

Bayesian Adaptive CNS Trials

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THE UNIVERSITY OF TEXAS
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CANCER CENTER

Baltimore, July 18, 2006

Janet Woodcock, CDER Director :

“Improved utilization of adaptive and Bayesian methods” could help resolve low success rate of and expense of phase III clinical trials

The Coming Bayesian Tsunami of Clinical Development

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ベイズ統計学パート2

臨床開発にベイズ統計学の「津波」が到来

テキサス大学アンダーソン癌センター生物統計学科主任、ドナルド・A・ベリー博士

30年以上もの間、ドナルド・ベリー博士は臨床試験のデザインと分析におけるベイズ統計学の利用を提唱してきた。この発想は当初、行政機関や製薬業界、および臨床研究における教育を受けておらず認識の乏しい統計学者から無視された。しかし、同氏が統計学の薬剤開発への適用に情熱を傾け続けた結果、数十年後、FDAと製薬会社が同氏に耳を傾け、ベイズ的アプローチの利点を理解するようになった。ベリー博士にこの変化について聞いた。

——薬剤開発のデザインと分析において、統計は乱用されているとお考えですか。

ベリー 統計学者を含め、臨床試験のデザ



ドナルド・A・ベリー博士

OUTLINE

- **Bayesian adaptive design?**
- **Predictive probabilities in design**
- **Adaptive dose-finding:
A stroke trial**

Practical Advantages of Bayes

1. On-line learning
2. Predictive probabilities
3. Modeling
 - Hierarchical
 - Longitudinal
4. Decision analysis

Current Use of Bayesian Adaptive Designs

- MDACC (> 300 trials)
- Many device companies
(> 20 PMAs, many 510(k)s)
- All top drug companies; many
biotechs

Some Bayesian Device Applications

- Orthopedics (esp. spinal implants)
- Diagnostics & screening
- Stents
- Shunts
- Defibrillators
- Bronchial thermoplasty
- Ablation catheters
- PFO closure
- Contraceptives
- Neurostimulation

Some areas of application of Bayesian adaptive drug trials

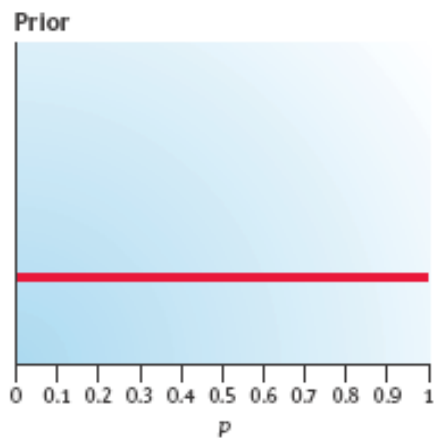
- Oncology
- Spinal Cord Inj
- Migraine
- HIV
- Rh Arthritis
- Hep C
- Lupus
- Pre-term labor
- Sepsis
- Constipation
- Diabetes
- Micturition
- Obesity
- Alzheimer's
- Stroke
- Parkinson's

Bayesian Adaptive Trials

- Stopping early (or late)
 - Efficacy
 - Futility
- Dose finding (& dose dropping)
- Seamless phases
- Population finding
- Adaptive randomization
- Ramping up accrual

Bayesian Updating

- Paired observations, T vs C
- $P(S) = P(T \text{ wins pair})$
- $H_0: P(S) = 1/2$
- Data: SFSS FSSSF
SFSSS SS



Prob:
8/12

Prob:
4/12

Predictive Probabilities

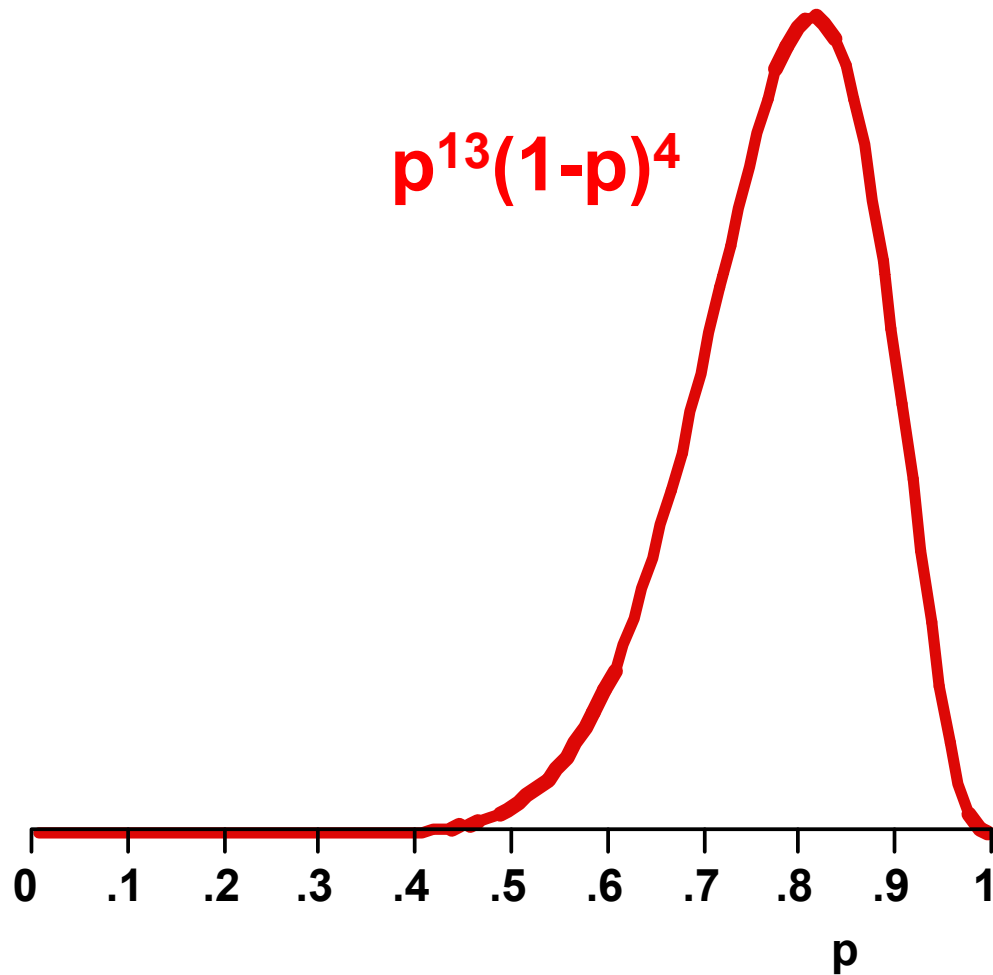
- Essential for monitoring trials
- Critical component of experimental design

