

# Dose Response Analysis for Combination Products

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# Scope

- Two drugs in a fixed dose combination
- At least one of the drugs is effective alone
- One (continuous) clinical endpoint
- Assume there is no PK interaction
- Literature review

# Example Combination Drug Product

## Vytorin

Ezetimibe – inhibits cholesterol absorption

Simvastatin – inhibits cholesterol production

Indication: Hypercholesterolemia

Doses: 10/10, 10/20, 10/40, 10/80

Vytorin (eze/simva) *	10/10	10/20	10/40	10/80
Lowering of LDL(%)	-45	-52	-55	-60
Ezetimibe *	10 mg			
Lowering of LDL(%)	-39			
Simvastatin *	10 mg	20 mg	40 mg	80 mg
Lowering of LDL(%)	-33	-34	-41	-49

[ \* Vytorin label (PDR) ]

# Outline

- Central Questions
- Multiple comparison procedures (MCP)
  - Minimum efficacious dose combination (MED)
  - Therapeutic synergy and contribution
- Modeling methods
  - Polynomial response surface
  - Mechanistic modeling
  - Empirical synergy
- Conclusions

# What do Regulators Want?

- 21 CFR 300.50 and ICH E4 (1994) require that 'each component make a **contribution** to the claimed effects'
- at the fixed doses, the combination must be superior to each of its two components alone at the same dose.
- $\text{Eff}(d_1, d_2) > \text{Eff}(d_1, 0)$  and  $\text{Eff}(d_1, d_2) > \text{Eff}(0, d_2)$

# Central Questions (part 1)

## Multiple Comparisons

1. Is there evidence that each drug contributes at some dose combination? (comparators are each drug alone)
2. What dose combinations give better results than each drug alone (at the same dose)?
3. Is there a dose combination that is *therapeutically synergistic*?

Therapeutic synergy: effect of combo is greater than max effect achievable by either component  
[Laska et al 1997 Stat in Med 16:2211-2228]

# Central Questions (part 2)

## Modeling

4. What response models give good predictions?
5. What is the optimal dose combination?
6. What do the results contribute to the understanding of the mechanism of action of the drugs alone or in combination?

[Adapted from Ruhberg 1995 & Pinheiro et al 2005]

# MCP methods

- Provide assessment of contribution and therapeutic synergy
- Treat dose as categorical
- Less sensitive to assumptions
- Requires less prior knowledge
- Allow for strong control of FWER



# Minimum Efficacious Dose (MED) combo

MED = “lowest” dose combination where both drugs contribute

K x N combo drug trial with  
 Drug A at doses  $i = 0, \dots, K$   
 Drug B at doses  $j = 0, \dots, N$   
 parallel groups of  $n$  subjects

mean response  $\mu_{ij}$  (monotone)

	0	1	...	N
0		$\mu_{01}$	$\mu_{0s}$	
1	$\mu_{10}$	$\mu_{11}$		
	$\mu_{r0}$		$\mu_{rs}$	
K				$\mu_{KN}$

expected gain  $\theta_{ij} = \min(\mu_{ij} - \mu_{i0}, \mu_{ij} - \mu_{0j})$

dose  $(r,s) = \text{MED}$  if  $\theta_{rs} > 0$  and

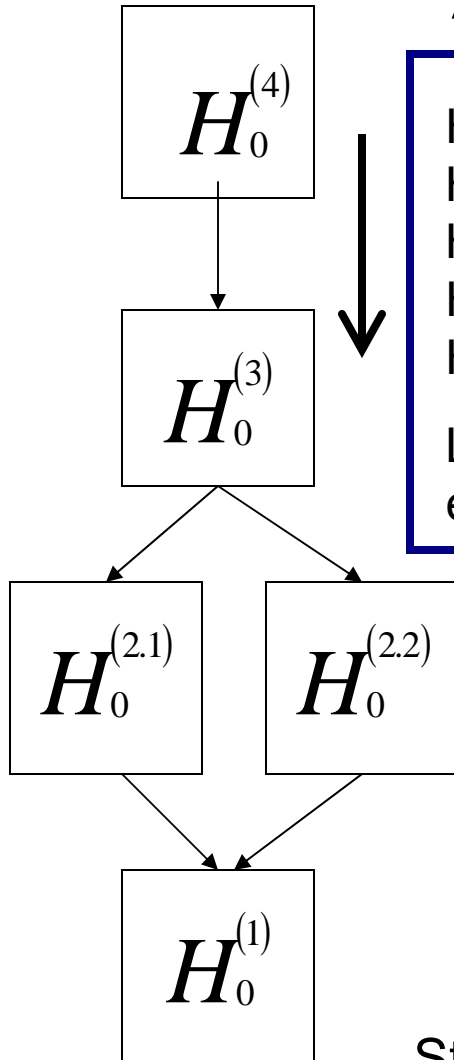
$\theta_{ij} = \theta_{rj} = \theta_{is} = 0$  for all  $i < r, j < s$

Then these are the lowest doses where both drugs contribute

$$\mu_{rs} > \mu_{r0}, \quad \mu_{rs} > \mu_{0s}$$

	1	...	N
1	$\theta_{11}$		$\theta_{1N}$
K	$\theta_{K1}$		$\theta_{KN}$

# AVE Test for MED Dose Combos



- $H_0^{(4)} : \theta_{11} = \theta_{12} = \theta_{21} = \theta_{22} = 0$
- $H_0^{(3)} : \theta_{11} = \theta_{12} = \theta_{21} = 0$
- $H_0^{(2.1)} : \theta_{11} = \theta_{12} = 0$
- $H_0^{(2.2)} : \theta_{11} = \theta_{21} = 0$
- $H_0^{(1)} : \theta_{11} = 0$

Last rejected  $H_0$  gives  
estimated MED dose combo

	0	1	2
0		$\mu_{01}$	$\mu_{02}$
1	$\mu_{10}$	$\mu_{11}$	$\mu_{12}$
2	$\mu_{20}$	$\mu_{21}$	$\mu_{22}$

$$T_A = \sum \sum \hat{\theta}_{ij} / (C_i \hat{\sigma})$$

Average over the  $C_i$  values  
in the set of estimated  $\theta$   
for the hypothesis

$$\hat{\theta}_{ij} = \min((\bar{y}_{ij} - \bar{y}_{i0}), (\bar{y}_{ij} - \bar{y}_{0j}))$$

