

***Combining Phase IIb and Phase III
Clinical Trials***

Christopher Jennison

Department of Mathematical Sciences,

University of Bath, UK

<http://people.bath.ac.uk/mascj>

Partnerships in Clinical Trials,

Brussels, November 2006

Plan of talk

1. Separate trials for Phase IIb (dose finding) and Phase III (confirmation)

Making use of a break between phases

Implications of delay in reaching a positive conclusion

2. Seamless transition: Joint planning of Phase IIb and Phase III trials

Writing a single protocol

“Non-statistical” gains of a combined approach

Statistical methodology for combining phases

Efficiency gains from using Phase IIb data in a combined analysis

3. Further good practices for Phase IIb/Phase III trials

1. Running separate Phase IIb and Phase III trials

Phase IIb

The trial compares several dose levels of a treatment with a control in order to select a dose and provide evidence of improvement against the control.

A Phase III

The trial is run as a confirmatory study to demonstrate superiority against control of the treatment selected in Phase IIb.

Stages:

Write Phase IIb protocol, seek ethical & regulatory approval, (FDA, IRBs, ...)

Run Phase IIb, analyse data, reach conclusions.

Write Phase III protocol, seek ethical & regulatory approval, (FDA, IRBs, ...)

Run Phase III, analyse data, reach final conclusion.

Planning the Phase III trial

Planning the Phase III trial after Phase IIb allows investigators to make use of information gained in Phase IIb.

They may decide to modify:

Treatment definition,

Target population,

Primary endpoint,

Sample size.

Positive results in Phase IIb will help recruitment for participation in Phase III.

But, planning and gaining approval for the Phase III trial can be time-consuming.

If the final outcome is positive, the sooner this conclusion is reached, the better.

2. Joint planning of Phase IIb and Phase III trials

Requirements

A single protocol for the combined Phase IIb and Phase III trials.

Rules for a committee managing the trials to follow as they:

Decide whether to proceed to Phase III,

Select the treatment, with regard to both efficacy and safety outcomes,

Respond to information from Phase IIb, e.g., using estimated response variance to set Phase III sample size.

NB, expect everyone else to be blinded to the Phase IIb results.

Potential benefits

Eliminating the “white space” between phases,

Gaining efficiency from using Phase IIb data in the Phase III analysis.

When to combine Phase IIb and Phase III data?

No:

1. A Phase IIb trial providing evidence of efficacy of the selected treatment relative to the control can sometimes be used in place of a second confirmatory study.

If this is a possibility, the benefit will be much greater than that gained from using some of the Phase IIb data in Phase III.

2. A Phase IIb trial may use a more rapid patient response, such as a bio-marker or an earlier measurement of the long term outcome.

If the primary endpoint for Phase III is not available for Phase IIb patients, the data cannot easily be combined.

Possibly yes when:

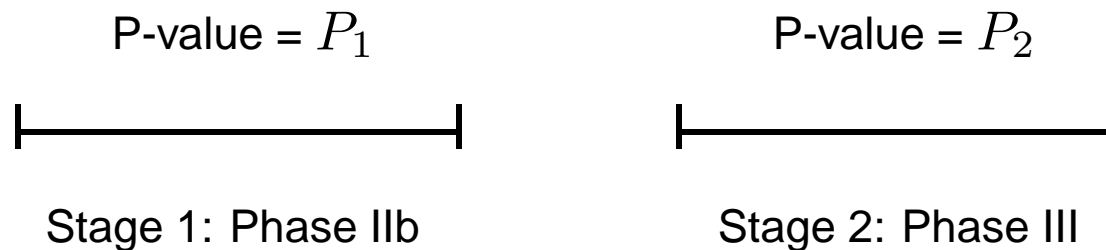
Treatment, patient population and endpoint are the same in both cases.

The efficiency gains are sufficient to make the exercise worthwhile.

3. Combining Phase IIb and Phase III data

Data from Phase IIb and Phase III trials can be analysed together using a combination test, just as in flexible adaptive designs.

For example, following Bauer and Köhne (*Biometrics*, 1994), one calculates a P-value from both “stages” and combines these values.