**“We must ask where we are
and whither we are tending.”**

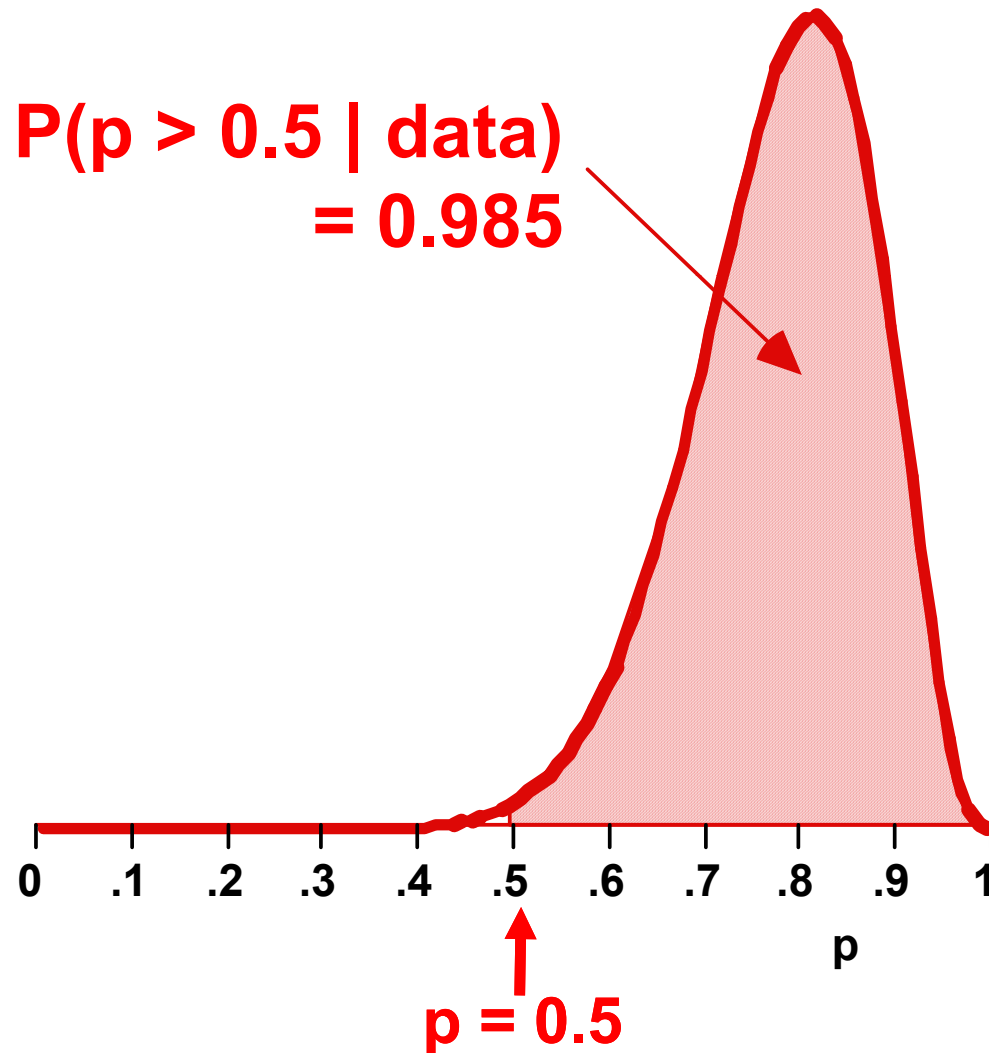
— Abraham

Lincoln

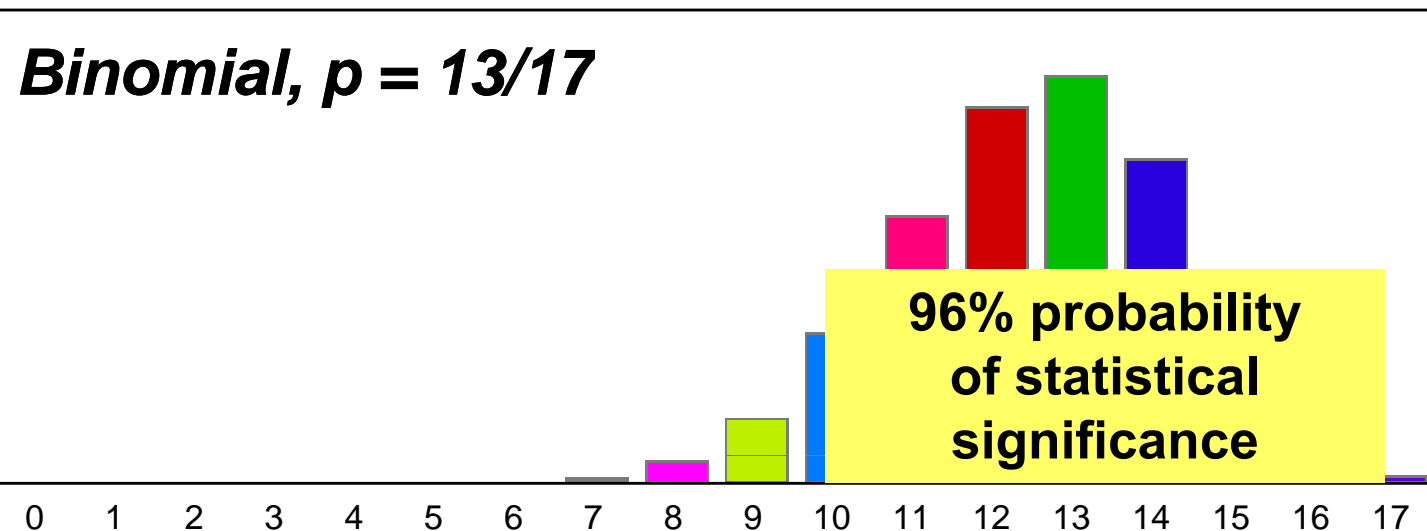
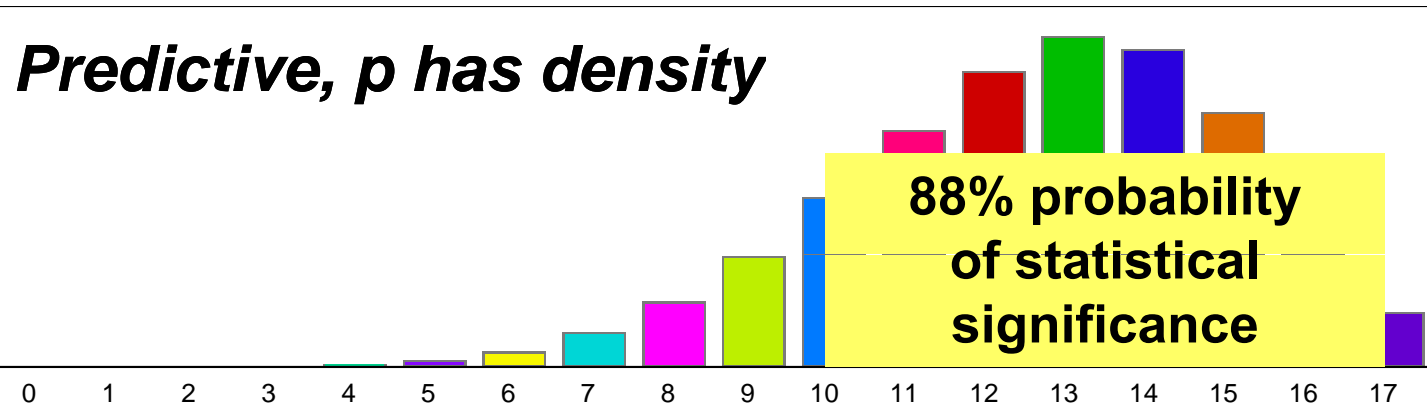
Current (posterior) for $p = P(S)$



Posterior Probability that $p > 0.5$ is Area Under Curve



Probabilities for # successes in next 17 observations



Why Bayes?

- **Smaller trials (usually!)**
- **More accurate conclusions**
- **Objective can include better treatment of patients in trials**
- **Complicated trials, but control Type I error via simulation**

Predicting Trial Results

- **Simulate**
- **Model uncertainty**
- **Incorporate info (Bayesian-wise) on various outcomes**
- **Model relationships among early and late endpoints**
- **Consider alternative designs**

Prediction in Adjuvant Breast Cancer Trial

- Patients ≥ 65 yrs old
- Capecitabine (Xeloda®) vs standard therapy (CMF or AC)
- Hypothesis: Noninferiority
- N = 600 to 1800

The NEW ENGLAND JOURNAL of MEDICINE

ESTABLISHED IN 1812

MAY 14, 2009

VOL. 360 NO. 20

Adjuvant Chemotherapy in Older Women with Early-Stage Breast Cancer

Hyman B. Muss, M.D., Donald A. Berry, Ph.D., Constance T. Cirrincione, M.S., Maria Theodoulou, M.D., Ann M. Mauer, M.D., Alice B. Kornblith, Ph.D., Ann H. Partridge, M.D., M.P.H., Lynn G. Dressler, Ph.D., Harvey J. Cohen, M.D., Heather P. Becker, Patricia A. Kartcheske, B.S., Judith D. Wheeler, M.P.H., Edith A. Perez, M.D.,

A Bayesian statistical design was used with a range in sample size from 600 to 1800 patients.

BACKGROUND

Older women with breast cancer are underrepresented in clinical trials, and data on the effects of adjuvant chemotherapy in such patients are scant. We tested for the noninferiority of capecitabine as compared with standard chemotherapy in women with breast cancer who were 65 years of age or older.

METHODS

We randomly assigned patients with stage I, II, IIIA, or IIIB breast cancer to standard chemotherapy (either cyclophosphamide, methotrexate, and fluorouracil or cyclophosphamide plus doxorubicin) or capecitabine. Endocrine therapy was recommended after chemotherapy in patients with hormone-receptor–positive tumors. A Bayesian statistical design was used with a range in sample size from 600 to 1800 patients.

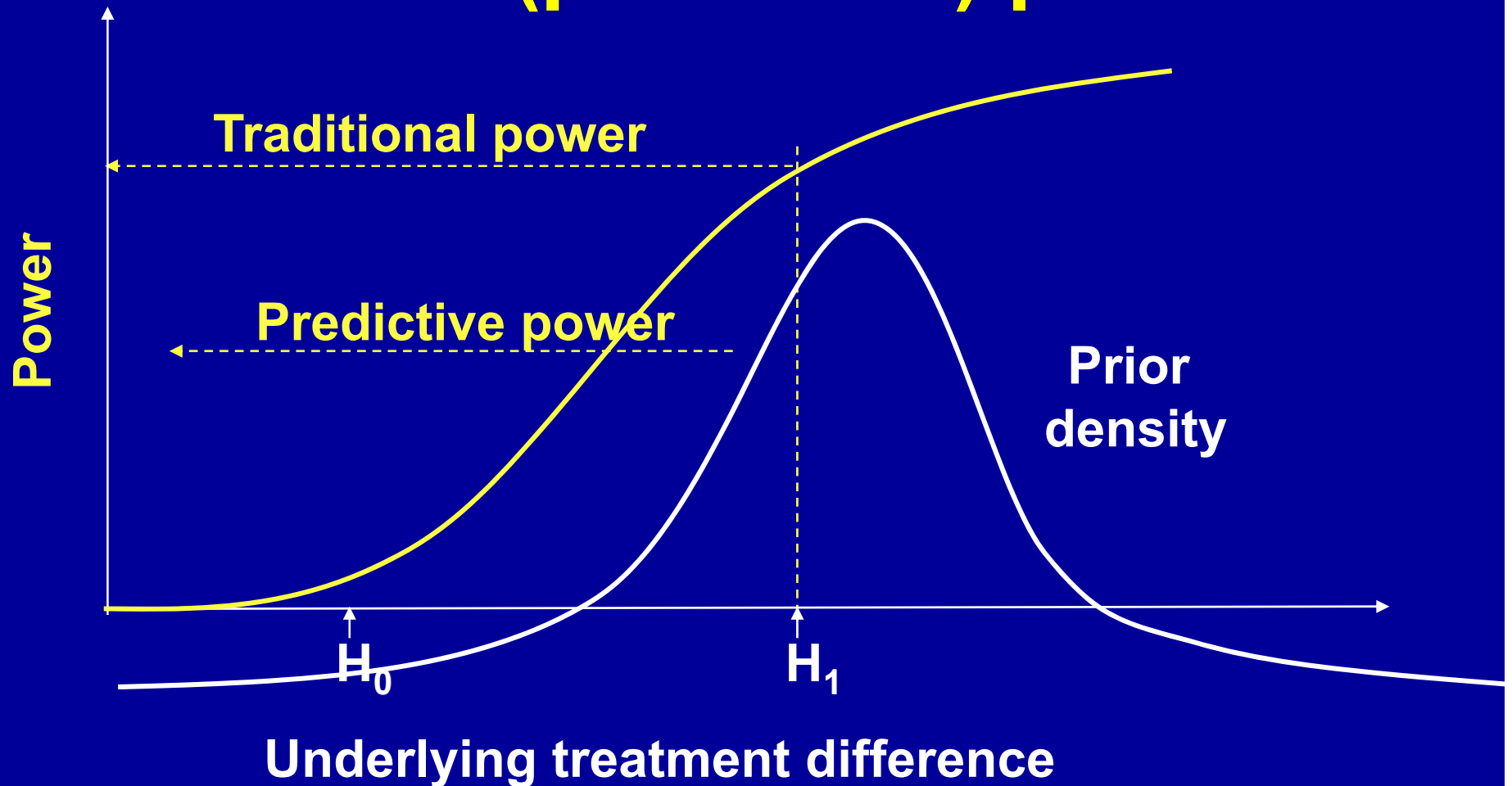
The primary end point was relapse-free survival.

From the University of Vermont, Burlington (H.B.M.); the M.D. Anderson Cancer Center, Houston (D.A.B.); the Cancer and Leukemia Group B (CALGB) Statistical Center, Duke University Medical Center (C.T.C., P.A.K.) and Duke University Medical Center (H.J.C., J.D.W., A.A.M.) — both in Durham, NC; Memorial Sloan-Kettering Cancer Center, New York (M.T., L.N., C.A.H.); CALGB, Chicago (A.M.M., H.P.B.); the Dana-Farber Cancer Institute, Boston (A.B.K., A.H.P., H.J.B., E.P.W.); the University of North Carolina, Chapel Hill (L.G.D.); the North Central Cancer Treatment Group, Rochester, Minn. (A.P.); the

**A purely statistical reason
for sorry performance of
drugs in phase III ...**

Power considerations

True (predictive) power



The NEW ENGLAND JOURNAL of MEDICINE

ORIGINAL ARTICLE

NXY-059 for Acute Ischemic Stroke

Kennedy R. Lees, M.D., Justin A. Zivin, M.D., Tim Ashwood, Ph.D.,
Antonio Davalos, M.D., Stephen M. Davis, M.D., Hans-Christoph Diener, M.D.,
James Grotta, M.D., Patrick Lyden, M.D., Ashfaq Shuaib, M.D.,
Hans-Göran Hårdemark, M.D., and Warren W. Wasiewski, M.D.,
for the Stroke–Acute Ischemic NXY Treatment (SAINT I) Trial Investigators*

N ENGL J MED 354;6 WWW.NEJM.ORG FEBRUARY 9, 2006

Results of SAINT I:

RESULTS

Among the 1699 subjects included in the efficacy analysis, NXY-059 significantly improved the overall distribution of scores on the modified Rankin scale, as compared with placebo ($P=0.038$ by the Cochran–Mantel–Haenszel test). The common odds ratio for improvement across all categories of the scale was 1.20 (95 percent confidence interval, 1.01 to 1.42).