Strongly controls FWER (assuming monotonicity:  $\theta_{ij} \geq 0$ )

Decision Tree  
for 2x2

# Example: AVE Test for MED Dose Combo

Mean reduction in DBP

Drug A

	<b>0</b>	<b>1</b>	<b>2</b>
<b>0</b>	0	4	5
<b>1</b>	5	7	9
<b>2</b>	5	8	10

Gain  
Drug A

$$\hat{\theta}_{ij} = \min((\bar{y}_{ij} - \bar{y}_{i0}), (\bar{y}_{ij} - \bar{y}_{0j}))$$

	<b>1</b>	<b>2</b>
<b>1</b>	2	4
<b>2</b>	3	5

	<b>1</b>	<b>2</b>
<b>1</b>	$\theta_{11}$	$\theta_{12}$
<b>2</b>	$\theta_{21}$	$\theta_{22}$

$$H_0^{(4)} : \theta_{11} = \theta_{12} = \theta_{21} = \theta_{22} = 0$$

$$H_0^{(3)} : \theta_{11} = \theta_{12} = \theta_{21} = 0$$

$$H_0^{(2.1)} : \theta_{11} = \theta_{12} = 0$$

$$H_0^{(2.2)} : \theta_{11} = \theta_{21} = 0$$

$$H_0^{(1)} : \theta_{11} = 0$$



Ho	Ave( $\hat{\theta}$ )		Crit value (0.05-level)	
4	3.5	>	1.85	Reject
3	3	>	2.02	Reject
2.1	3	>	2.61	Reject
2.2	2.5	<	2.61	Accept
1	2		3.03	

Last rejected Ho gives estimated MED dose combo

# Pairwise Test for Contribution or for Therapeutic Synergy

Therapeutic Synergy: Effect of combo is greater than max effect achievable by either component ( $\mu_{A0}$  ,  $\mu_{0B}$ )

$$H_0 = H_{01}: \mu_{rs} \leq \mu_{A0} \cup H_{02}: \mu_{rs} \leq \mu_{0B}$$
$$H_1: \mu_{rs} > \mu_{A0} \text{ and } \mu_{rs} > \mu_{0B}$$

Test  $H_{01}$  and  $H_{02}$  at  $\alpha$ -level to get an  $\alpha$ -level of  $H_0$   
Under normality assumptions these can be two t-tests

A similar procedure provides test of contribution for a specific combination

# Example Test for Therapeutic Synergy

Treatment	Lowering LDL (%) (N=146-150)	Compare to combo
		p-value
Eze (10 mg)	-18.9	<0.001
Simva (80 mg)	-48.5	<0.001
Eze/Simva (10/80)	-60.2	

- > Assume 10 mg Eze and 80 mg Simva as providing maximal effect when given alone
- > the combo is better than each of these alone
- > conclude **Therapeutic Synergy at  $p < 0.001$**

[Adapted from Bays et al 2004 Clin Ther 20:1758-1773]

# Summary of MCP

- Useful even if few dose combinations are studied
- Provide assessments of contribution and of therapeutic synergy
- Closed series of tests can control FWER

# Modeling

- Dose as continuous
- Assume parametric form for relationships
- Requires better understanding/description of dose-response relationship
- May be more useful for simulations and for planning future studies
- May provide a way to investigate the mechanism of action

# Types of modeling efforts

- Polynomial response surface modeling
- Mechanistic modeling
- Empirical synergy



# Response Surface Methods (RSM)

- Polynomial is fit to the 3 dimensional surface (“French curve”)
- May provide a description of surface when number of doses is too few for fitting more mechanistic models
- No convergence issues
- No clear interpretation of parameter values
- Dangerous for extrapolation

# Polynomial Response Surface

$$E_k = \beta_0 + \beta_1 \text{dose}_{1k} + \beta_2 \text{dose}_{2k} + \beta_3 \text{dose}_{1k} \text{dose}_{2k} \\ + \beta_4 \text{dose}_{1k}^2 + \beta_5 \text{dose}_{2k}^2 + \varepsilon_k \quad , \text{ for } k\text{-th subject}$$

Fitting by OLS or linear mixed effects

May be useful to scale doses to -1 to +1

- avoids numerical problems
- allows each drug to be evaluated in an equivalent manner .

$$X_{ik} = (\text{dose}_{ik} - \text{mean}(\text{dose}_{ik})) / (\text{max}(\text{dose}_{ik}) - \text{mean}(\text{dose}_{ik}))$$

drugs  $i=1,2$

# Example Combo Drug Study

Two drugs dosed in factorial combinations  
e.g. 16 parallel groups of subjects,  
2 drugs: atorvastatin (AD), gemcabene (GD)  
15 subjects per group

<b>AD</b>	<b>GD</b>	<b>AD</b>	<b>GD</b>	<b>AD</b>	<b>GD</b>	<b>AD</b>	<b>GD</b>
0	0	0	300	0	600	0	900
10	0	10	300	10	600	10	900
40	0	40	300	40	600	40	900
80	0	80	300	80	600	80	900

[Herman et al 2005, PAGE Meeting, Pamplona, Spain]

# Polynomial Response Surface Fit

Parameter	Estimate	SE	p-value
Intercept	-3.03	6.78	0.656
Atorva (AD)	-1.18	0.124	<0.0001
Gemca (GD)	-0.0295	0.00984	0.003
AD*AD	0.00935	0.00140	<0.0001
AD*GD	0.000203	8.80E-05	0.0222
GD*GD	1.18E-05	1.02E-05	0.245
Baseline LDL	-0.0791	0.0353	0.0259

Response variable is percent change in LDL

Parameters significant except quadratic term on GD (this term could be dropped)

# Predicted Maximum Effective Combo

Stationary point: AD = 54.8 mg, GD = 776 mg;  
Predicted value: -60.8 (% change in LDL)

