

# A Unified Approach to Flexible and Powerful Modeling Of Pre-Clinical Combination Studies

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Discovery Statistics

AstraZeneca

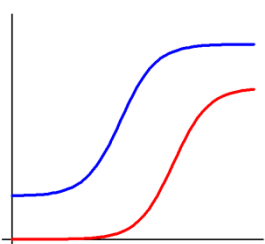
# Why Drug Combinations?

- Making better use of our assets
- In many disease areas, e.g oncology, cardiovascular, HIV, polypharmacy is the norm
- Numerous examples of approvals for drug combinations
- Increasingly focussed and selective compounds
- Increased molecular & pathway level understanding
  - Hypothesise and understanding synergistic actions
  - Link with systems biology
- Recent FDA call for comments on combination drug treatments



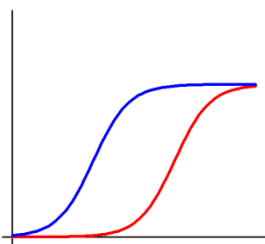
# Classification Of Combinations

## • Efficacy Enhancing



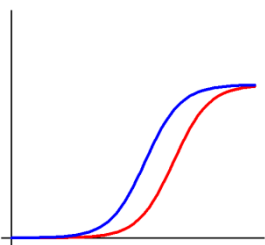
- Increased efficacy in combination beyond what can be achieved by single agent
- Synergy : greater response than expected under additivity
- Indicative of a positive mechanistic interaction
- Potential for Patentable Intellectual Property

## • Dose Sparing



- Efficacy in combination achieved at lower doses than by a single agent
- Synergy : greater response than expected under additivity
- Indicative of a positive mechanistic interaction
- Potential for Patentable Intellectual Property

## • Beneficial

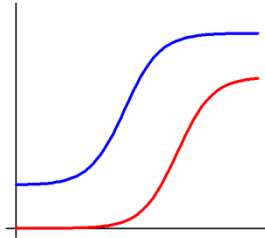


- Efficacy in combination achieved at lower doses than by a single agent
- FDA guidelines on combinations refer to as contributing
- May be compatible with compounds sharing pathways and mechanism of action