The study design must ensure P_1 and P_2 are independent and have $U(0, 1)$ distributions under the null hypothesis.

Then, we can combine P_1 and P_2 through Fisher's test and reject the null hypothesis for low values of $P_1 P_2$.

Combining Phases IIb and III: Bauer and Köhne, method (a)

Let $\theta_i, i = 1, \dots, n$, denote the effect size of dose level i vs the control treatment.

Obtain P_1 from a test of $H_{01}: \theta_1 = \dots = \theta_n = 0$ using Phase IIb data.

One might, for example, test for a positive trend in effect as dose increases.

Select dose level i^* , then use Phase III data to test $H_{02}: \theta_{i^*} = 0$, yielding P_2 .

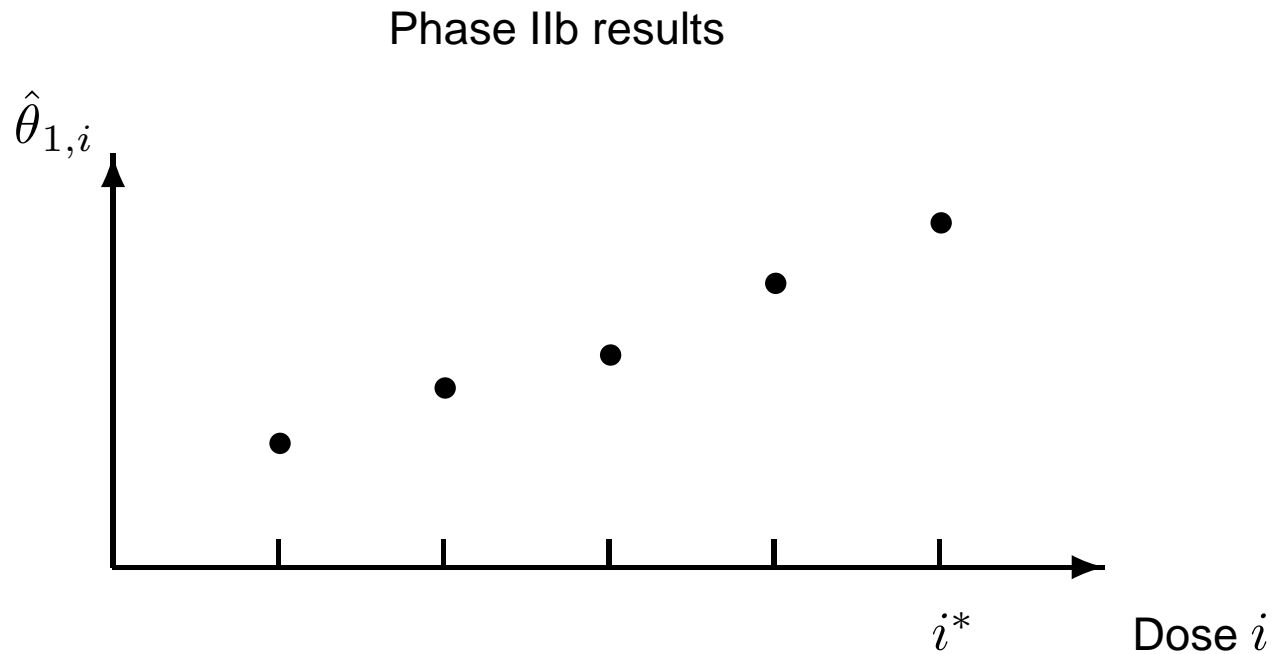
NB, P_2 is still $U(0, 1)$ under H_{02} , even if the Phase III design is adapted to Phase IIb findings.

The combined test based on $P_1 P_2$ has overall null hypothesis equal to the intersection of H_{01} and H_{02} , i.e.,

$$H_0 = H_{01} \cap H_{02} = H_{01}.$$

Does rejection of H_0 imply $\theta_{i^*} > 0$?