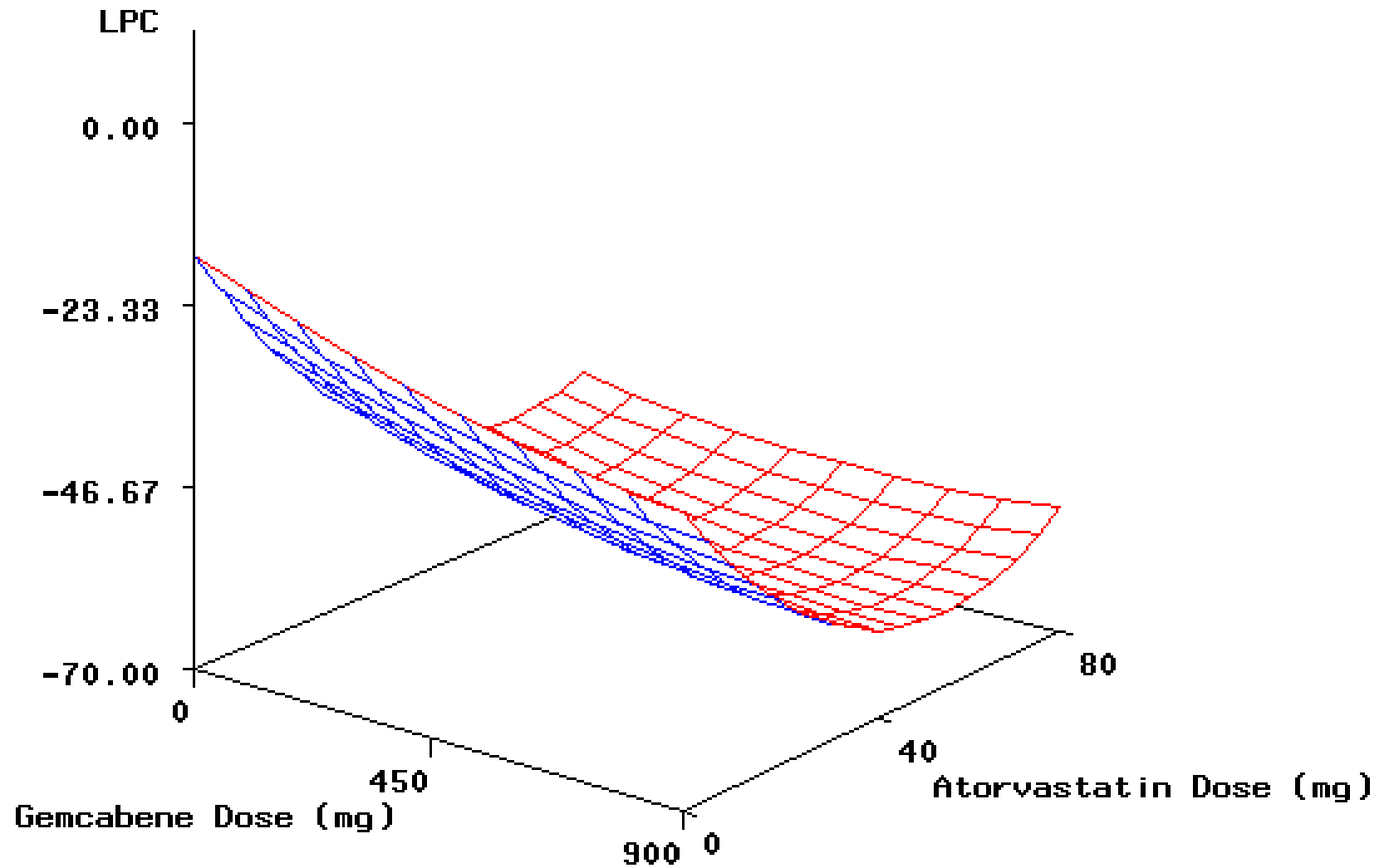
For confidence region on dose levels for maximum combo see:  
Peterson et al 2002 Biometrics 58:422-431

## By comparison

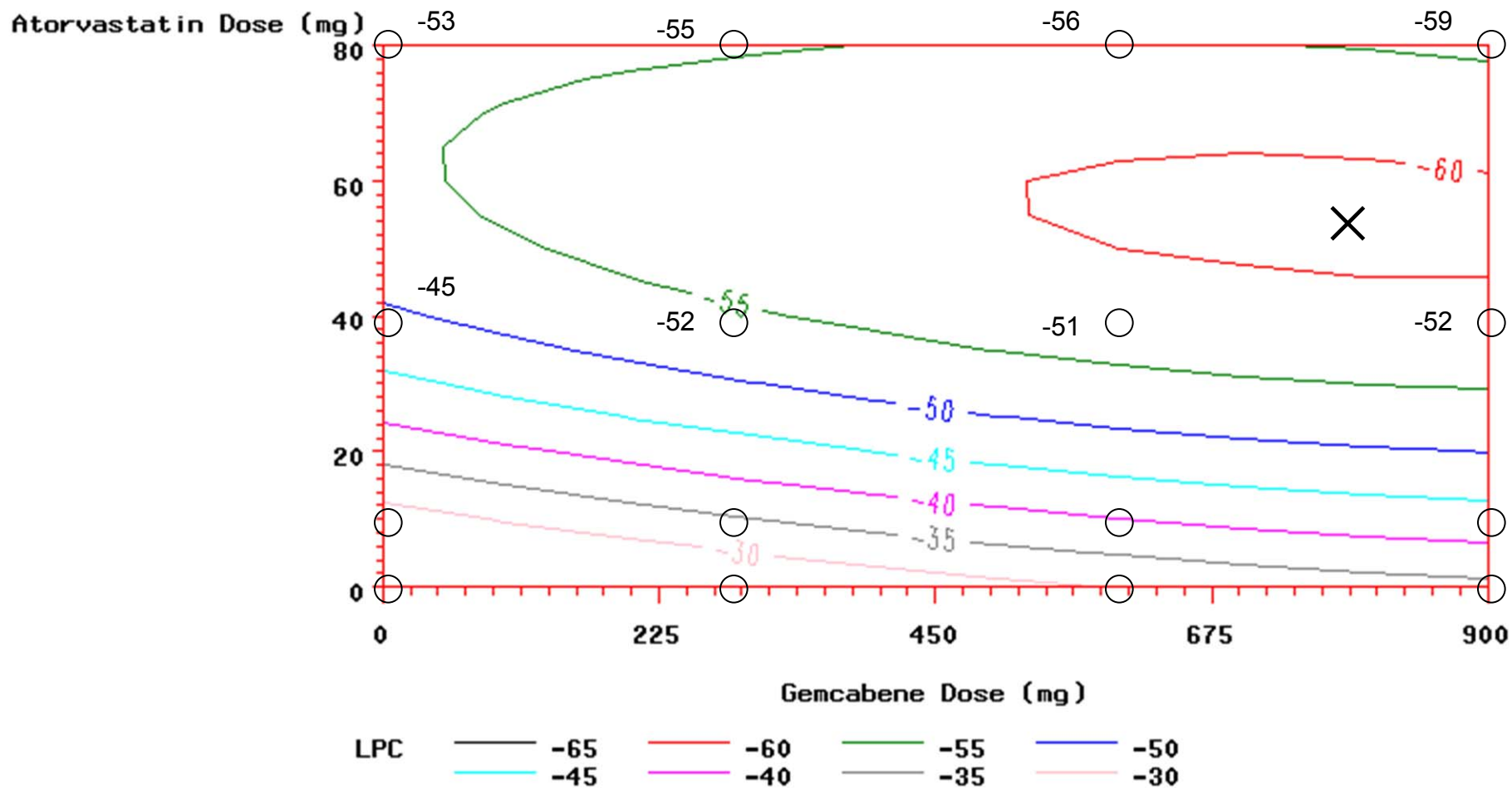
Vytorin (eze/simva) *	10/10	10/20	10/40	<b>10/80</b>
Lowering of LDL(%)	-45	-52	-55	<b>-60</b>
Atorvastatin *	10 mg	20 mg	40 mg	<b>80 mg</b>
Lowering of LDL(%)	-39	-43	-50	<b>-60</b>
Gemcabene	300 mg	600 mg	900 mg	
Lowering of LDL(%)	-18	-25	-31	