Design of SAINT II:

- **N = 3200 (up from 1700)**
- **Power 80% for Odds Ratio 1.20**

In SAINT I:

RESULTS

Among the 1699 subjects included in the efficacy analysis, NXY-059 significantly improved the overall distribution of scores on the modified Rankin scale, as compared with placebo ($P=0.038$ by the Cochran–Mantel–Haenszel test). The common odds ratio for improvement across all categories of the scale was 1.20 (95 percent confidence interval, 1.01 to 1.42).

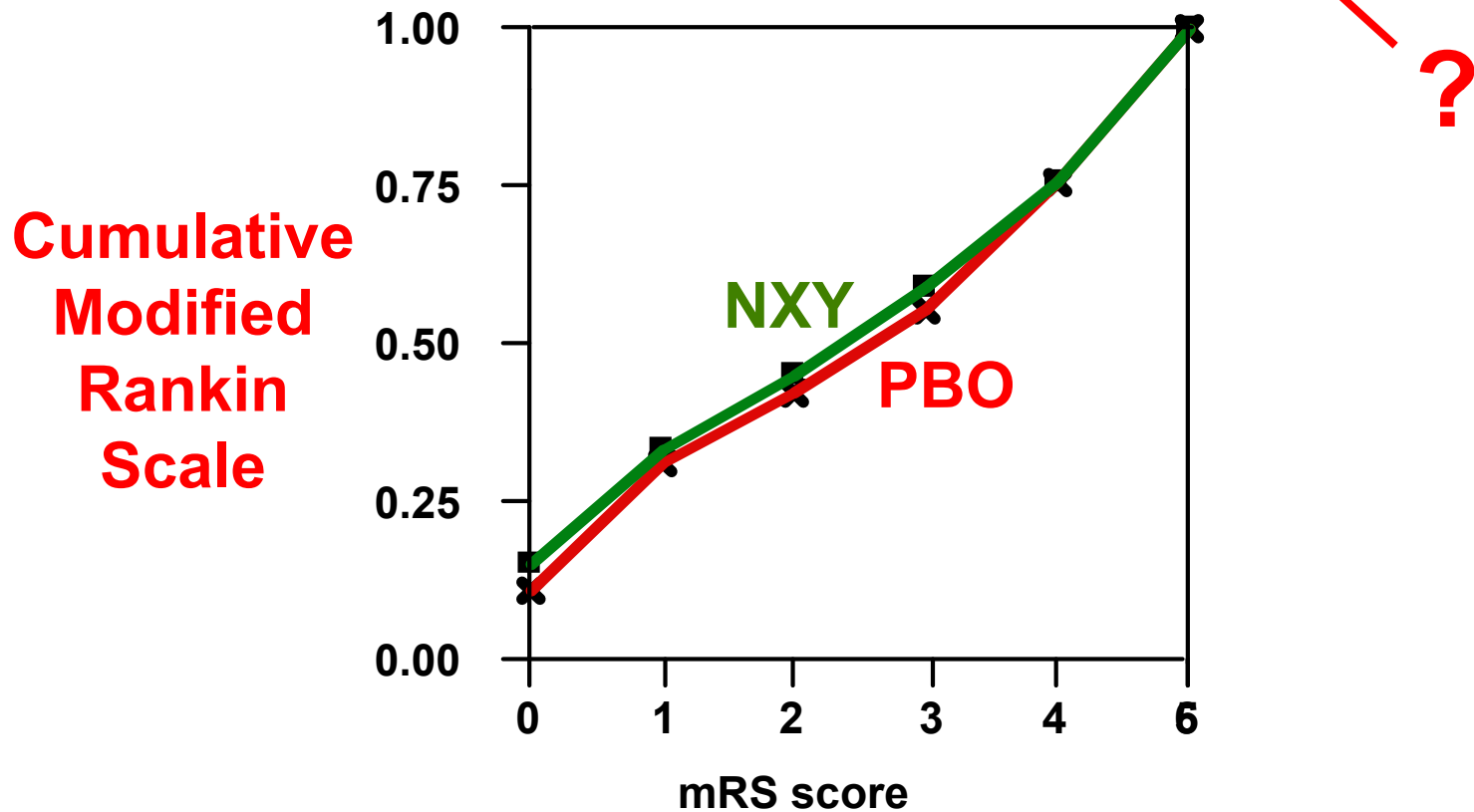


Table 2. Efficacy of the Study Drug at Day 90 or at the Last Rating.*

Outcome Variable	Placebo Group	NXY-059 Group	Difference between NXY-059 and Placebo† % or score (95% CI)	P Value
Modified Rankin scale score (primary end point)				
No. of patients	849	850		
Score — no. (%)				
0	93 (11.0)	131 (15.4)	4.4	
1	170 (20.0)	153 (18.0)	-2.0	
2	99 (11.7)	97 (11.4)	-0.3	
3	108 (12.7)	121 (14.2)	1.5	
4	175 (20.6)	144 (16.9)	-3.7	
5 (or death)	204 (24.0)	204 (24.0)	0	0.038
Change from baseline in total NIHSS score (coprimary outcome)				
No. of patients	851	851		
Score — LSM ±SE	-1.7±0.5	-1.8±0.5	-0.1 (-1.4 to 1.1)	0.86
Barthel index (dichotomized analysis)				
No. of patients	848	850		
Score, ≥95 — no. (%)	346 (40.8)	368 (43.3)	2.5	0.14
Stroke Impact Scale				
No. of patients	676	669		
Score — LSM ±SE	63.4±1.1	66.2±1.1	2.8 (-0.3 to 5.9)	0.08
EuroQoL EQ-5D (weighted index)				
No. of patients	816	819		
Score — LSM ±SE	0.43±0.013	0.47±0.013	0.04 (-0 to 0.07)	0.06
EuroQoL EQ-5D (VAS)				
No. of patients	671	670		
Score — LSM ±SE	62.0±0.9	64.5±0.9	2.5 (-0.1 to 5.0)	0.05

Not probability of null hypothesis



**Advertised power of
SAINT II = 80%**

**Naïve predictive power
of SAINT II = 60%**

**Other info from SAINT I:
My pre-probability
that SAINT II was going
to be positive: 10%**

Press release, Oct 26, 2006

“Results from the SAINT II (Stroke Acute Ischemic NXY-059 Treatment) trial: ...

NXY-059 **did not meet its primary outcome** of a statistically significant reduction in stroke-related disability, as assessed by the modified Rankin Scale (mRS) ($p=0.33$, odds ratio 0.94) compared to placebo.

“The company plans **no further development** of NXY-059 in acute ischemic stroke.”

Morals

- P-value is not probability of H_0
- Predictive power < “power”
- Use adaptive design
 - Start trial cautiously
 - Look at the data!!
 - Ask “whither we are tending”
 - Adapt

Types of adaptive trials

- **Stopping early (or late)**
 - **Efficacy**
 - **Futility**
- **Dose finding (& dose dropping)**
- **Seamless phases**
- **Adaptive randomization**
- **Identifying responding biomarker profile**
- **Ramping up accrual**

Why adaptive design?

- **Smaller trials (usually!)**
- **More accurate conclusions**
- **Better treatment of patients in trials, depending on trial goal**

REFERENCES

- Berry D (1996). *Statistics: A Bayesian Perspective*. Duxbury Press.**
- Berry D (2006). Bayesian clinical trials. *Nature Reviews Drug Discovery*.**
- Spiegelhalter D, Abrams K, Myles J (2004). *Bayesian Approaches to Clinical Trials and Health-Care Evaluation*. Wiley.**

Websites for JHU/FDA Workshop

www.prous.com/bayesian2004

**[http://www.cfsan.fda.gov/~frf/
bayesdl.html](http://www.cfsan.fda.gov/~frf/bayesdl.html)**

CDRH Bayesian Guidance

<http://www.fda.gov/cdrh/osb/guidance/1601.html>

March 12, 2001

BioCentury 100™ Indicators

Week ended 3/9/01

PRICES	VOLUME
2117.52 dn 6%	606.3M shrs up 14%

PAGE A1 OF 20

The adaptive trial computer simulations described by the authors* start with 16 possible doses or placebo. As data come in each week on patient responses, the next set of doses assigned are allowed to depend on information from patients treated previously.

Center in Houston, and co-...ails an adaptive Bayesian trial stroke drugs. It is to appear in *es in Bayesian Statistics*".

aptive trial computer simulation described by the authors start with 16 possible doses or placebo. As data come in each week on patient responses, the next set of doses assigned are allowed to depend on information from patients treated previously. Thus as data accumulate that a certain dose is more effective, more patients

There's no flexibility in the frequentist straightjacket.

— Donald Berry of M.D. Anderson

Bio...
Cov...

Bayes...
faster...
inform...

stances than more familiar frequentist statistics.

Technology Focus

Psoriasis still up for grabs/Page A7
Protocol counts in Parkinson's/Page A9

...experience with Bayesian methods among biostatisticians have limited use of the technique. In fact, Bayesian statistics have never been used for regulatory approval of therapeutics.

That is about to

***Berry, et al. Case Studies in Bayesian Statistics 2001**

CM.../Page A10

...dose finding phase of a therapeutic...
...using an adaptive Bayesian design. The

...The investigator remains blinded to the dose...
...("Adaptive Dose Finding Trial

