula
- Varying numbers of tau parameters
- Convert group index vector into a text string in the right format



# Iterative Fitting of Formula

Iterative Non-linear curve-fitting performed by nls() :  
monotherapy and tau parameters

tau.model(d<sub>1</sub>,d<sub>2</sub>,m<sub>1</sub>,m<sub>2</sub>,lower<sub>1</sub>,lower<sub>2</sub>,ldm<sub>1</sub>,ldm<sub>2</sub>,taus)

For each observation :

Make initial estimate of Y

Calculate D<sub>1</sub> & D<sub>2</sub> –

monotherapies required to achieve Y using Hill equation

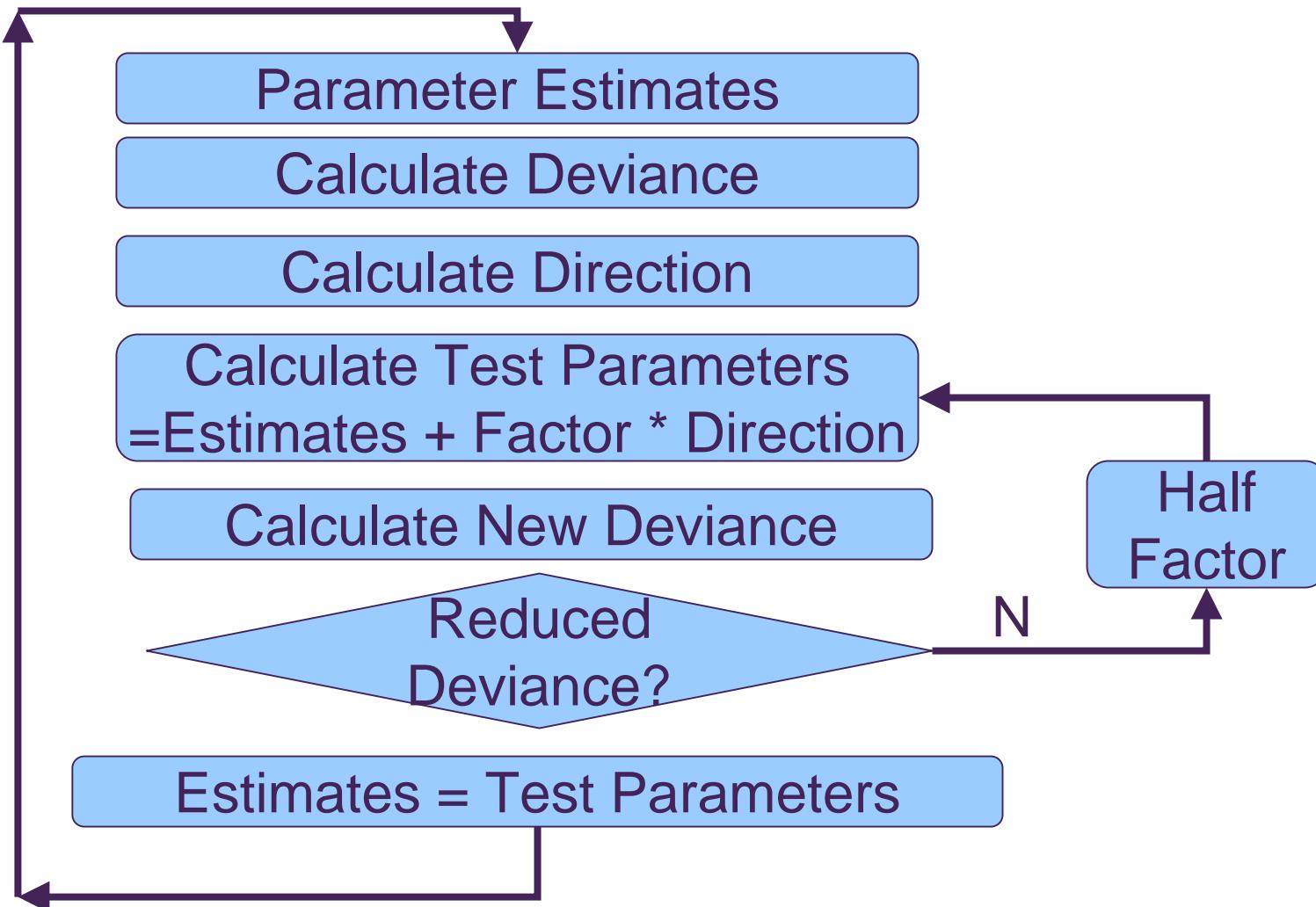
Adjust Y up or down depending on whether

$$\frac{d_1/\tau_{(i)}}{D_1} + \frac{d_2/\tau_{(i)}}{D_2} \quad \text{is } >1 \text{ or } < 1$$