[\* Product labels (PDR) ]

# 3-D Surface Plot



# Contour Plot



# Summary of Polynomial RSM

- Provide a description of 3-D effect surface
- Allow for interpolation and prediction of optimal dose combination
- Parameter values have little meaning
- Need to check for lack of fit and consider relevance to physiology



# Mechanistic Modeling

- Drug1 + Drug2 + System = Combined Response
- Quite elaborate system models have been devised to describe and predict biology and the actions of drug.
- For illustrative purposes we will discuss only a few very simple models.

# Indirect Response Models

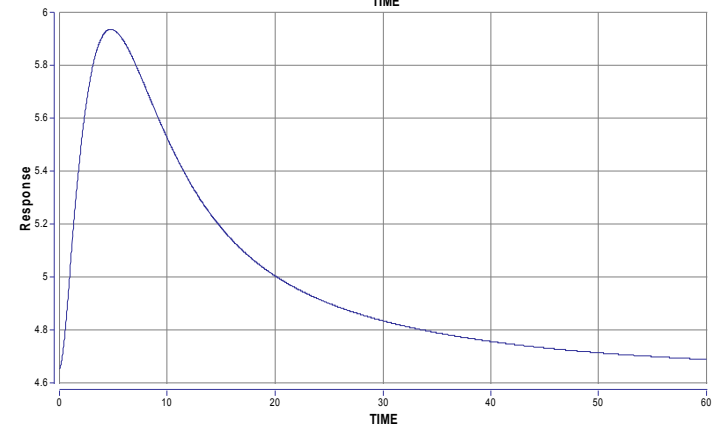
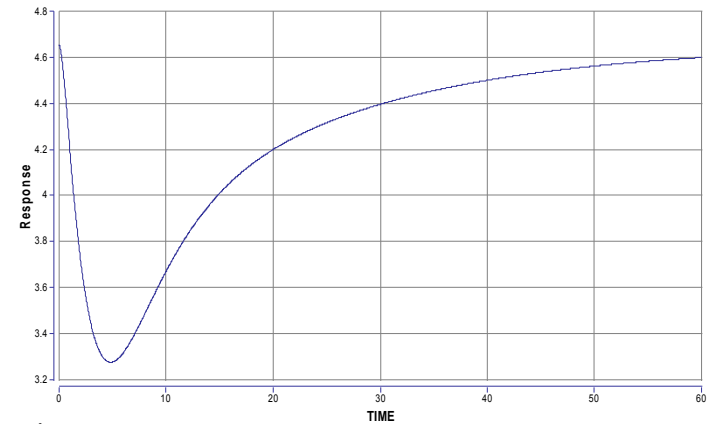
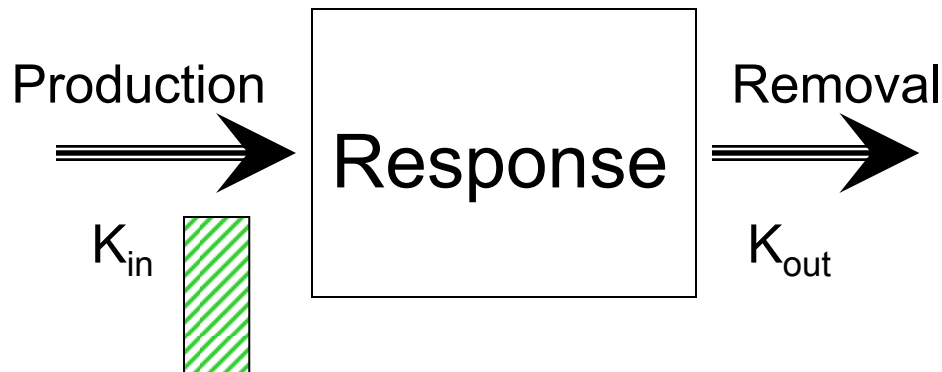
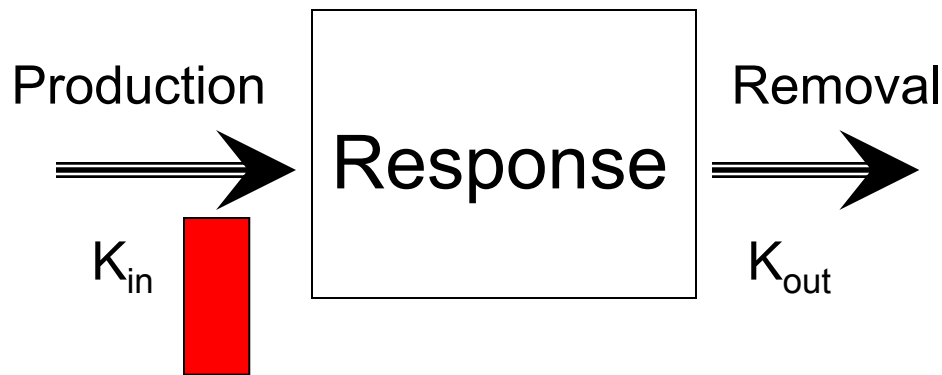
## Homeostatic System