Stroke and adaptive dose-response

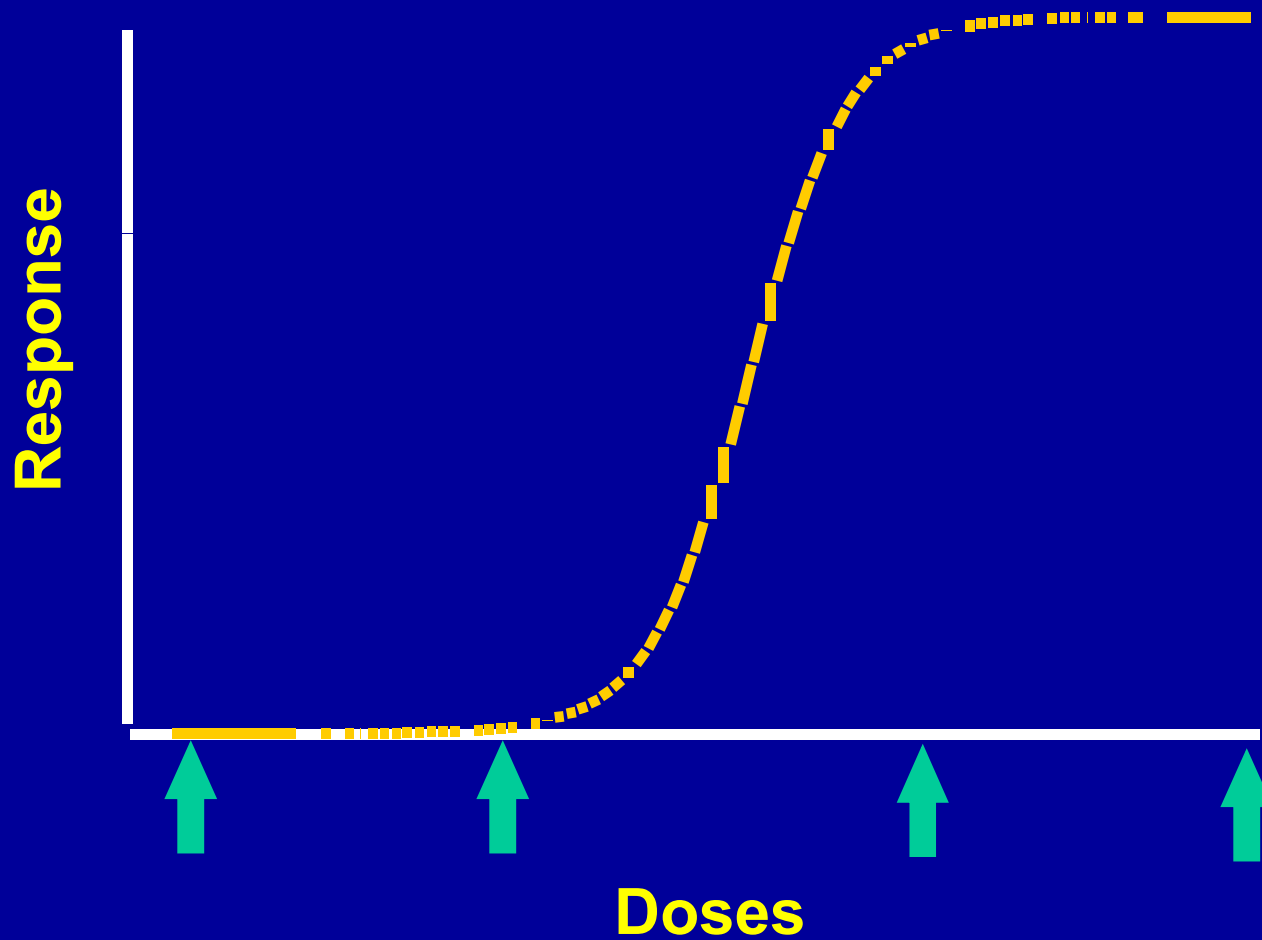
- **Adaptive doses in Phase II setting: learn efficiently and rapidly about dose-response relationship**
- **Pfizer trial of a neutrofil inhibitory factor**
- **Endpoint: stroke scale at week 13**
- **Early endpoints: weekly stroke scale**

Standard Parallel Group Design

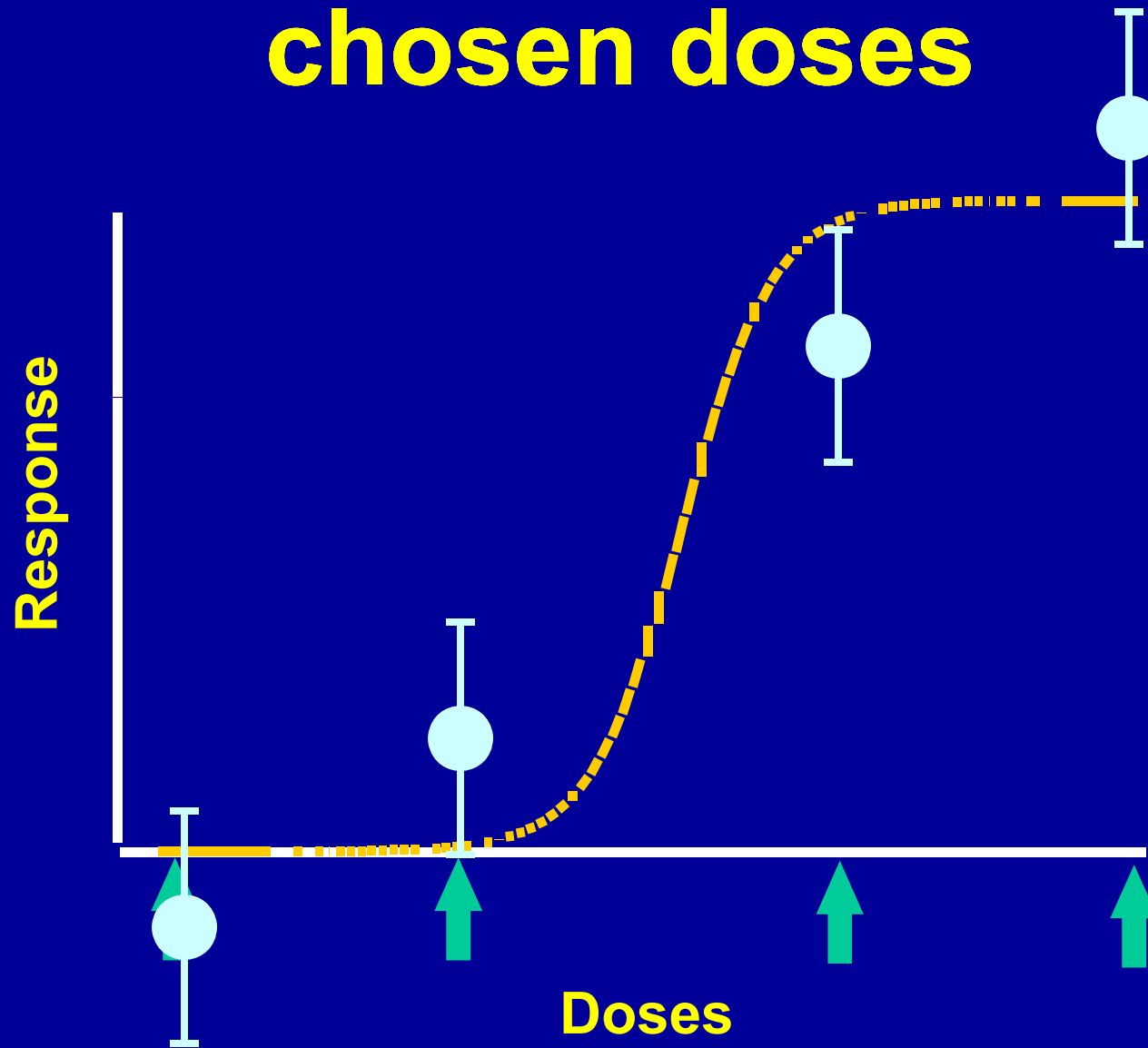
Equal sample sizes at each of k doses.



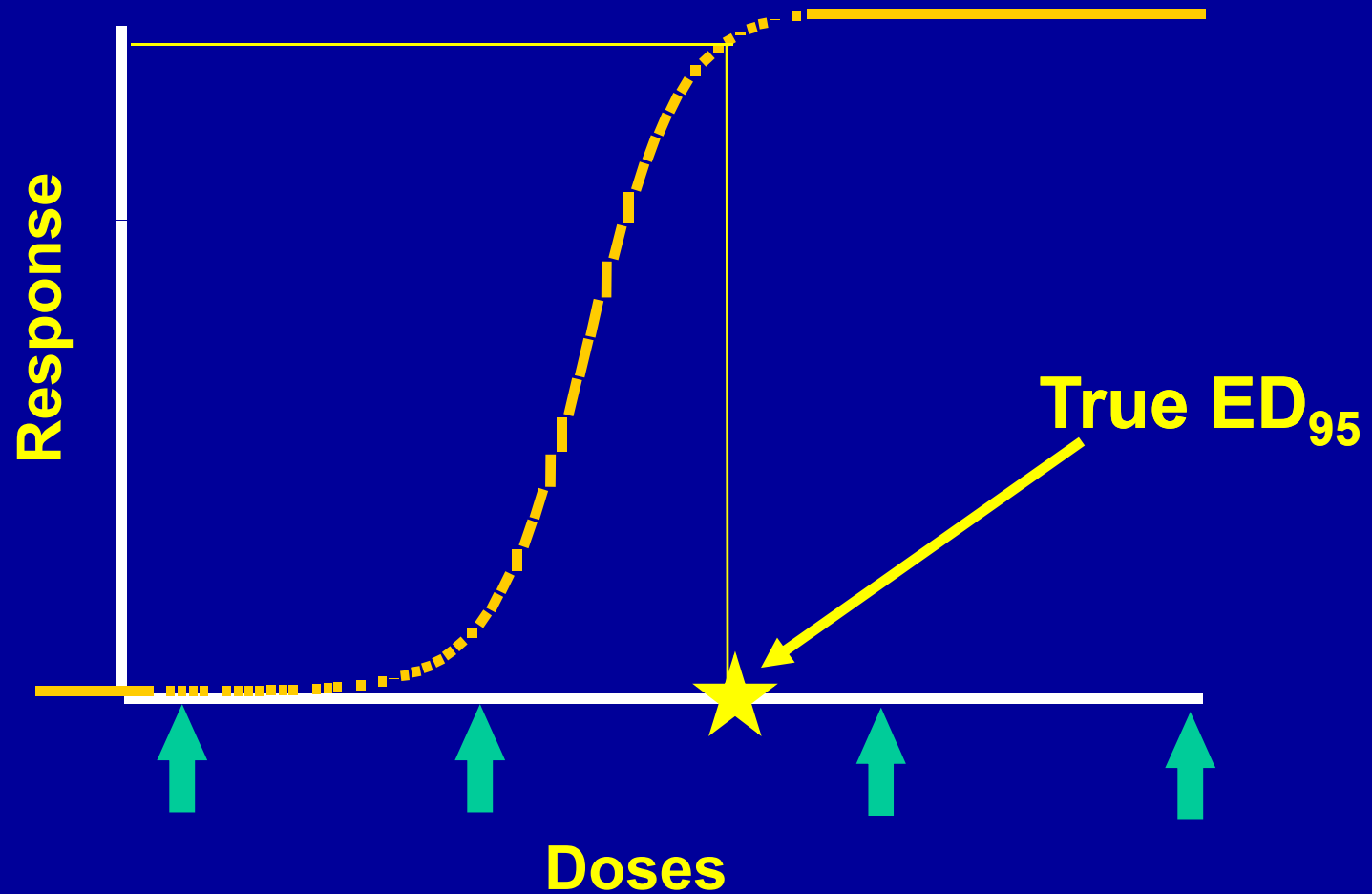
True dose-response curve (unknown)