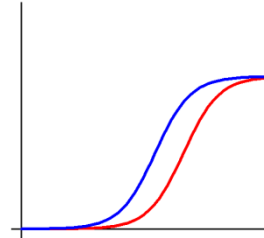


# Analysis Of Combinations

Efficacy Enhancing



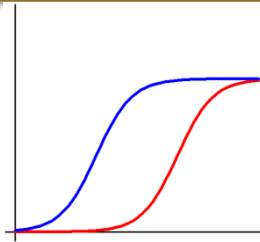
Beneficial



Relatively simple comparisons

ANOVA t-test or separate parameters in curve fit

Dose Sparing



Synergy : greater response than expected under additivity

Harder



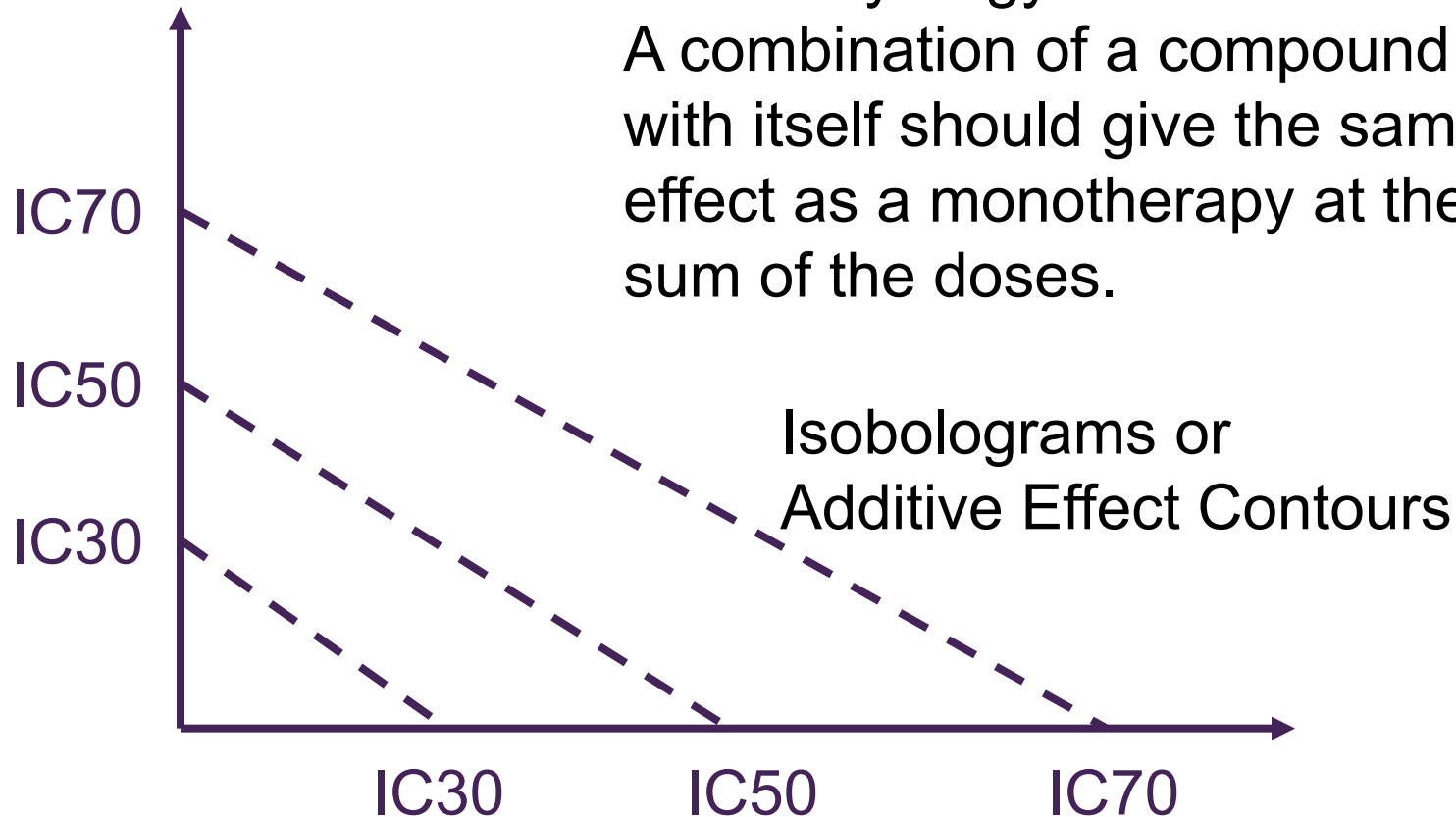


# Assessing Synergy

## Loewe Additivity

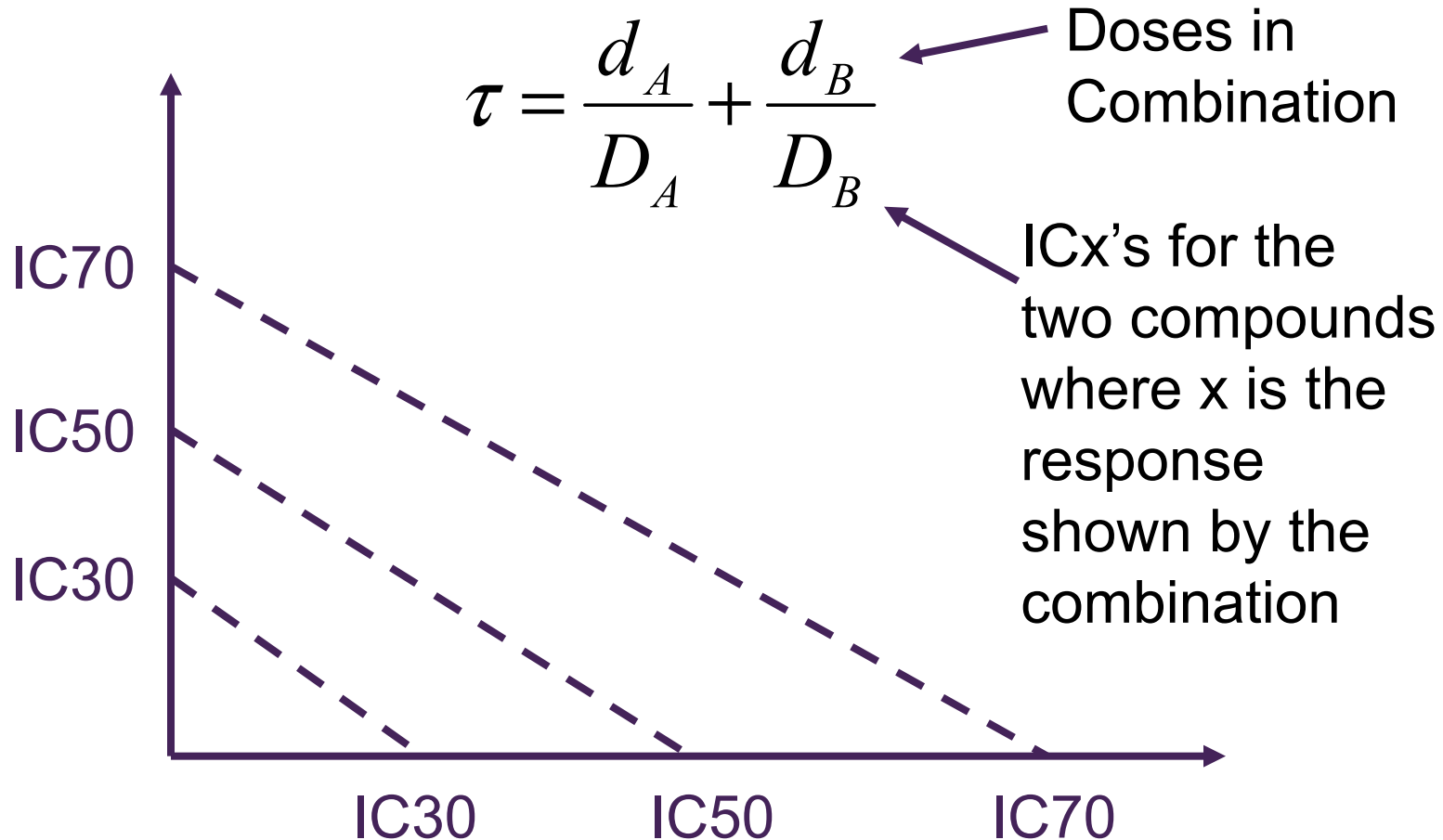
Based around “sham synergy” or “self synergy”