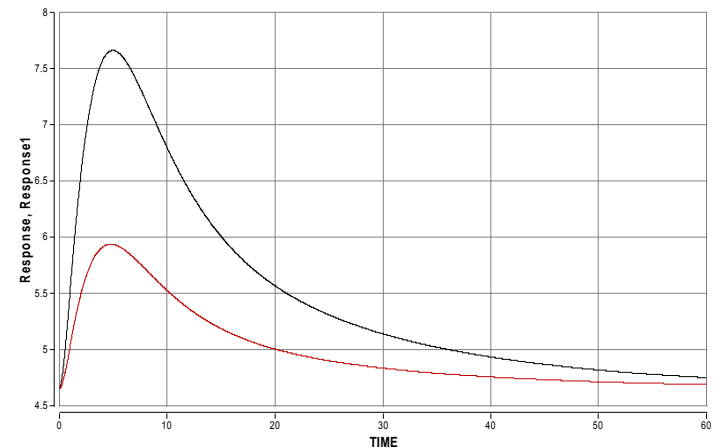
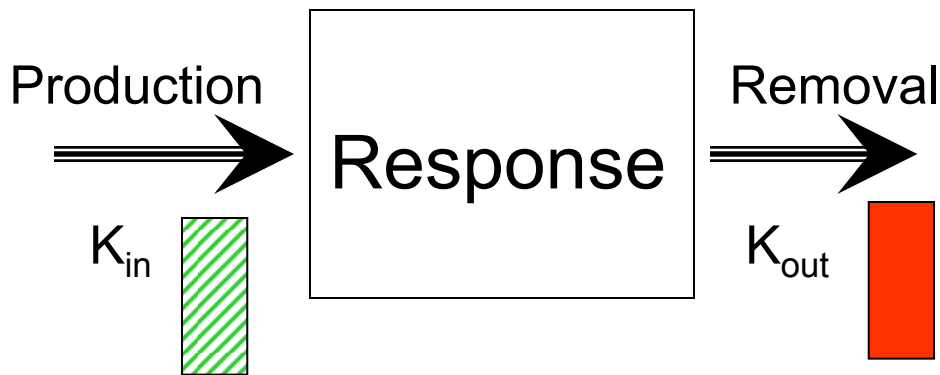
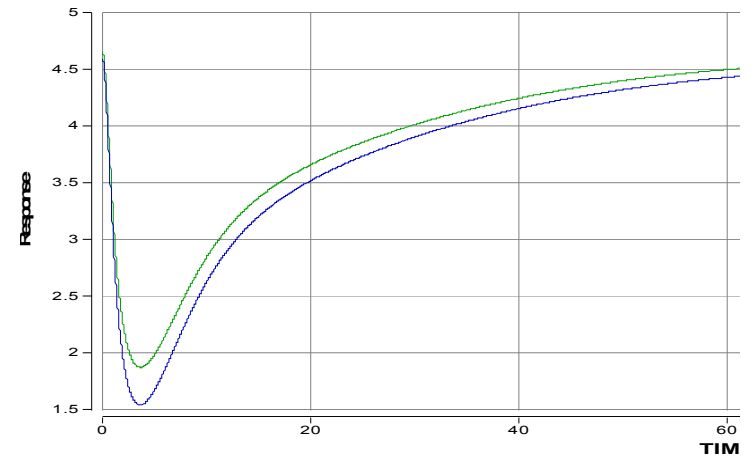
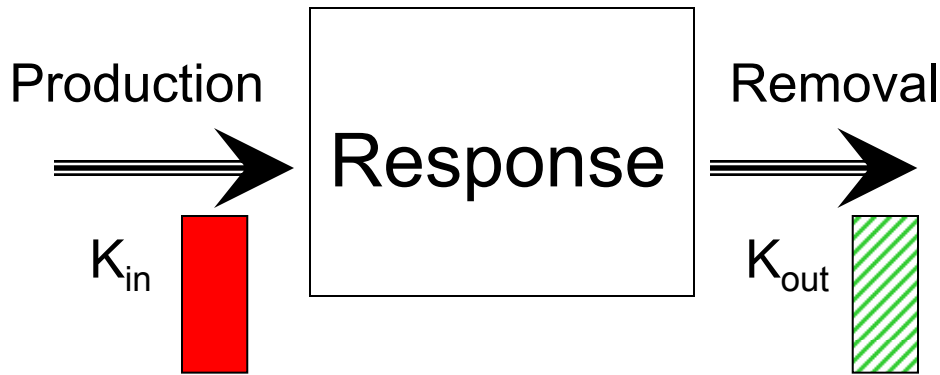
System is in equilibrium until addition of drug perturbs the system.