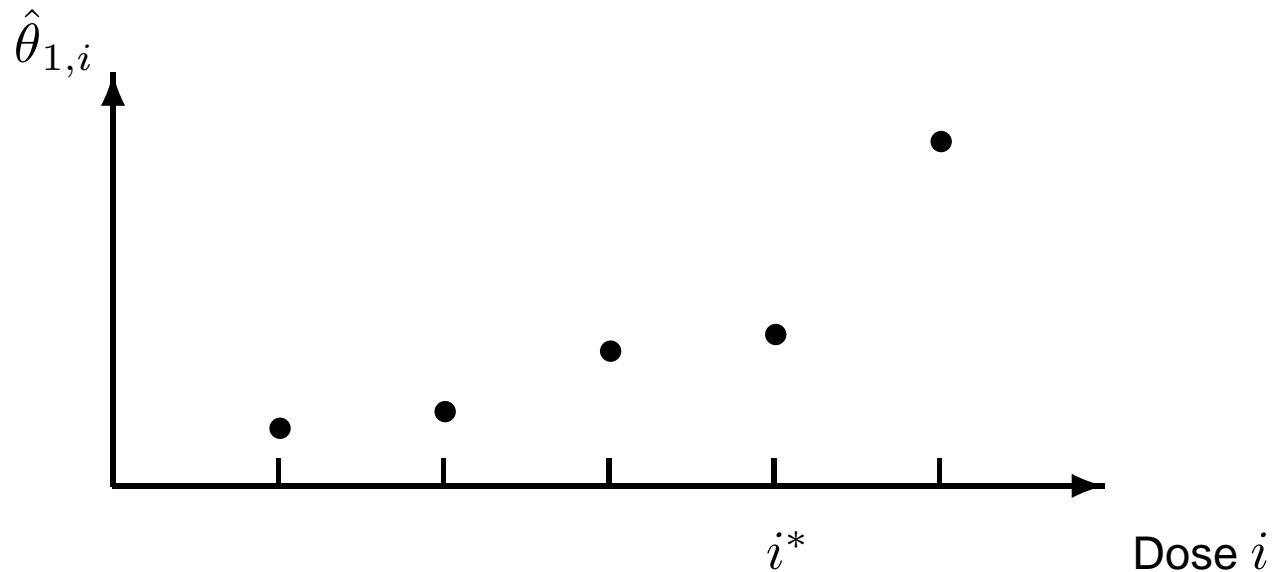
Bauer and Köhne, method (a)



Clear evidence of a simple dose response relationship in Phase IIb supports rejection of $H_{02}: \theta_{i^*} = 0$, as well as of $H_{01}: \theta_1 = \dots = \theta_n = 0$.

Bauer and Köhne, method (a)

Phase IIb results



Evidence of a dose response relationship is dominated by results at the highest dose, but this dose is not selected for Phase III due to safety problems.

Rejecting $H_0 = H_{01} \cap H_{02}$ does *not necessarily* imply $\theta_{i^*} > 0$.

Combining Phases IIb and III: Bauer and Köhne, method (b)

See also Bretz, Schmidli et al. (*Biometrical Journal*, 2006).

Procedure

Select dose level i^* to advance to Phase III.

Hope to reject $H_{i^*}: \theta_{i^*} = 0$ at significance level α , with allowance for the multiple comparisons involved in Phase IIb.

Formally

Define $H_i: \theta_i = 0$ for $i = 1, \dots, n$.

We want to control the **family wise error rate**, i.e., for all $(\theta_1, \dots, \theta_n)$,

$$Pr\{\text{Reject any true } H_i\} \leq \alpha.$$

Then, the probability of falsely claiming significance for the selected i^* is at most α .

Bauer and Köhne, method (b)

To protect the family wise error rate, use the **Closure Principle**:

For each subset I of $\{1, \dots, n\}$, define a level α test of the intersection hypothesis

$$H_I = \bigcap_{i \in I} H_i.$$

The simple hypothesis $H_j: \theta_j = 0$ is rejected if, and only if, H_I is rejected for every set I containing index j .

Proof of overall error rate

Let \tilde{I} be the set of all true hypotheses H_i .

For a “family wise” error to be committed, $H_{\tilde{I}}$ must be rejected. Since $H_{\tilde{I}}$ is true,

$$Pr\{\text{Reject } H_{\tilde{I}}\} = \alpha$$

and so the probability of a family wise error is no greater than α .