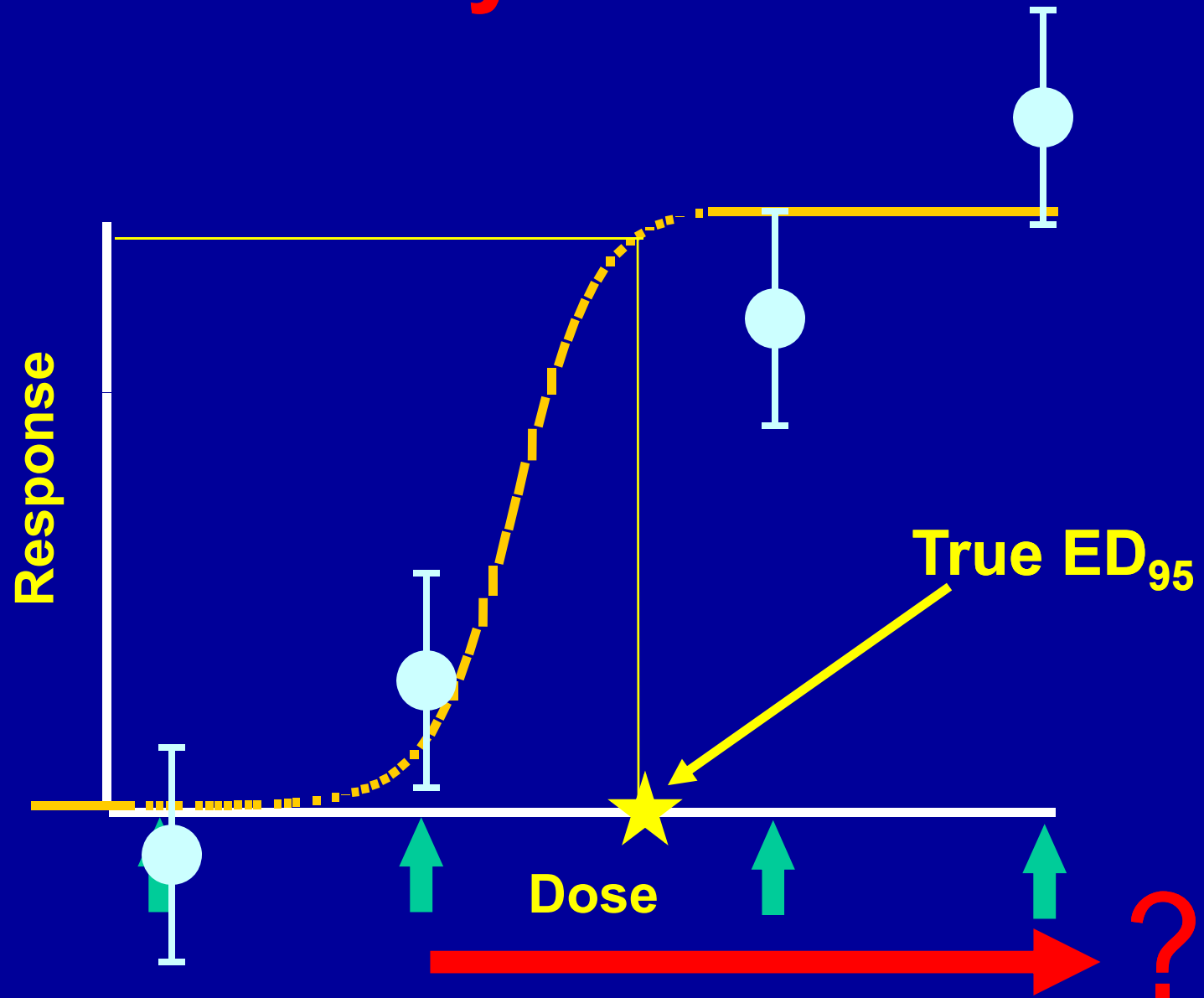
Observe responses (with error) at chosen doses