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

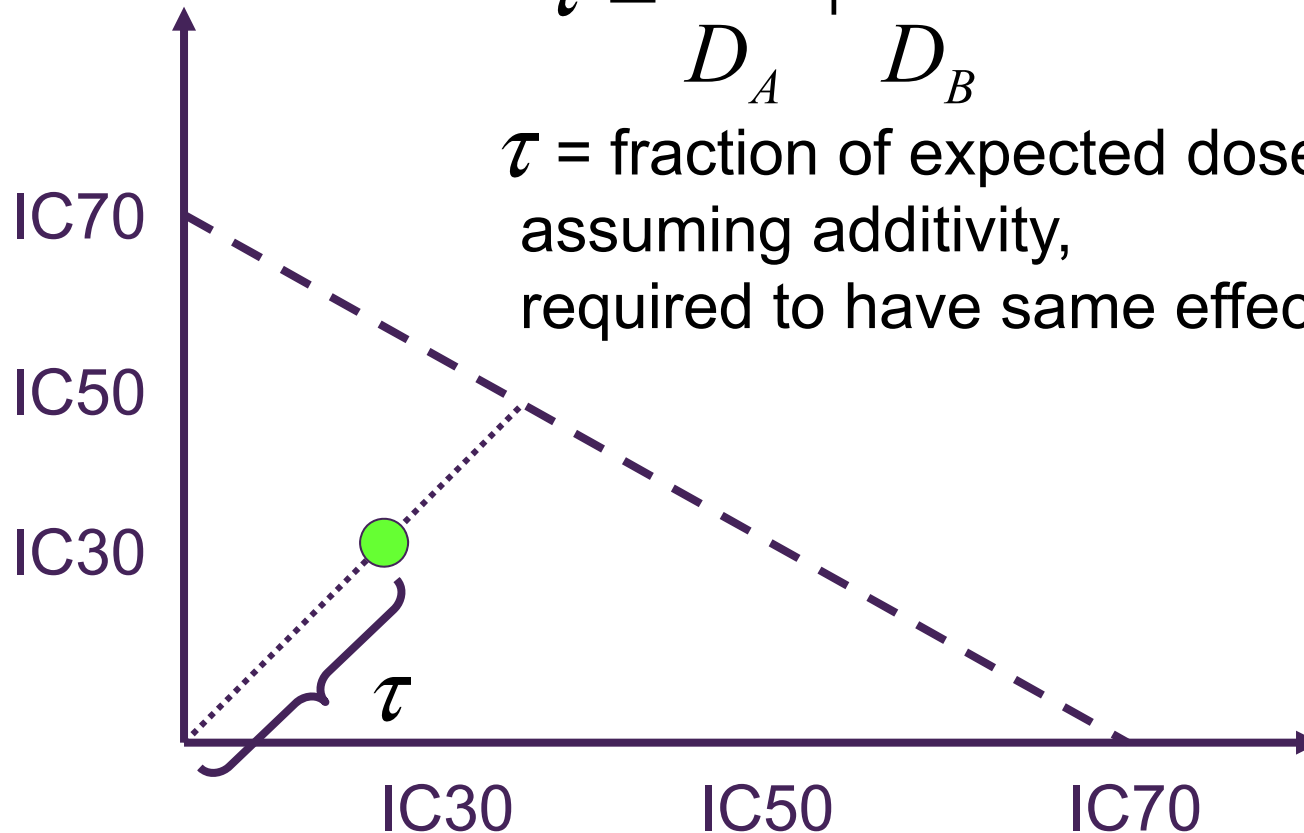




# Interaction Index – Berenbaum Combination Index – Chou & Talalay

$$\tau = \frac{d_A}{D_A} + \frac{d_B}{D_B}$$

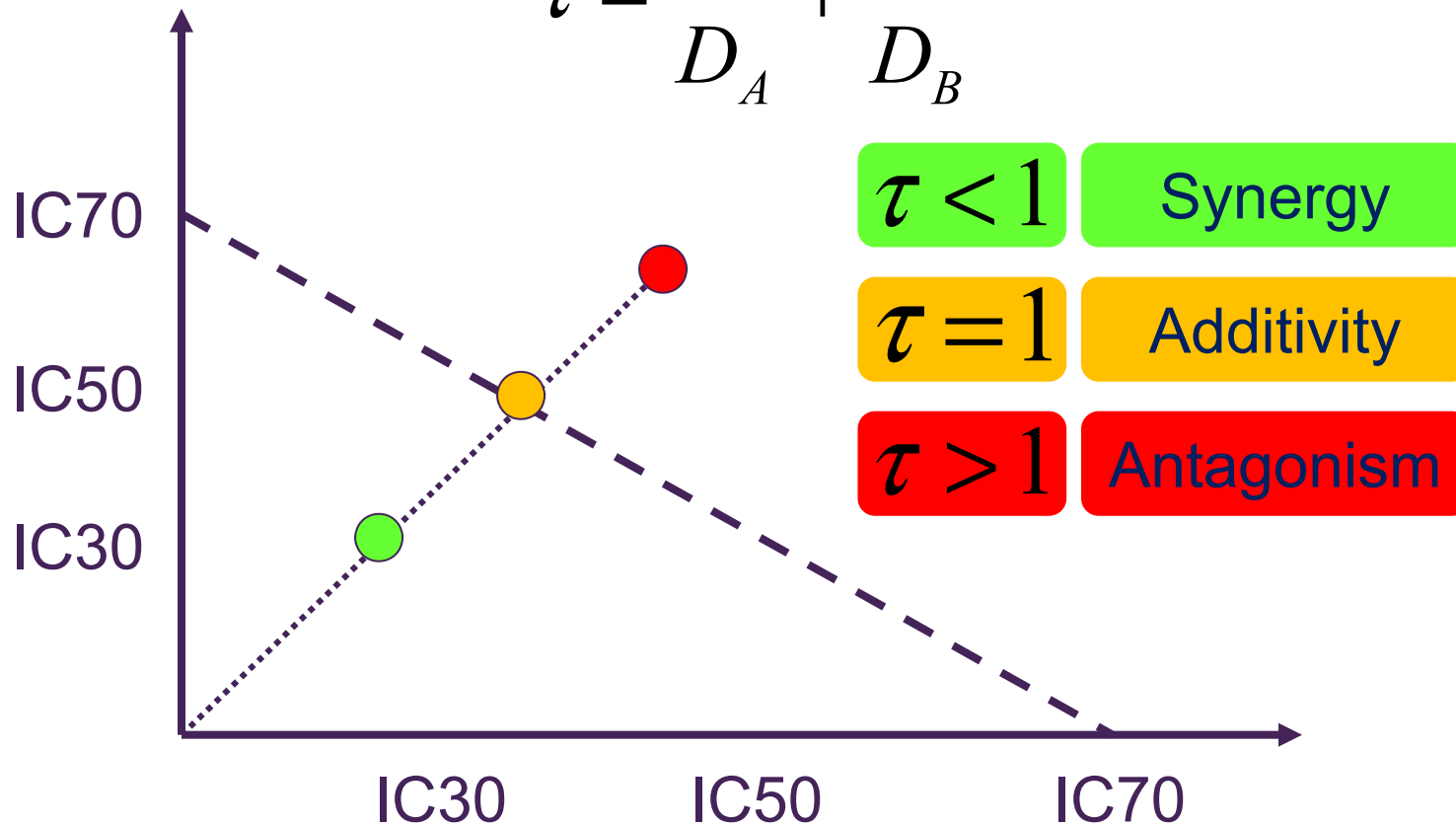
$\tau$  = fraction of expected dose,  
assuming additivity,  
required to have same effect





# Interaction Index – Berenbaum Combination Index – Chou & Talalay

$$\tau = \frac{d_A}{D_A} + \frac{d_B}{D_B}$$







# Interaction Indices

- Wish to calculate these:
  - With standard errors / confidence intervals
  - Statements of confidence and significance tests
- Flexibly and powerfully
  - Utilise all data in a wide variety of designs
  - Combining combination doses together
  - Overall assessments of synergy
- Covering a wide variety of situations
  - Partial responses
  - Inactive compounds as monotherapies
  - Multiple subjects or plates