Bauer and Köhne, method (b)

Using the Closure Principle in combination Phase IIb/Phase III trials

Phase IIb

Observe estimated treatment effects $\hat{\theta}_{1,i}$, $i = 1, \dots, n$.

Select treatment i^* to go forward to Phase III.

Treatment i^* will have a high estimate $\hat{\theta}_{1,i^*}$ and good safety outcomes.

Phase III

Test treatment i^* against control.

Analyse the full collection of data, from both phases, in the hope of rejecting $H_{i^*}: \theta_{i^*} = 0$, and so establishing the effectiveness of dose i^* .

Bauer and Köhne, method (b)

Using the Closure Principle in combination Phase IIb/Phase III trials

Formally:

In order to reject $H_{i^*}: \theta_{i^*} = 0$, we need to reject each intersection hypothesis H_I with $i^* \in I$ at level α , based on combined Phase IIb and Phase III data.

Intuitively:

Dose i^* is chosen for the good results observed at this dose in Phase IIb.

We must adjust for this selection effect when adding the Phase IIb data on dose level i^* to the final analysis after Phase III.

Under a global null hypothesis of no treatment effect at any dose, the Phase IIb data on dose i^* should be viewed as the *best results out of n ineffective doses*, rather than typical results at a single, pre-specified dose.

Bauer and Köhne, method (b)

Testing an intersection hypothesis $H_I: \theta_i = 0$ for all $i \in I$

- a) Need to test an intersection hypothesis.
- b) Need to combine data from two stages, Phase IIb and Phase III.

Take problem (b) first

We can use a combination test, following Bauer and Köhne (1994).

Denote the P-value for testing H_I in Phase IIb by $P_{1,I}$.

Denote the P-value for testing H_I in Phase III by $P_{2,I}$.

Reject H_I if

$$-\log(P_{1,I} P_{2,I}) > \frac{1}{2} \chi_{4, 1-\alpha}^2.$$

Bauer and Köhne, method (b)

Testing an intersection hypothesis

(a) Testing H_I is most complex in Phase IIb

Suppose we calculate a P-value, $P_{1,i}$, for each $H_i: \theta_i = 0$.

Using the Bonferroni inequality, the overall P-value for testing H_I is m times the minimum $P_{1,i}$ over $i \in I$, where m is the number of indices in I .

Schmidli, Bretz et al. (*Biometrical Journal*, 2006) propose using Simes' (*Biometrika*, 1986) modification of the Bonferroni inequality:

Let $P_{1,(k)}$, $k = 1, \dots, m$, denote the m P-values in increasing order.

Then the P-value for testing H_I is

$$P_{1,I} = \min_{k=1, \dots, m} (m P_{1,(k)} / k).$$

Bauer and Köhne, method (b)

Testing an intersection hypothesis

Using Simes' method:

The P-value for testing H_I , where I has m elements, is

$$P_{1,I} = \min_{k=1,\dots,m} (m P_{1,(k)} / k).$$

If treatment i^* has the highest $\hat{\theta}_{1,i}$ and smallest P-value of all n treatments, we have $P_{1,(1)} = P_{1,i^*}$ in any set I containing i^* .

The term $m P_{1,(k)} / k$ with $k = 1$ becomes $m P_{1,i^*}$, the usual “Bonferroni adjusted” version of P_{1,i^*} .

Simes' method allows other low P-values to reduce the overall result: if a second treatment performs well, $P_{1,(2)} / 2$ may be smaller than P_{1,i^*} , reducing $P_{1,I}$.

Bauer and Köhne, method (b)

Testing an intersection hypothesis

(a) Testing H_I in Phase III

In order to reject $H_{i^*}: \theta_{i^*} = 0$, we need to reject each H_I with $i^* \in I$.

Only treatment i^* is studied in Phase III, so a test of such an H_I using Phase III data is based on $\hat{\theta}_{1,i^*}$ — and there is just one such test.

Hence, all H_I of interest have a common P-value in Phase III, $P_{2,I} = P_{2,i^*}$.