# Indirect Response Models

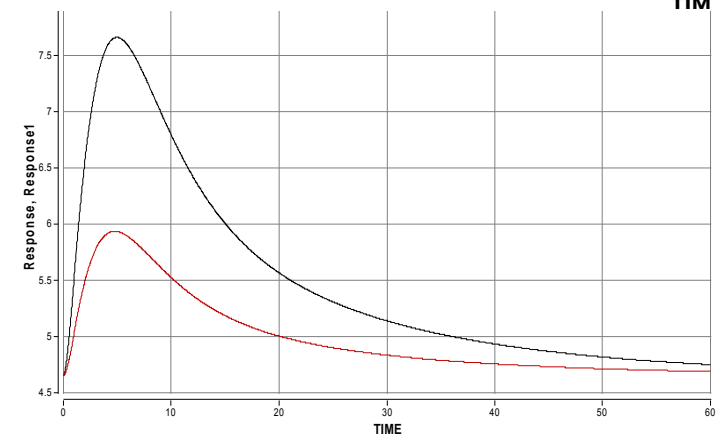
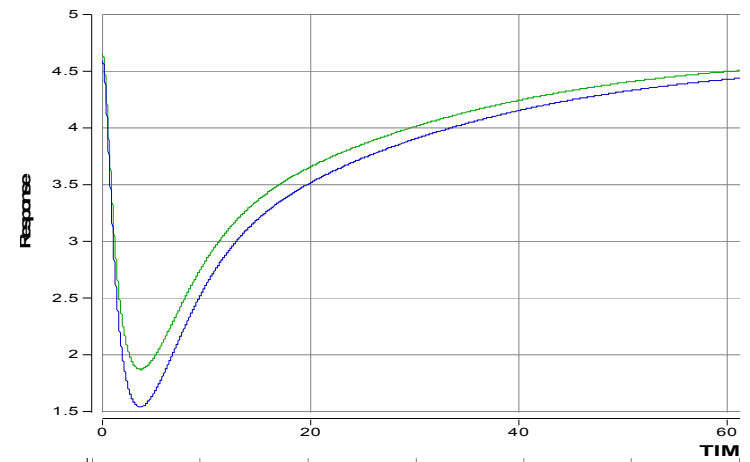
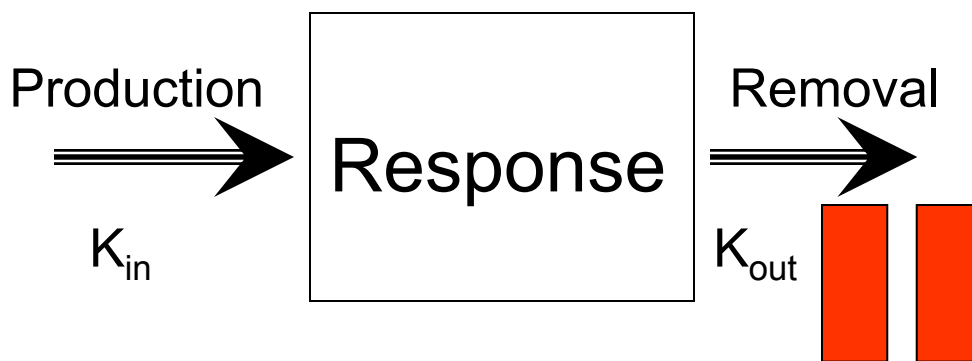
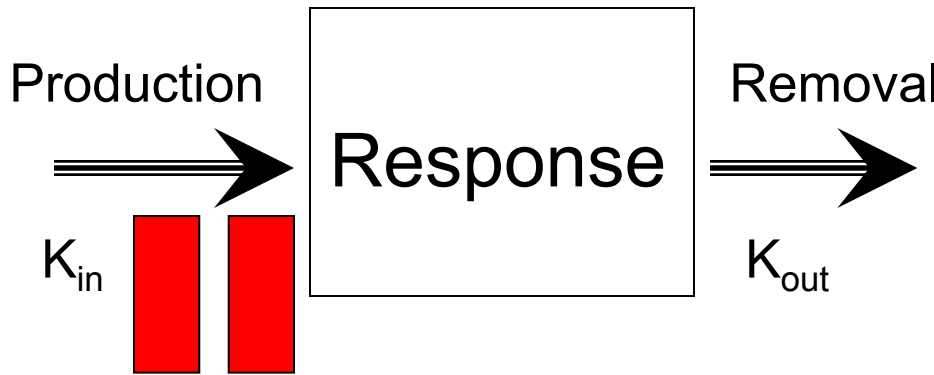
## System and Action of Drug 1



# Joint action involving different processes



# Joint action involving the same process (non-competitive)



# Steady-state Dose-Response

$$C_{ss} = \text{Dose} \cdot F / (\text{CL} \cdot \tau) = (\text{Dose} / \tau) / (\text{CL} / F)$$

$$\frac{dR}{dt} = k_{in} \{1 + H_1(C_1)\} - K_{out} \{1 + H_2(C_2)\} R$$

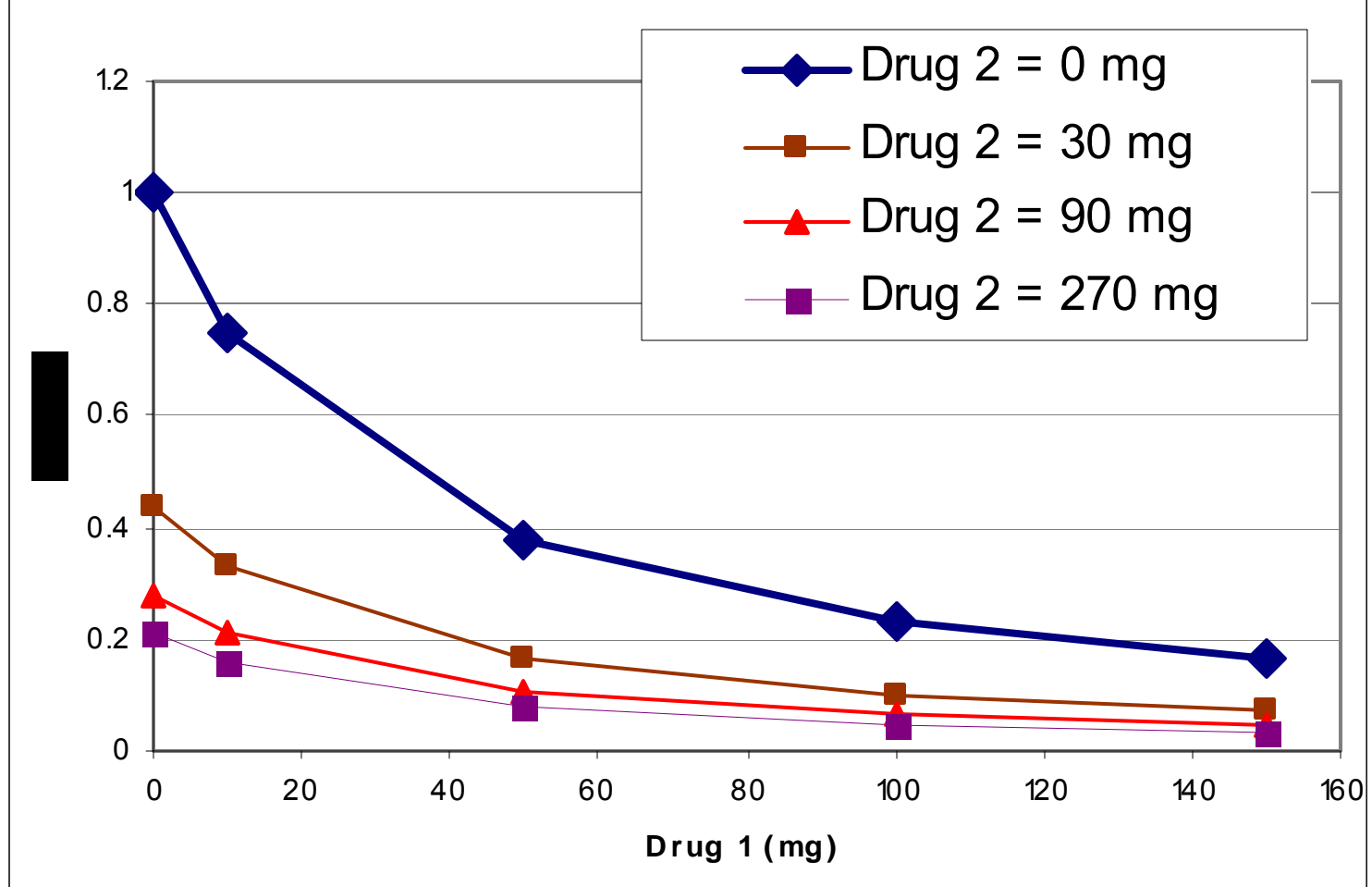
where  $H_n(C_i) = - \text{Imax}_i C_i / (\text{IC50}_i + C_i)$   
 or  $= \text{Smax}_i C_i / (\text{SC50}_i + C_i)$