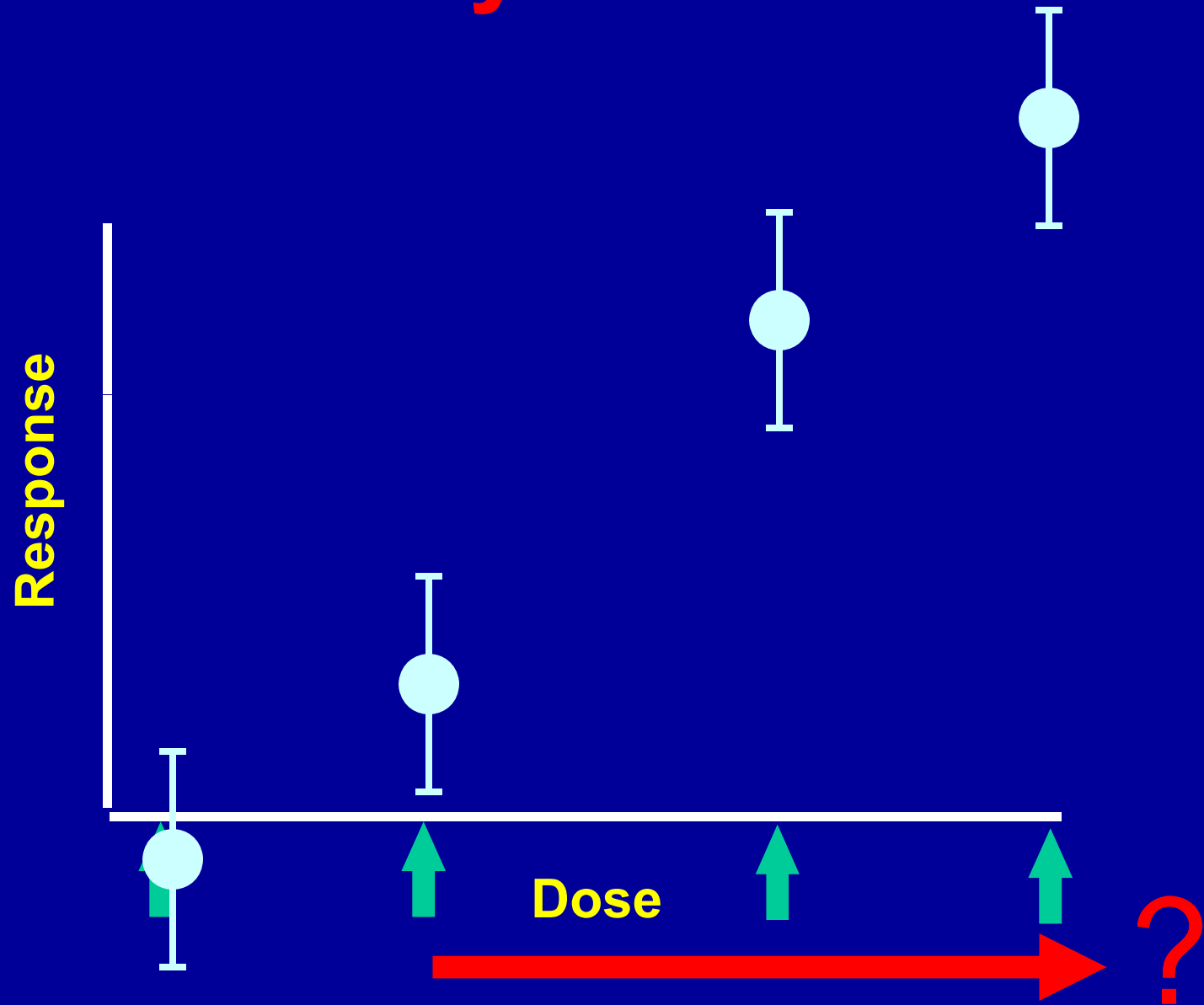
Dose at which 95% max effect



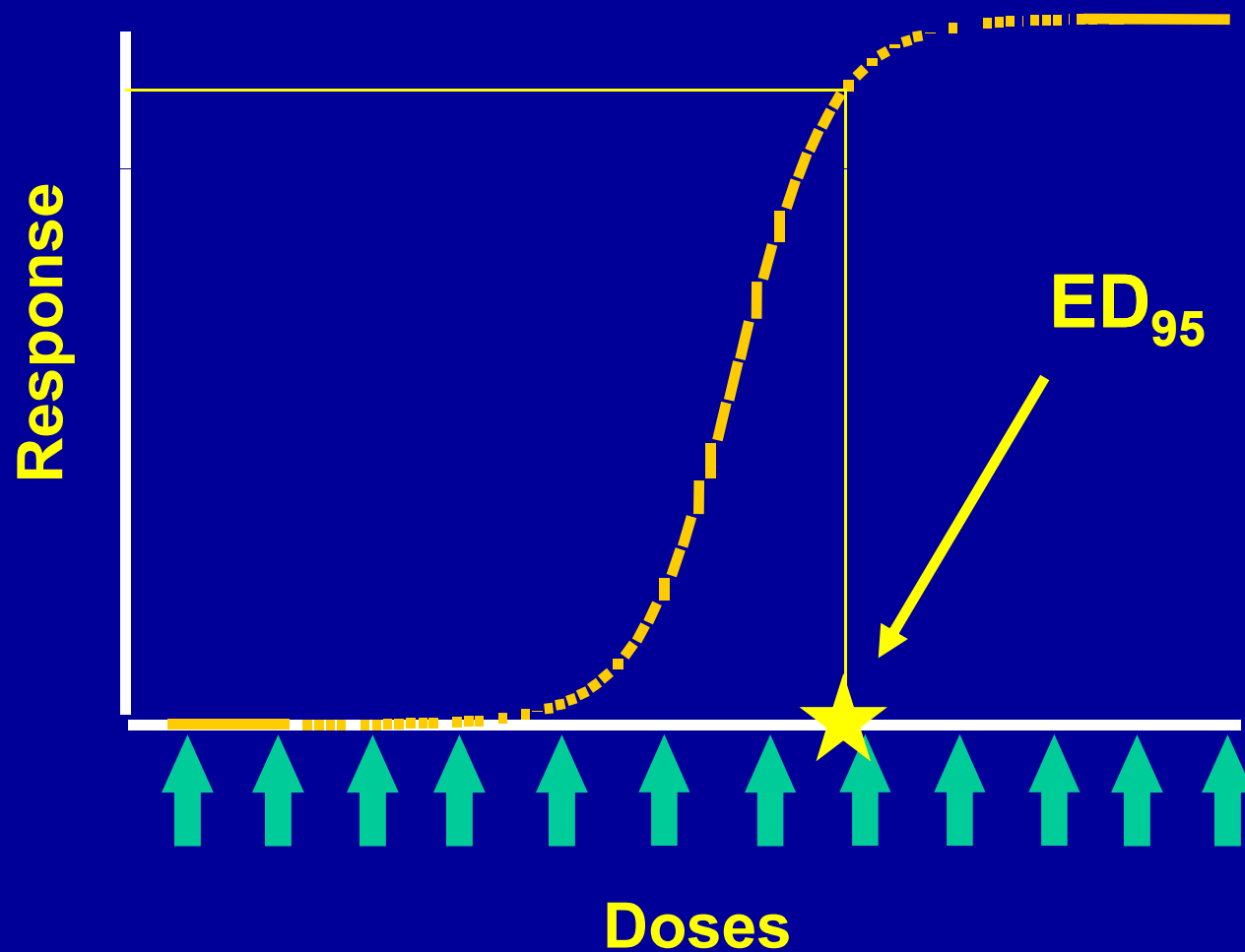
Uncertainty about ED95



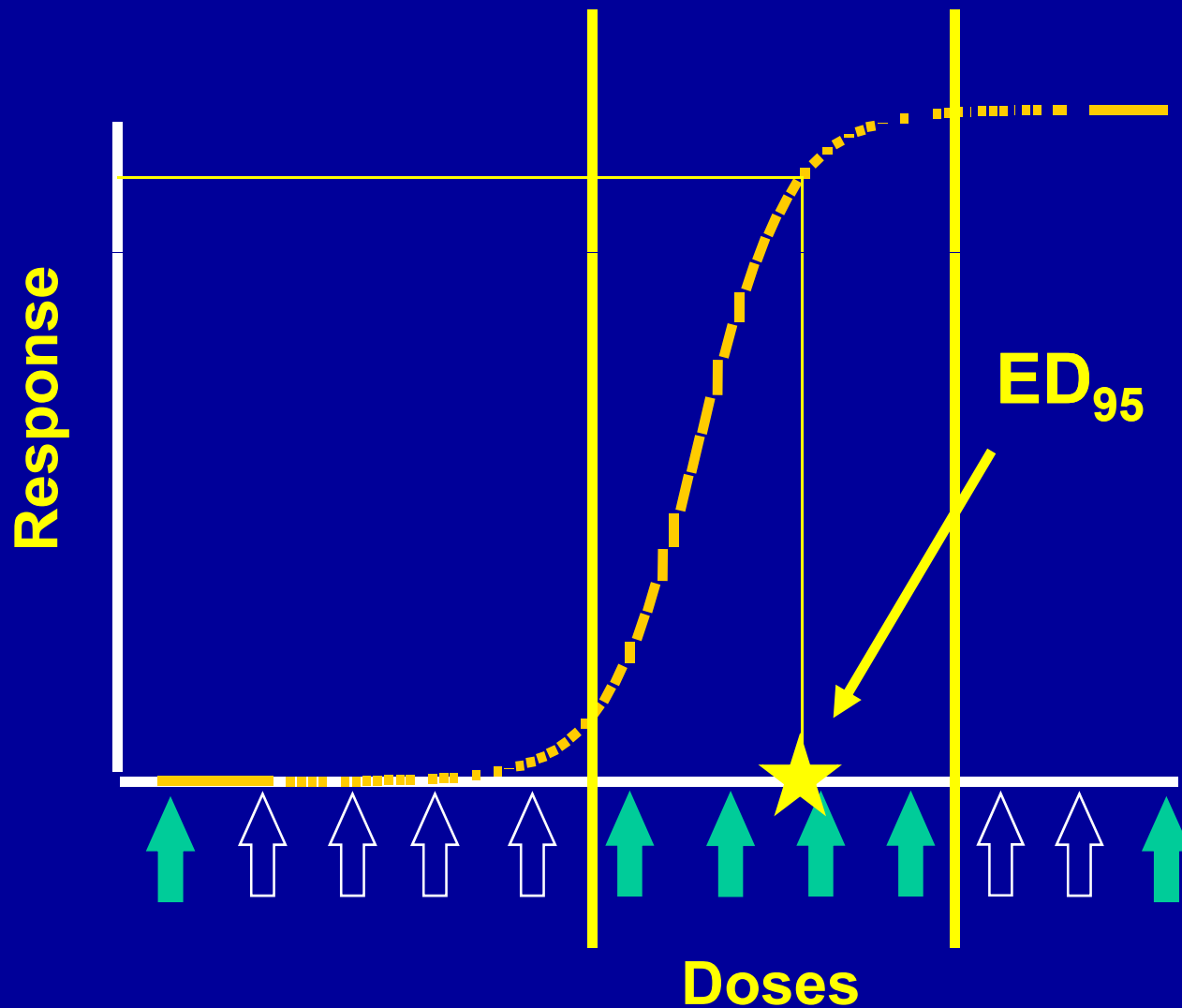
Uncertainty about ED95



Solution: Increase number of doses



***But, enormous sample size, and . . .
wasted dose assignments—always!***



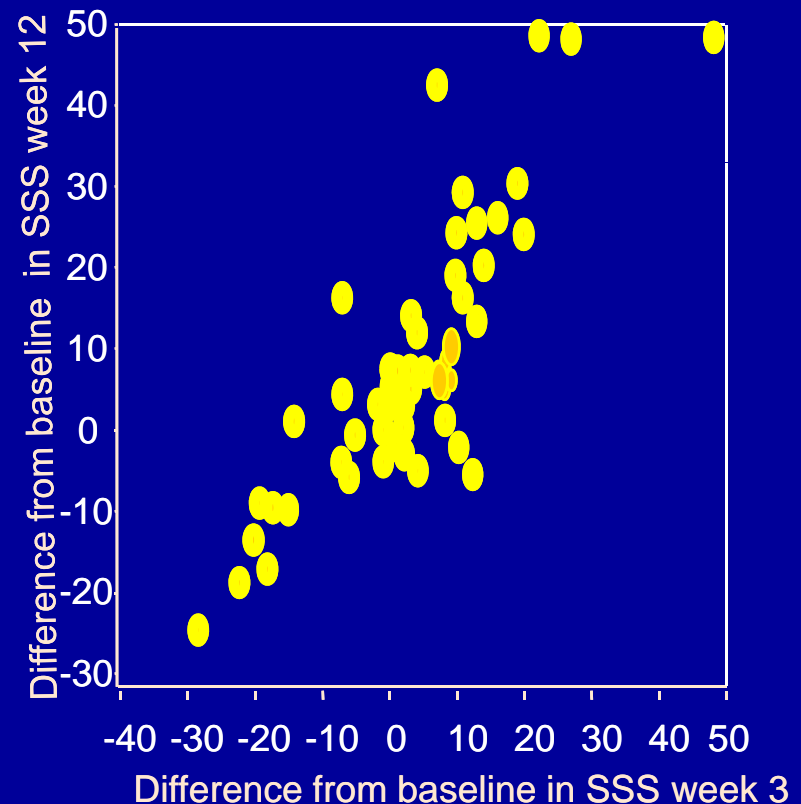
Our adaptive approach

- Observe data continuously
- Select next dose to maximize information about ED95, given available evidence
- Stop dose-ranging trial when know ED95 & response at ED95 “sufficiently well”

Our approach (cont'd)

Info accrues
gradually
about each
patient;
prediction
using
longitudinal
model

Longitudinal Model
Copenhagen Stroke Database



Our approach (cont'd)

- **Model dose-response
(borrow strength from
neighboring doses)**
- **Many doses (logistical issues)**

Possible decisions each day:

- Stop trial and drug's development
- Stop and set up confirmatory trial
- Continue dose-finding (what dose?)

Size of confirmatory trial based on
info from dose-ranging phase

Choices by decision analysis
(Human safeguard: DSMB)

Dose-response trial

- Learn efficiently and rapidly about dose-response; if + go to Phase III
- Assign dose to maximize info about dose-response parameters given current info
- Use predictive probabilities, based on early endpoints
- Doses in continuum, or preset grid

Dose-response trial (cont'd)

- Learn about SD on-line
- Halt dose-ranging when know dose sufficiently well
- Seamless switch from dose-ranging to confirmatory trial—
2 trials in 1!

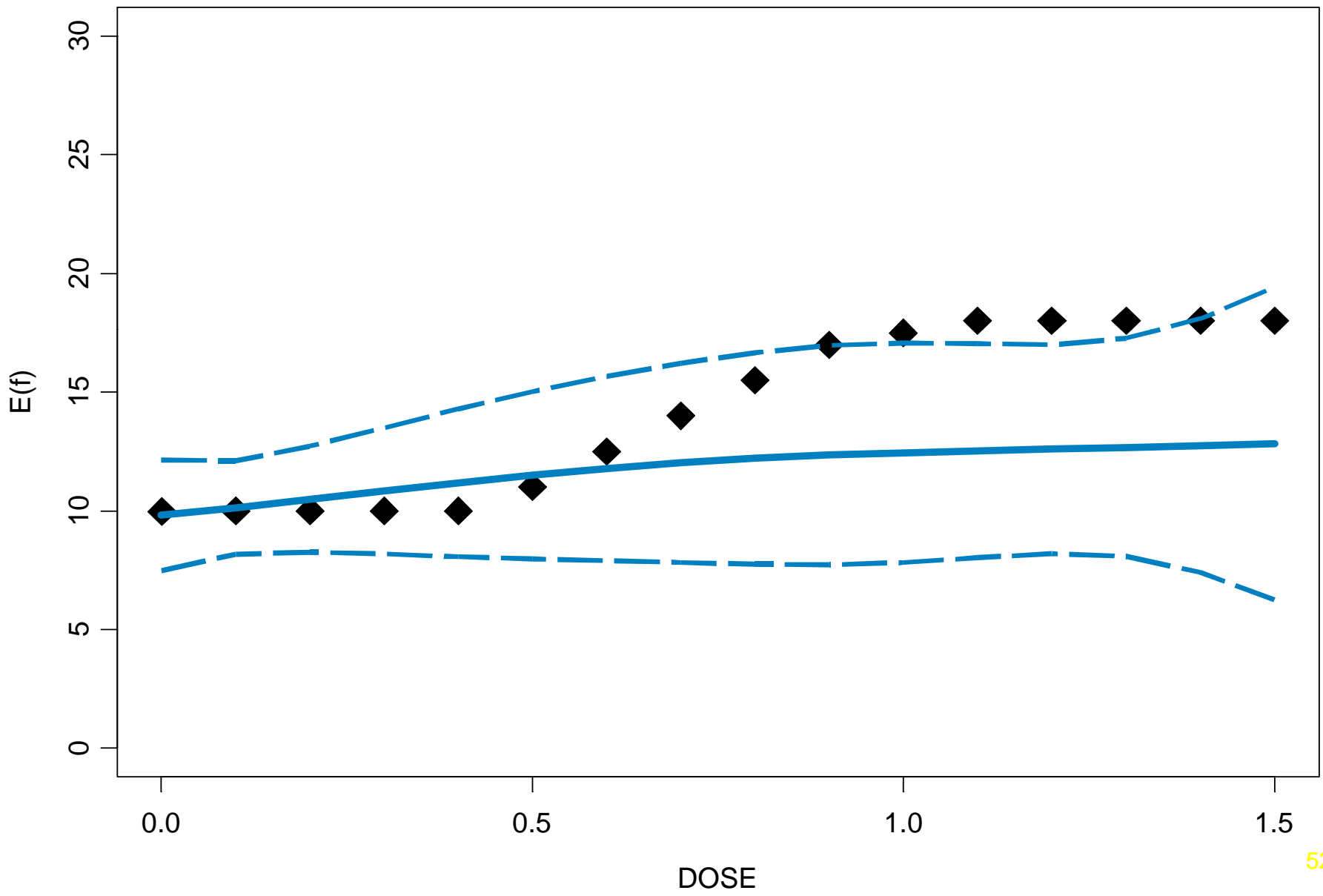
Advantages over standard design

- Fewer patients (generally); faster & more effective learning
- Better at finding ED95
- Tends to treat patients in trial more effectively
- Drops duds early—actual trial!