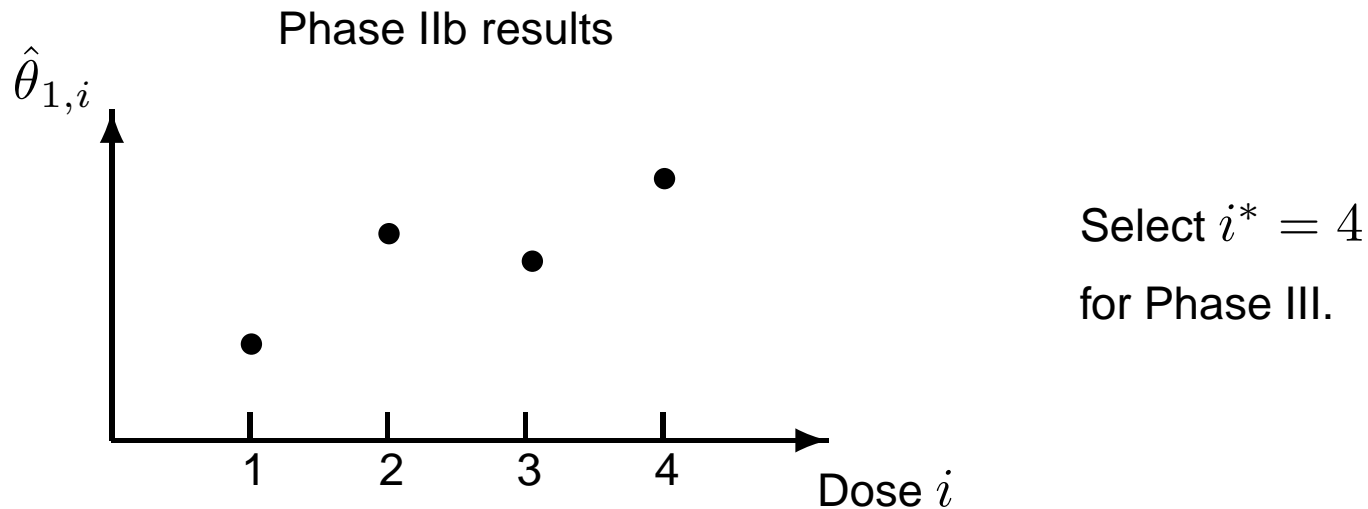
Since the combination test of H_I is based on $P_{1,I}P_{2,I}$, rejection of H_{i^*} depends on the highest value of $P_{1,I}P_{2,i^*}$.

This comes from the H_I with highest $P_{1,I}$, so the key statistic from Phase IIb is:

$$\max_I P_{1,I} \text{ over sets } I \text{ containing } i^*.$$

Bauer and Köhne, method (b)

Simes' test: Example



Suppose $P_{1,1} = 0.2$, $P_{1,2} = 0.04$, $P_{1,3} = 0.05$, $P_{1,4} = 0.03$.

For sets I containing $i^* = 4$, maximum $P_{1,I}$ comes from $I = \{1, 3, 4\}$.

Ordered P-values are $P_{1,(1)} = 0.03$, $P_{1,(2)} = 0.05$, $P_{1,(3)} = 0.2$:

$$P_{1,I} = \min_{k=1,\dots,3} (3 P_{1,(k)} / k) = 3 \times 0.05 / 2 = 0.075.$$

Summary of Bauer and Köhne, method (b)

In Phase IIb:

Select treatment i^* and carry forward $\max_{I:i^* \in I} P_{1,I}$.

In Phase III:

Test treatment i^* against control and find P_{2,i^*} .

Overall:

Combine the two P-values to see if $H_{i^*}: \theta_{i^*} = 0$ is rejected.

Flexibility:

Treatment i^* can be selected for efficacy, safety, or other factors

— not necessarily the treatment with maximum $\hat{\theta}_{1,i}$.

Efficiency:

Phase IIb data increases power or reduces Phase III sample size.

Combining Phase IIb and Phase III

Alternative methods

Two-stage procedures for treatment selection and testing include proposals by:

Thall, Simon and Ellenberg (*Biometrika*, 1988),

Schaid, Wieand and Therneau (*Biometrika*, 1990),

Stallard and Todd (*Statistics in Medicine*, 2003).