# Unified Tau

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

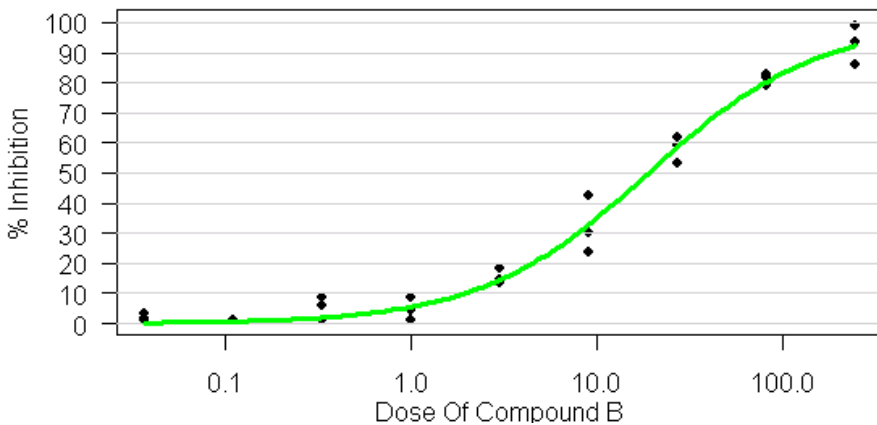
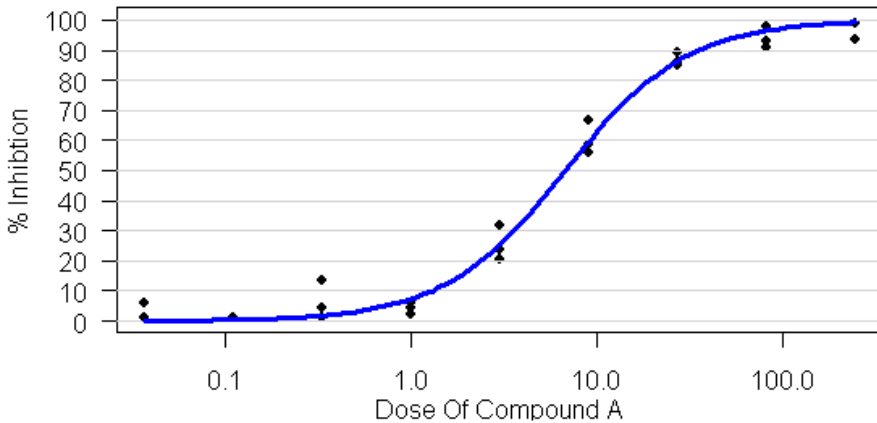
- Where  $\tau_{(i)}$  is either:

- 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
- a continuous function of dose or ray