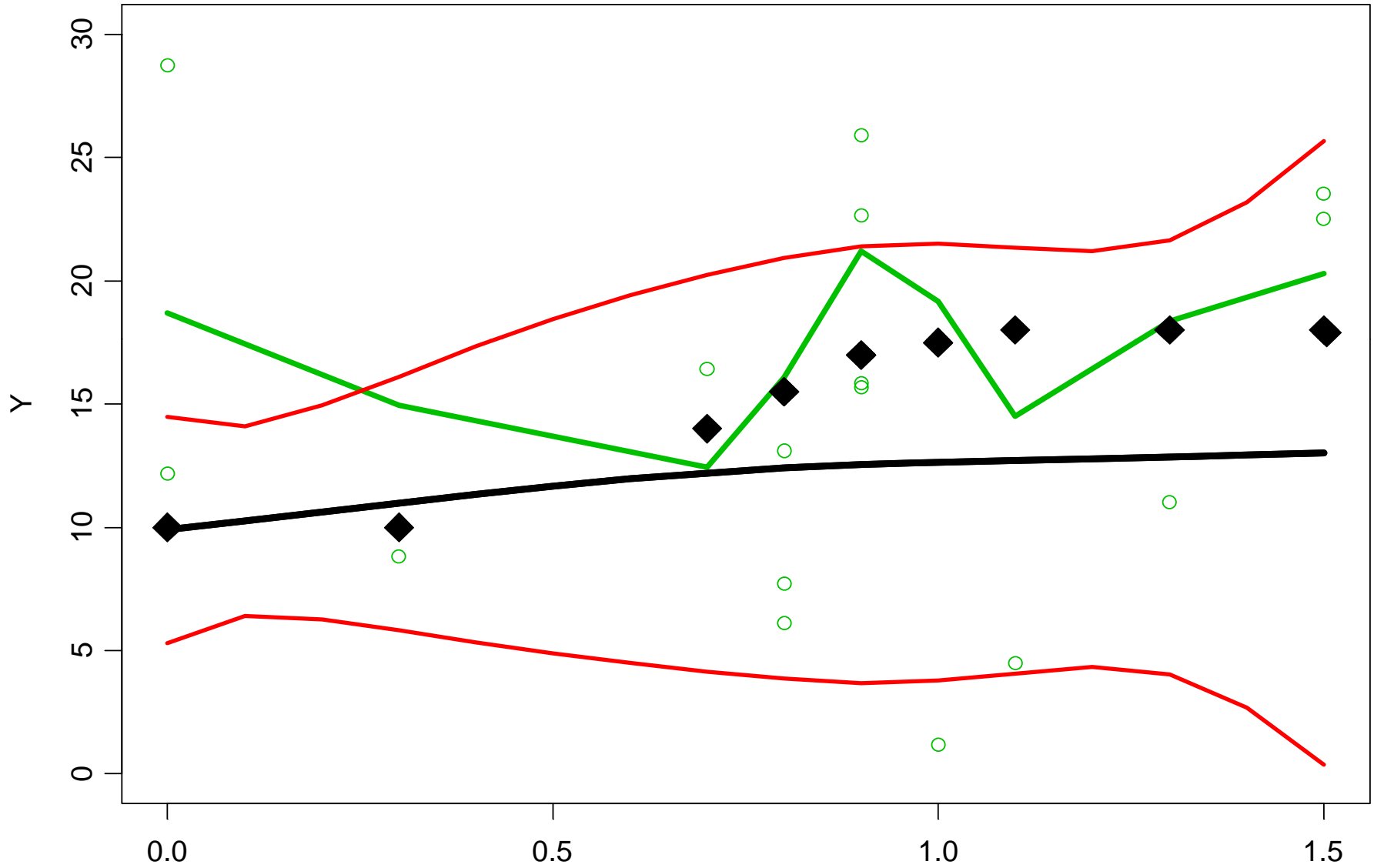
Dose-assignment simulation

- Assumes particular dose-response curve
- Assumes $SD = 12$
- Shows weekly results, several patients at a time (green circles)