These tests are not presented in terms of the closure principle, but they can be interpreted in that framework.

The tests can be shown to control the family wise error rate.

They have flexibility and can provide good statistical efficiency.

See Jennison and Turnbull's (*Biometrical Journal*, 2006), discussion of the Bretz, Schmidli et al. papers, for further details.

Efficiency gains from combining Phases IIb and III

Comparisons have been made of the total sample size in

Separate Phase IIb and Phase III trials vs

Combined design with Phase IIb data used at the end of Phase III.

Bretz et al. (2006) consider examples where sample size per treatment in Phase IIb is equal to sample size per treatment in Phase III.

The combined study saves 30% of the total sample size for selecting one of $n = 2$ treatments and testing vs control.

But, perhaps the Phase IIb trial could count as a supporting study instead?

Todd and Stallard (*Drug Information Journal*, 2005) present an example where sample size per treatment is 25 in Phase IIb and 1400 in Phase III.

Here, savings can be at most 2% of the total sample size!

Features of alternative methods

1. *More than one treatment may be carried forward to Phase III.*

2. *Sequential monitoring:*

Inferior treatments may be eliminated early in Phase IIb.

An early decision, positive or negative, may be reached in Phase III.

We should expect significant efficiency gains from sequential monitoring, eliminating poor treatments and stopping early, either for a positive outcome or for futility.

These are likely to outweigh the gains from combining Phase IIb and Phase III data.

Features of alternative methods

Allowing more than one treatment to progress to Phase III:

Schaid, Wieand and Therneau (*Biometrika*, 1990).

Group sequential analysis of combined Phase IIb and Phase III data (with adjustment for treatment selection in Phase IIb):

Stallard and Todd (*Statistics in Medicine*, 2003).

Elimination procedures and adaptive treatment allocation:

Paulson (*Annals of Mathematical Statistics*, 1964),

Robbins and Siegmund (*JASA*, 1974),

Jennison, Johnstone and Turnbull (*Purdue Symposium*, 1982).

Totally seamless procedure

In principle, one can combine

Interim monitoring,

Treatment elimination,

Carrying several treatments into Phase III,

Early stopping for a final decision.

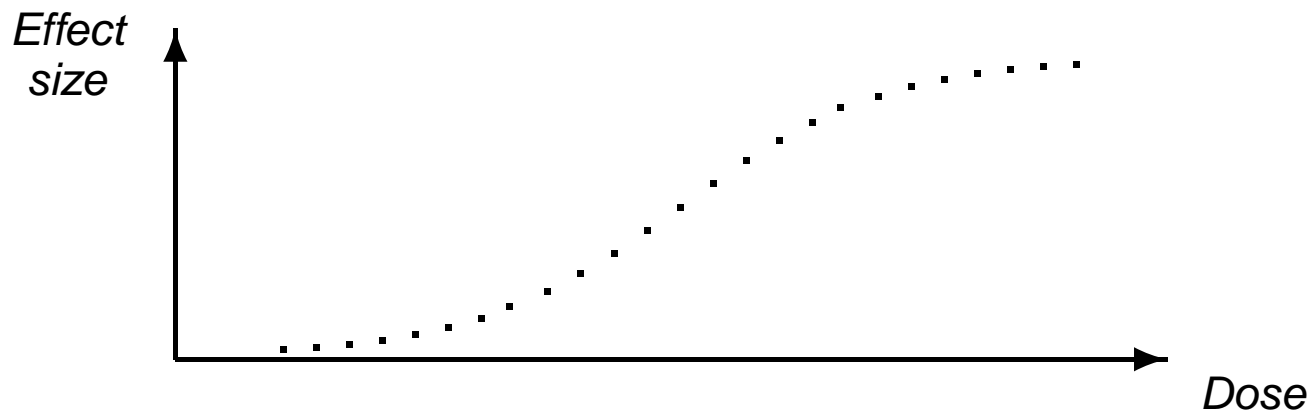
Then, the distinction between Phases IIb and III is blurred and a totally seamless procedure emerges.