Iterate until Y is accurately estimated

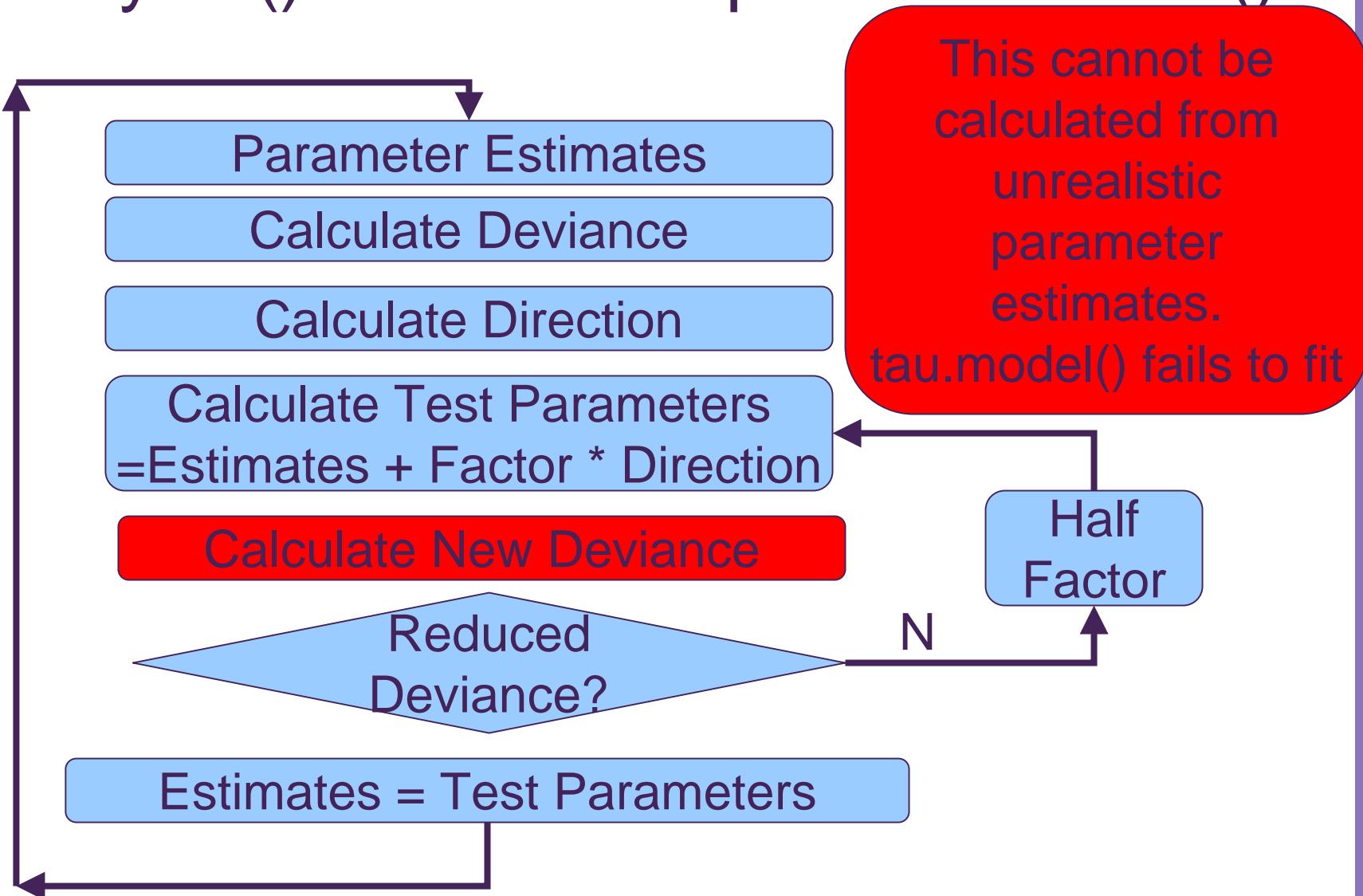


# mynls() : A less temperamental nls()



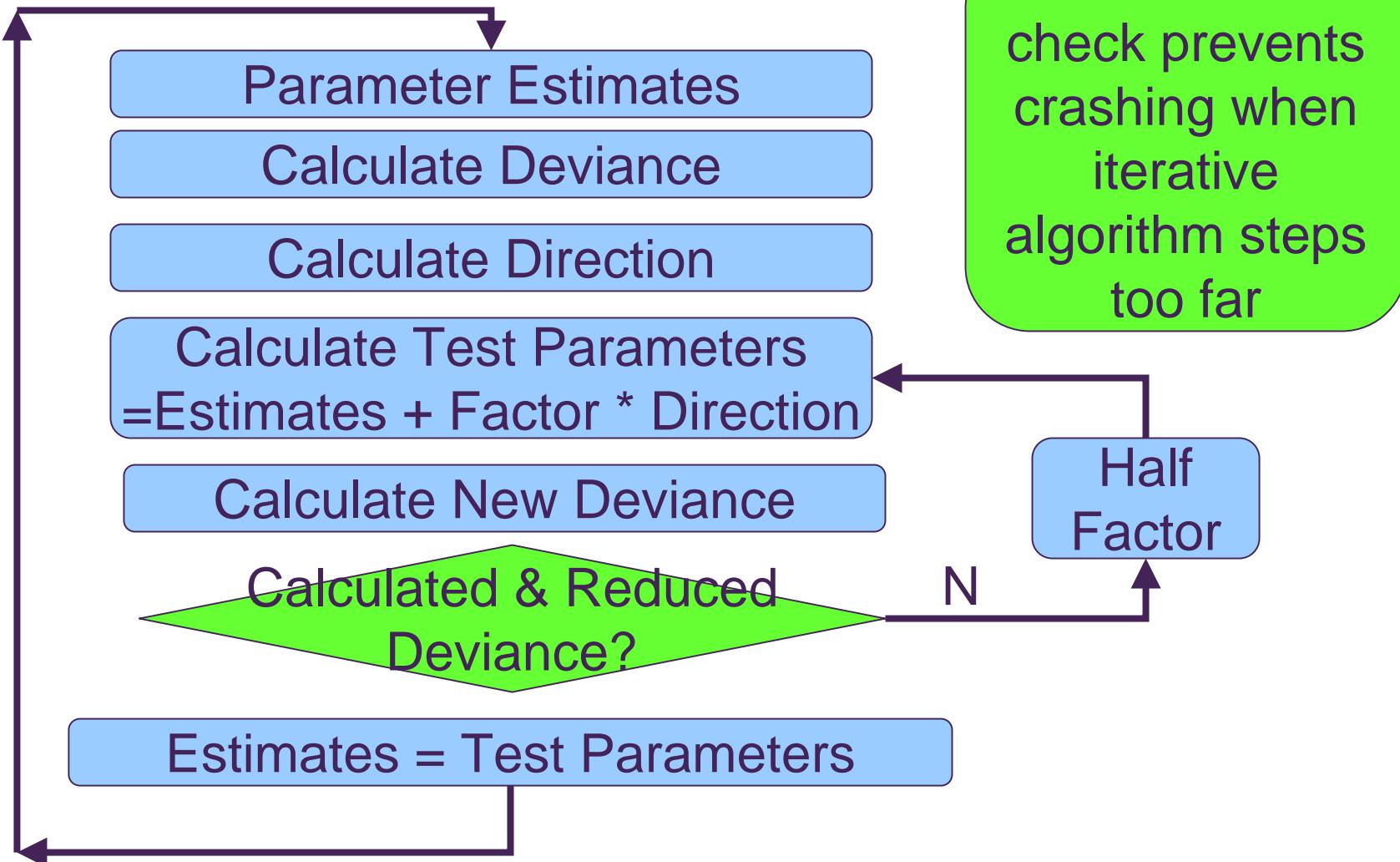


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# Starting Parameters

- Good starting parameters from fitting marginal distributions (e.g. monotherapies) and direct calculations
  - In some situations, this can be done exactly, so `nls()` converges immediately to the starting parameters, but with standard errors added
- Starting from multiple starting points decrease risks of local minima
- Identify and fix parameters likely to shoot off to infinity beforehand



# Summary

- Early identification of synergistic drug combinations of strategic importance within the pharmaceutical industry
- Powerful and flexible methodology for identifying and characterising synergy
- R provides a powerful environment for fitting and visualising these models
- Careful programming increases the robustness and success rate of R in fitting these models