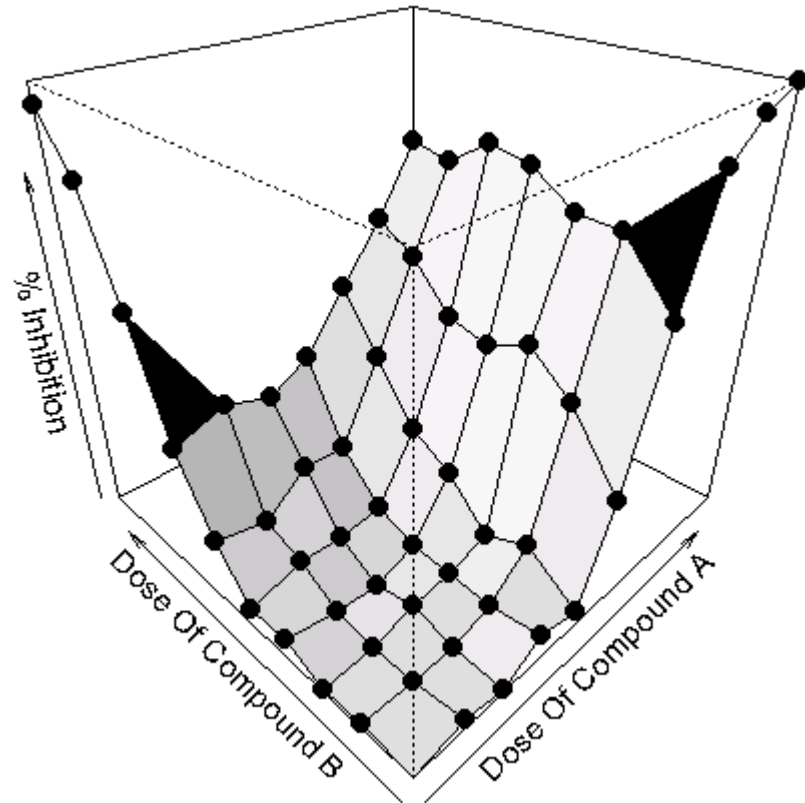
# An In-Vitro Example

## Inhibition of growth in cell lines

### Monotherapies



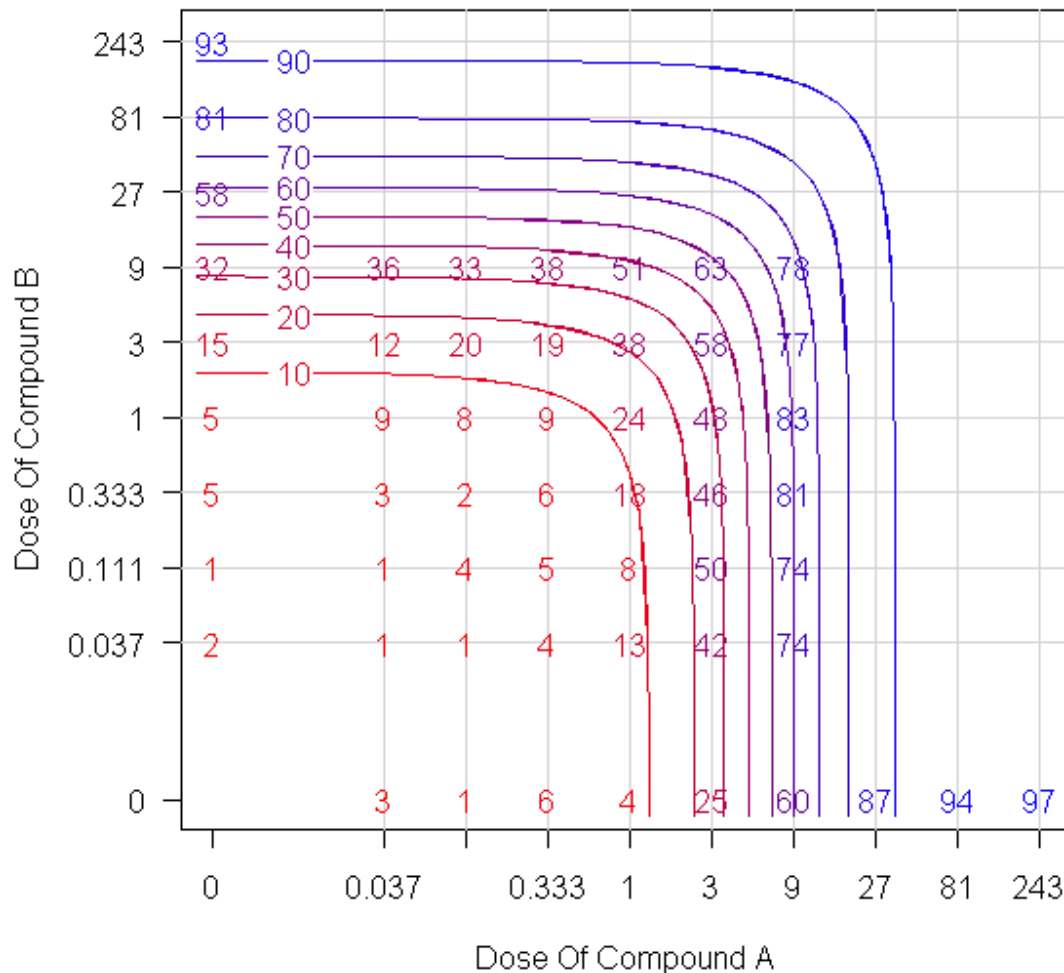