then 
$$\frac{R_{ss, \max}}{R_o} = \frac{1 + H_1(C_1)}{1 + H_2(C_2)} \quad \text{for different processes}$$

or 
$$\frac{R_{ss, \max}}{R_o} = \{1 + H_1(C_1)\} \{1 + H_2(C_2)\} \quad \text{if both on } K_{in}$$

and 
$$\frac{R_{ss, \max}}{R_o} = \frac{1}{\{1 + H_1(C_1)\} \{1 + H_2(C_2)\}} \quad \text{both on } K_{out}$$

## Action on Different Processes



Simulated data shows that amount of interaction varies with dose levels

# Summary of Mechanistic Modeling

- Parameter values have biological meaning
- Best done with time course data and data from several dose levels
- Can lead to a better understanding of the mechanism of action



# Empirical Assessment of Synergy

“The Search for Synergy”

something more than expected

Starts with choice of “what is expected”

- Loewe Additivity
- Bliss Independence

[Greco et al 1995]

# Loewe Additivity

Implies that each drug contributes to joint action independently in accord with its individual potency.

$$d_1 + d_2 \rho = \text{dose of drug 1 equiv to } (d_1, d_2) \text{ , where } \rho = \text{rel. potency}$$

$D_1$  and  $D_2$  are equi-effective doses of two individual drugs  
( $d_1, d_2$ ) dose pairs that in combination give the same effect as dose  $D_1$  alone or dose  $D_2$  alone

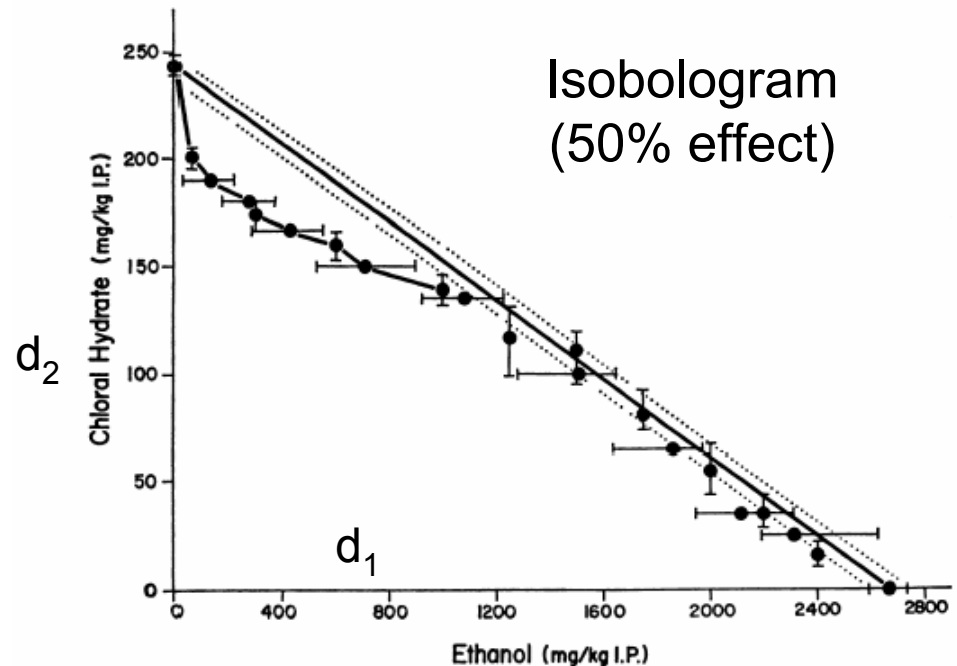
$$\frac{d_1}{D_1} + \frac{d_2}{D_2} = I$$

$I = 1$ , implies additivity

$I < 1$ , synergism

$I > 1$ , antagonism

Can roll in dose response functions for each drug alone and estimate them in pooled dataset simultaneously with  $I$  (using a root finder). URSA method: Greco et al 1990



[Gessner and Cabana. 1970. J Pharm Exp Ther. 174: 247-259]

# Bliss Independence (one of many forms)

Combined response  $C$  for two single compounds  
(Drugs 1 and 2) with effects  $A$  and  $B$  is

$$C = A + B + \alpha A * B * 0.01$$

$\alpha = 1$  for independence

$\alpha < 1$  for synergy

$\alpha > 1$  for antagonism

where each effect is expressed as  
a percentage decrease (-100,0)