Complex calculations would be required to define such a procedure meeting specified error rates and to evaluate its properties.

4. Further good practices for Phase IIb/Phase III trials

1. Modelling dose response vs control in Phase IIb

A “nice” relationship between response and dose is usually to be expected.



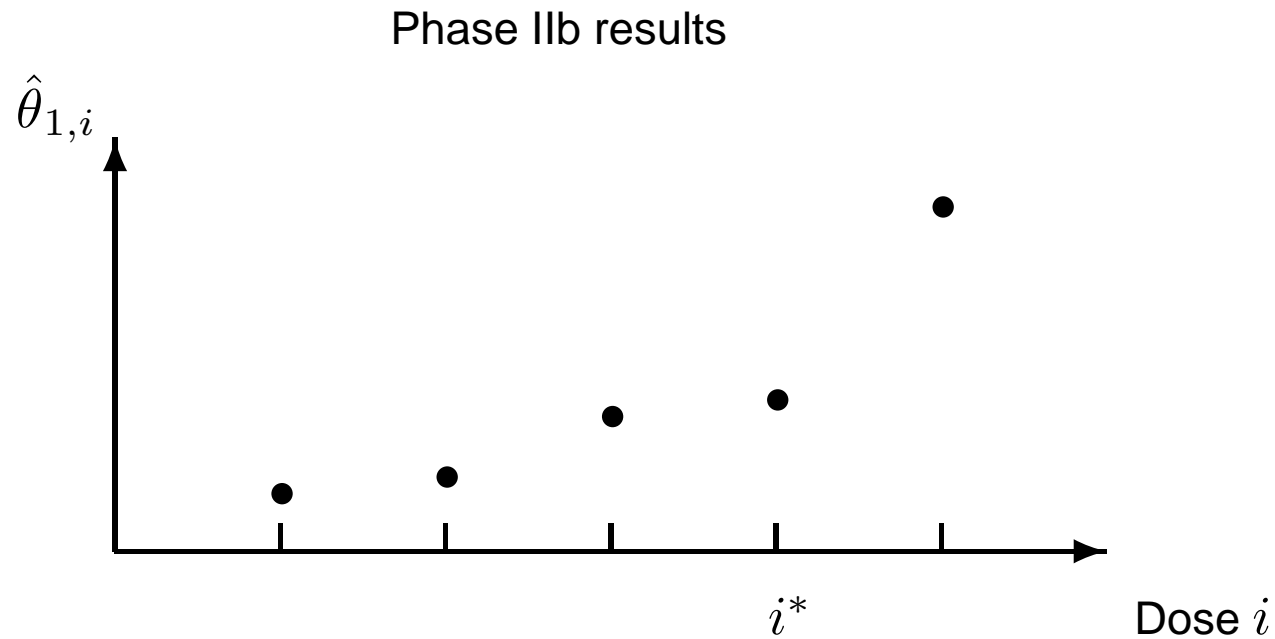
Using such a model can help in identifying the optimal dose.

The ASTIN trial (Krams et al. *Stroke*, 2003) is well worth studying for its use of modelling in the on-going design of a Phase IIb trial.

However, a model based analysis may not be viewed as providing clear cut evidence of a treatment effect at the selected dose level i^* .

Modelling dose response in Phase IIb

Previous example:



It may be that, under the assumed model, any positive trend implies positive treatment effects at all dose levels.

However, in this example, the effect size at dose i^* could still be slight and the evidence of a positive effect relies heavily on the assumed model.

Further good practices for Phase IIb/Phase III

2. Using a rapidly observable endpoint in Phase IIb

Different endpoints may be used in Phase IIb and Phase III.

A more rapidly available response (surrogate endpoint, bio-marker) may be used to select the treatment for which a long-term response is tested in Phase III.

There are important open questions:

How should efficient designs be derived in this case?

When is it useful to follow up (some) Phase IIb subjects to observe the long-term endpoint while the Phase III trial runs its course?

Does the delay involved in waiting for a long-term response change the nature of “efficient” trial designs?

***The challenging questions in this area involve practical issues
as much as statistical methodology.***