### Combinations





# EDA Suggests Synergy At Higher Doses Of Drug A

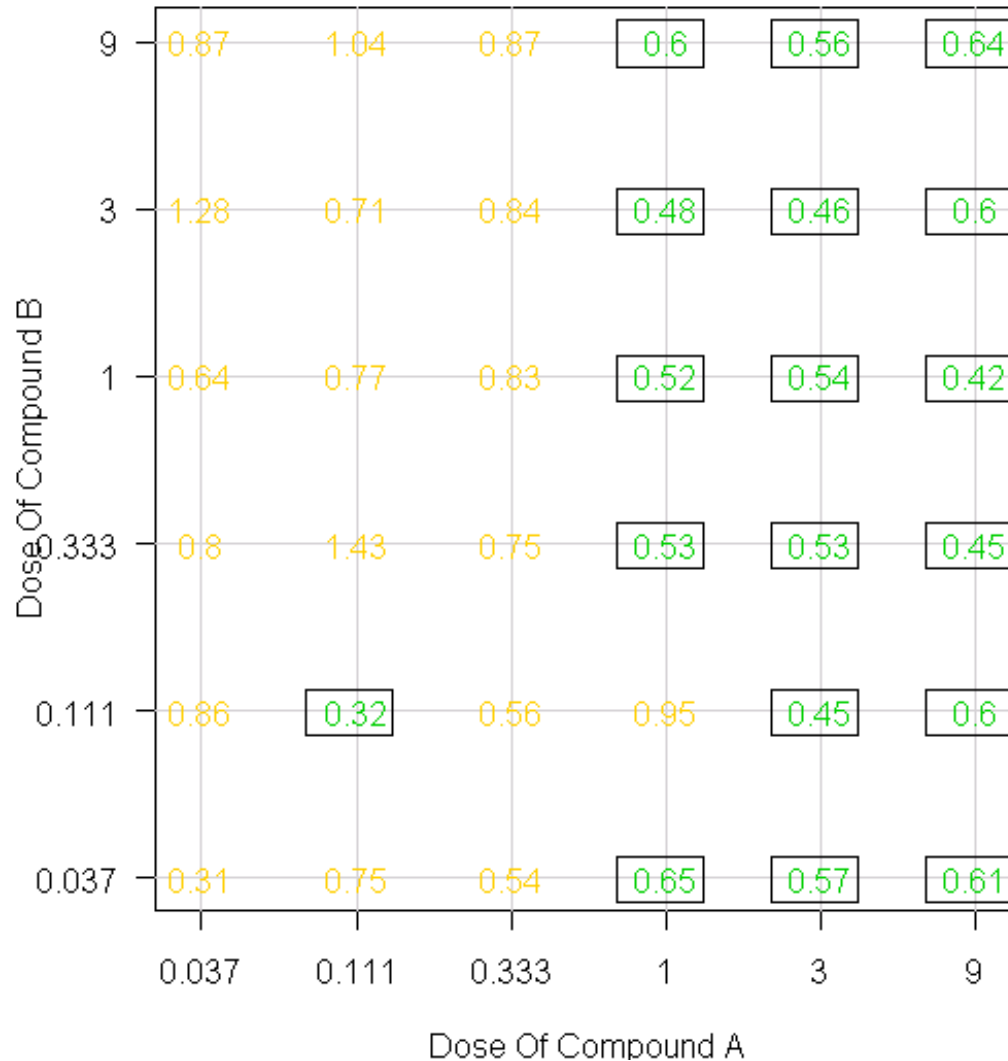
Data & Isobologram Assuming Additivity



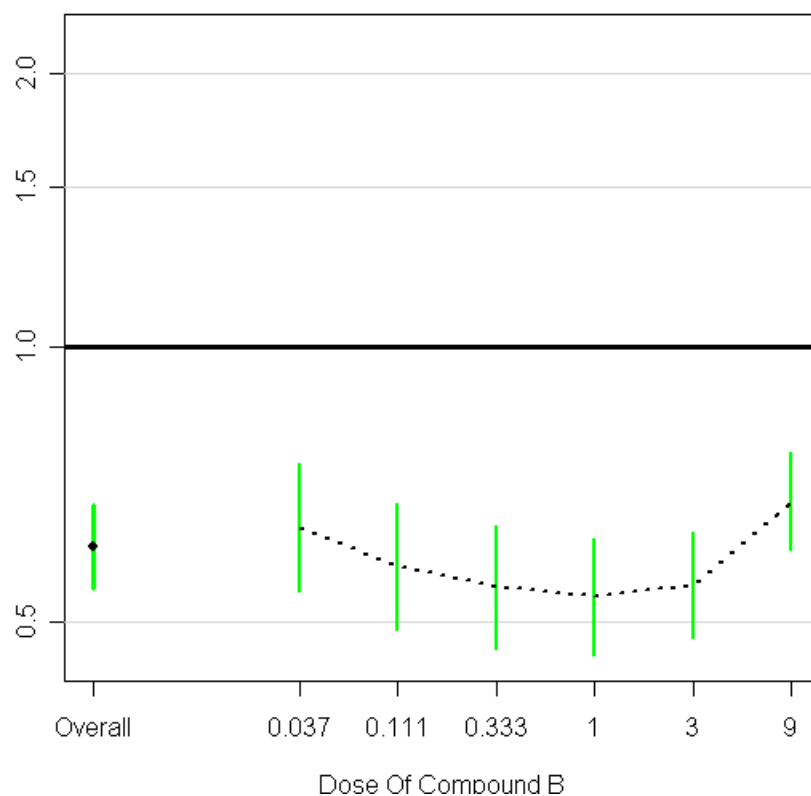
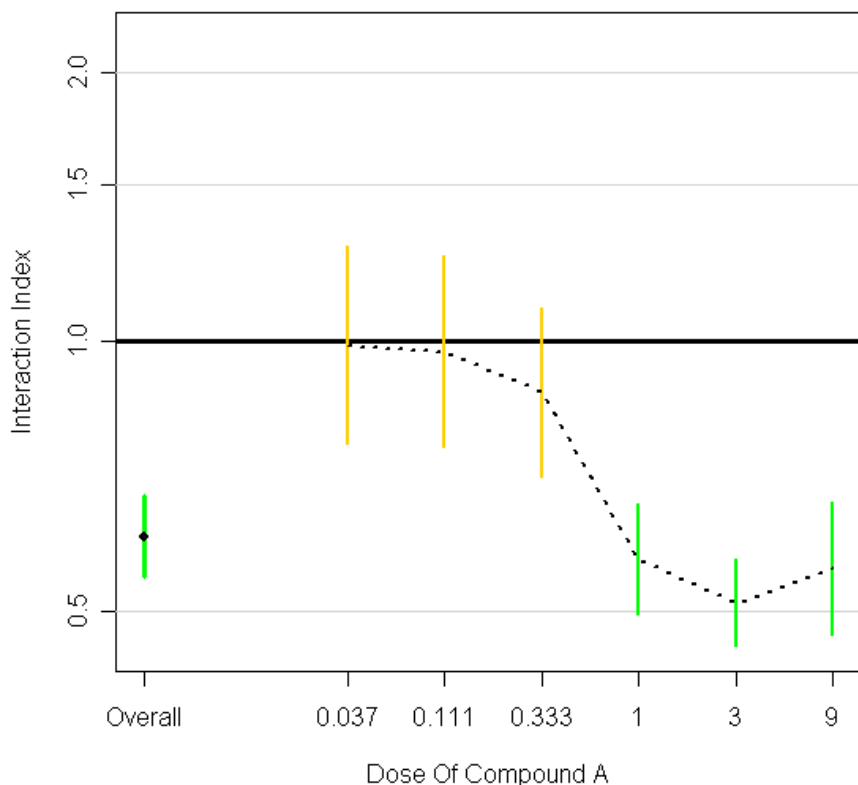
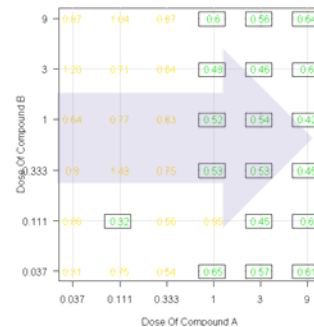
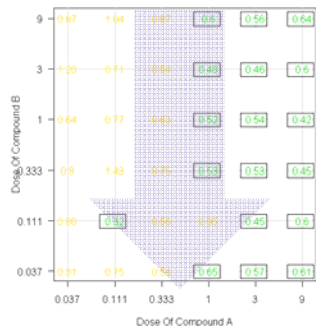
Wish to quantify  
this considering  
uncertainty in  
observations &  
monotherapy fits