Can roll in dose response functions for each  
drug alone, e.g. Effect =  $E_{max} * \text{Dose} / (\text{ED}_{50} + \text{Dose})$

Example:

Drug 1 causes a 50% decr in LDL

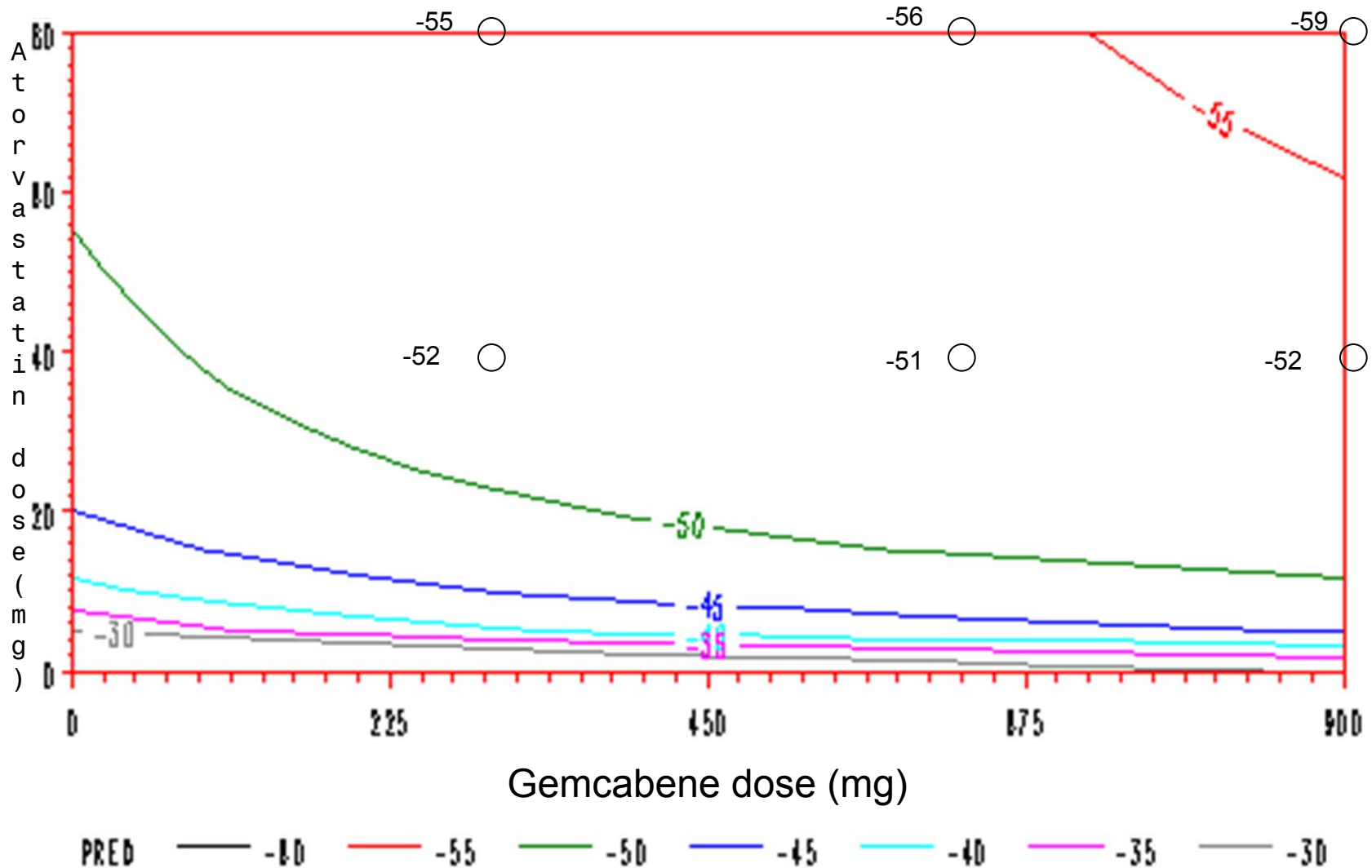
Drug 2 causes a 30% decr in LDL

If acting independently

100mg/dL  $\Rightarrow$  50mg/dL  $\Rightarrow$  35mg/dL  
(a 65% decr)

$$(-50) + (-30) + (1)(1500)(0.01) = -65$$

# Atorvastatin-Gemcabene Bliss Independence Model



# Assumptions of Empirical Synergy Methods

## Both Loewe additivity and Bliss independence:

- interaction relationship is the same quality and magnitude at all dose levels of both components

## Loewe additivity:

- the individual dose-response curves have “parallelism”, e.g.  $E_{max1} = E_{max2}$  and shapes of effect curves are similar

Extensions overcome the these issues by increasing number of parameters  
White et al (2003): adds up to ~30 parameters  
Kong and Lee (2006): adds up to 9 parameters

# Empirical Synergy versus Therapeutic Synergy

	Bliss independence factor ( $\alpha$ ) *	Loewe additivity factor (I)	Empirical Synergy (Bliss/Loewe)	Therapeutic Synergy
Atorva-Gemca	1.69 (1.49, 1.88) antagonism	0.696 (0.448, 0.944) synergy	antagonism / synergy	No
Simva-Eze	1 independence	Not available	independence / not assessed	Yes

[ \* Mandema et al 2005]

## versus Contribution

Atorva-Gemca	Dose	10/600	40/900	80/900	N = ~15
	p-value	0.02	0.05	0.19	
Simva-Eze **	Dose	10/10	40/10	80/10	N= ~150
	p-value	<0.001	<0.001	<0.001	

[ \*\*Adapted from Bays et al 2004]

# Summary of Empirical Synergy Methods

- Provides a way to combine the dose response curves of each drug alone and predict the response surface of the combination
- Gives one number to represent synergy level
- Applicable where relationships between dose and effect appear simple and direct
- Extensions are available to broaden the assumptions
- Lack of empirical synergy does not imply that the two drugs do not contribute
- Empirical synergy is unrelated to therapeutic synergy.

# Conclusions

- Assessment of dose response for combo is best done with good prior understanding of component drugs
- MCPs exist for analysis of studies with few dose combinations
- RSM and mechanistic modeling methods allow for more complete description and understanding of dose response
- Empirical synergy may not predict therapeutic synergy or contribution