# Identify Individual Combinations Significantly Demonstrating Synergy



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



# Testing Hierarchies of Models



Model	rdf	RSS	df	F	p-Value
Additive	158	5704			
Common $\tau$	157	3672	1	86.86	$2 \times 10^{-16}$

Linearly varying $\tau$ over doses of compound A	156	3130	1	27.00	$6 \times 10^{-7}$
Separately varying $\tau$ over doses of compound A	152	2834	4	3.96	0.004
Separate $\tau$ for each combination	122	2079	30	1.47	0.072

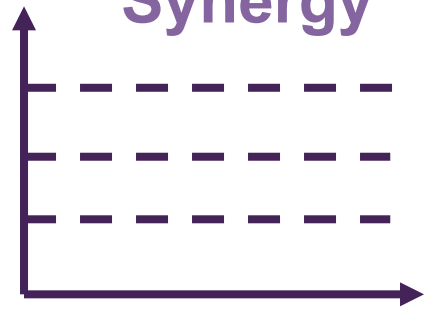
Linearly varying $\tau$ over doses of compound B	156	3672	1	0.00	0.984
Separately varying $\tau$ over doses of compound B	152	3334	4	3.85	0.005
Separate $\tau$ for each combination	122	2079	30	2.45	0.0003



# Naturally Extends To A Range of Situations

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

**Inactive Agent /  
Therapeutic  
Synergy**



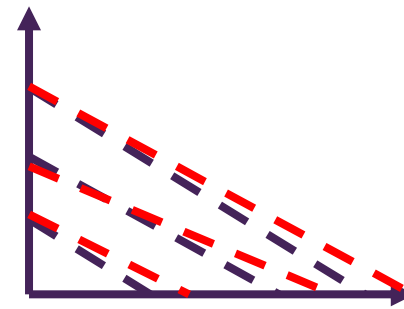
$$D_B = \infty$$

**Partial  
Response**



$$D_B = \infty : R > U_B$$

**Multiple  
Experiments**



Separate Monotherapy  
Curves  
Common  $\tau_{(i)}$  parameters

**Extendable to >2 agents**



# Summary

- Early identification of synergistic drug combinations of strategic importance within the pharmaceutical industry
- Powerful and flexible methodology for identifying, quantifying and characterising synergy
- Utilises all data to analyse any design with sufficient monotherapy data
- Consistent approach across different designs and response scenarios
- Found to be robust even to mis-specifications of response curves
- Implementation in R provides a powerful environment for fitting and visualising these models building upon standard functions e.g. `nls()` and `anova()`

A flexible unified approach to the analysis of pre-clinical combination studies, 2010,  
Statistics in Medicine, 29 (16) 1746-1756