Prior

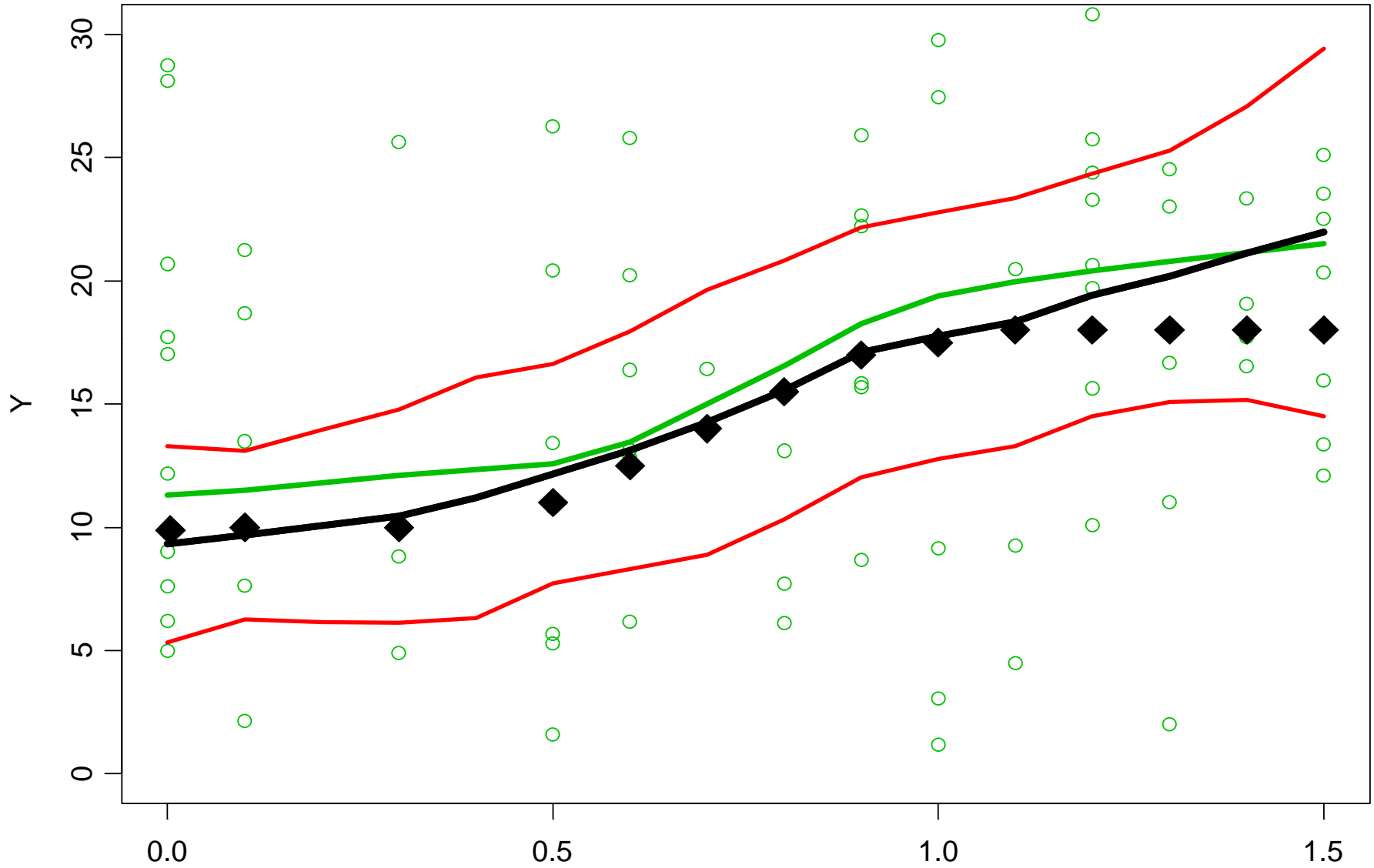


DATA



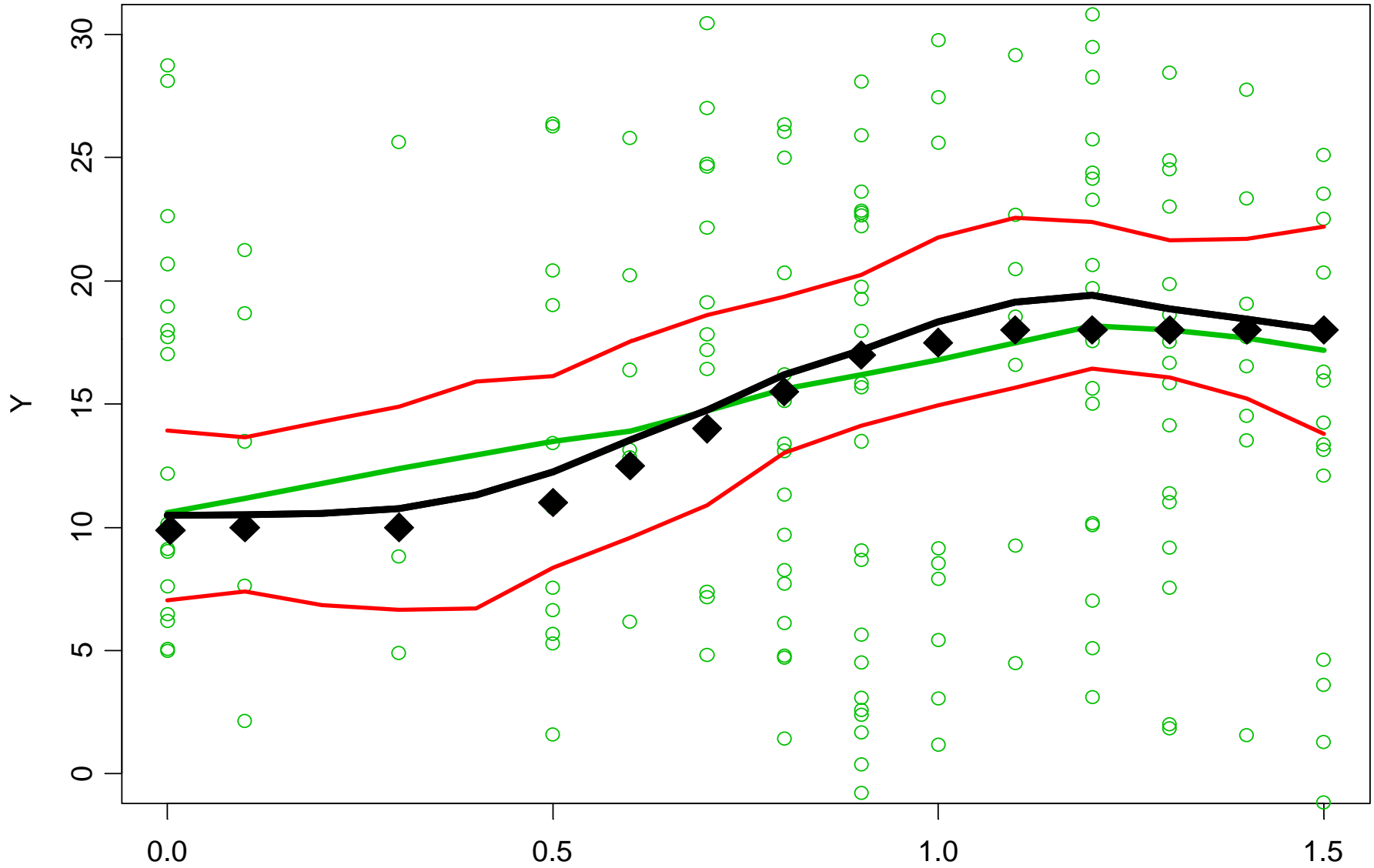
green=obs, blue=imputed, black=true mn

DATA



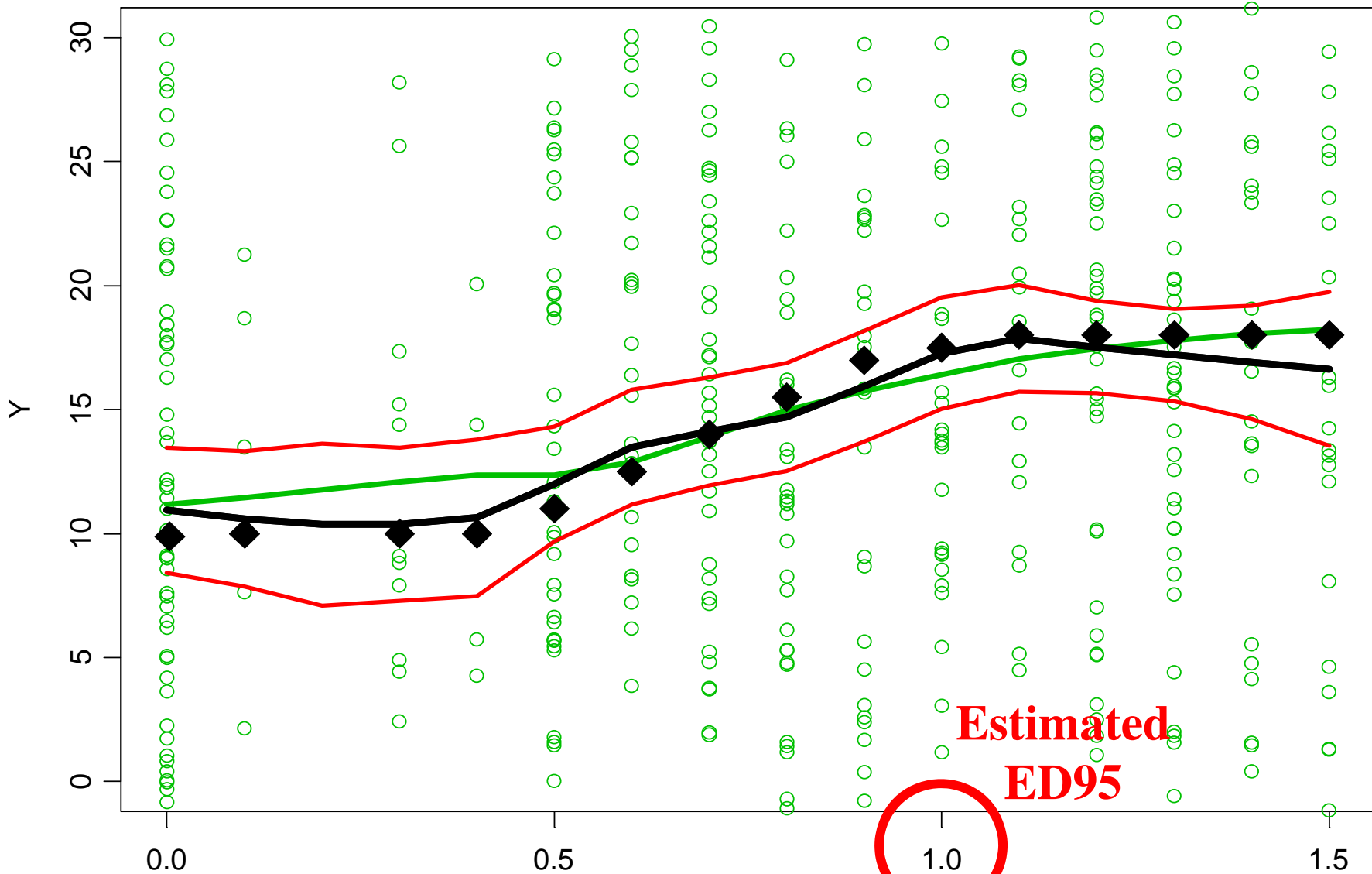
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DATA



green=obs, blue=imputed, black=true mn

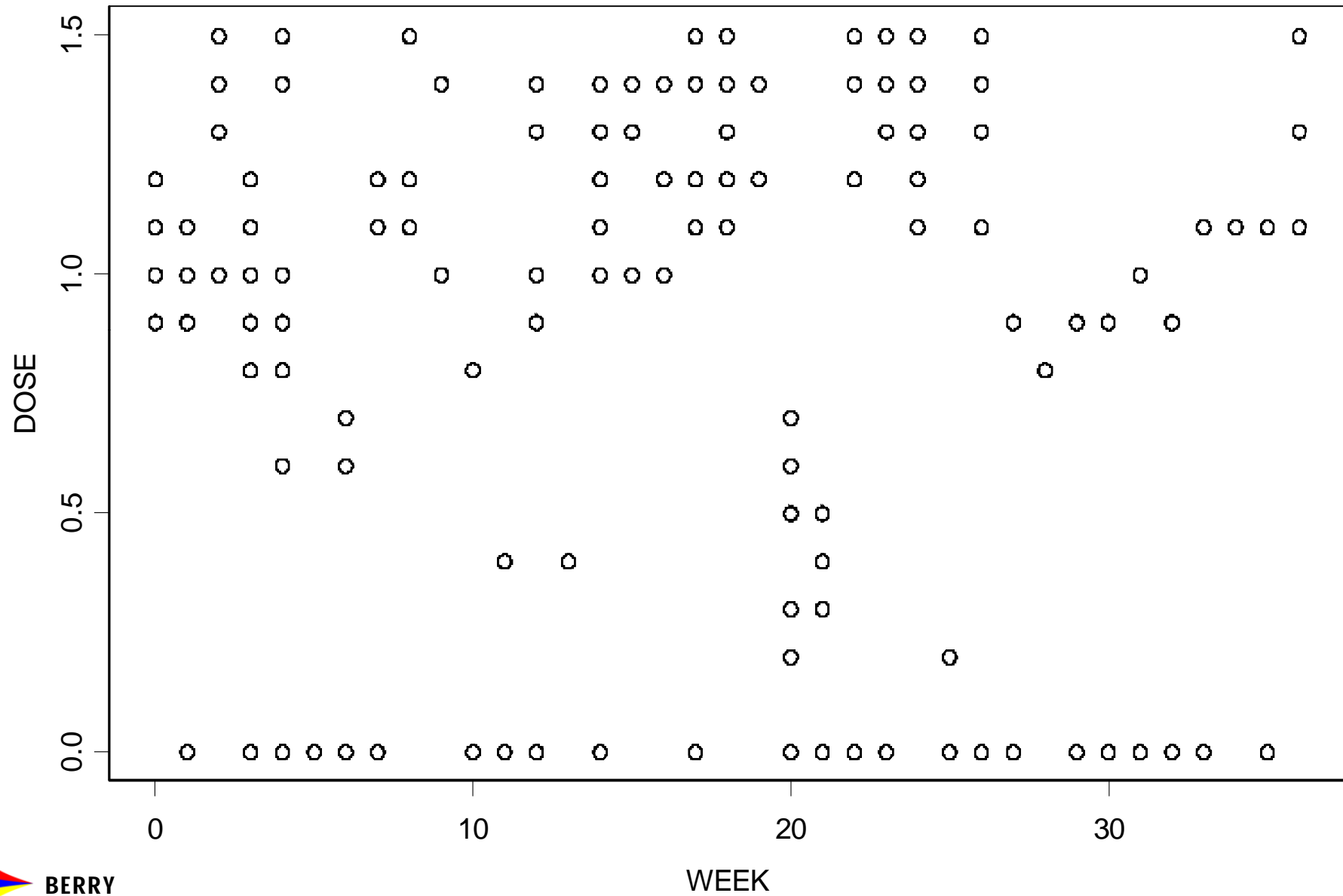
DATA



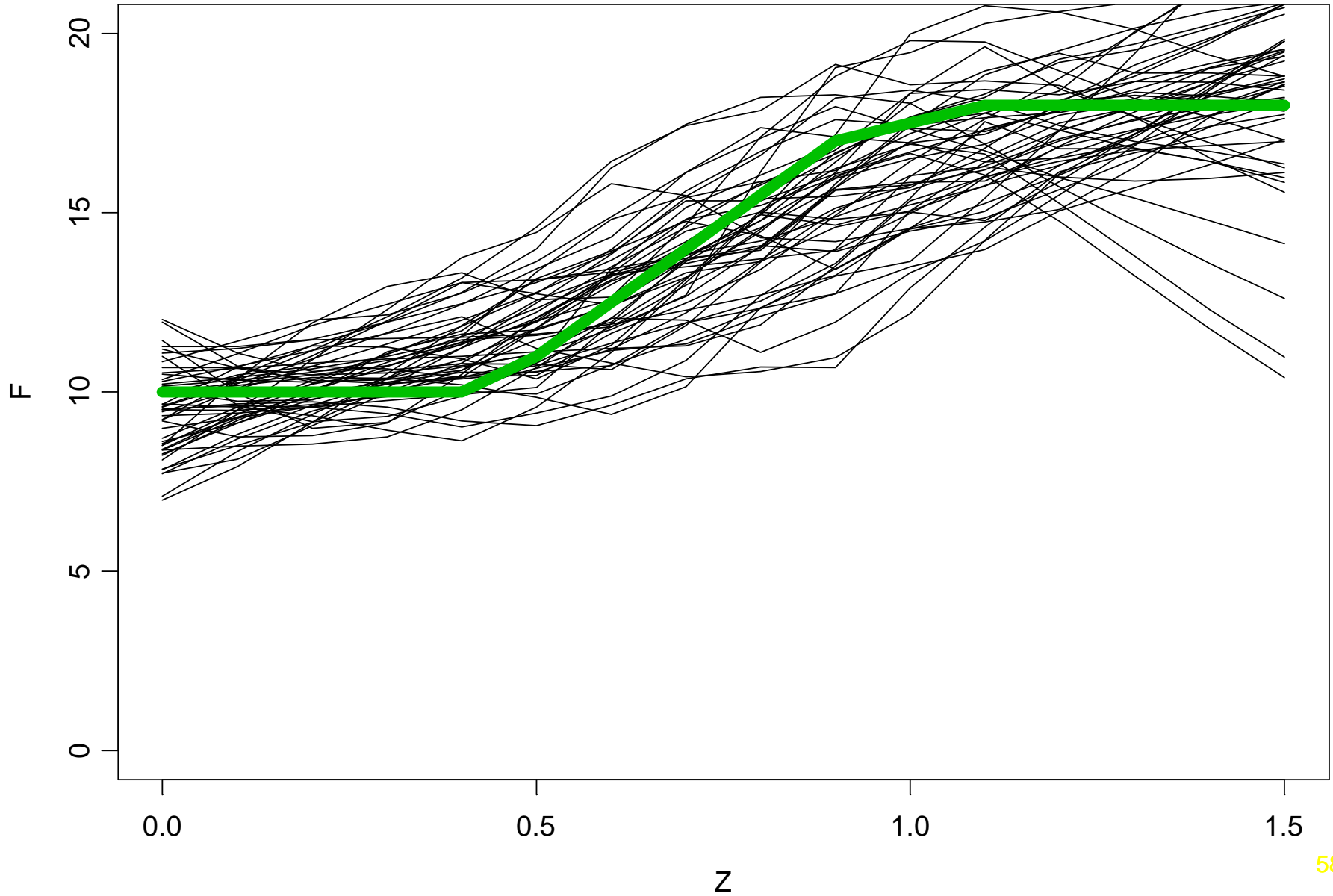
green=obs, blue=imputed, black=true mn

→ Confirmatory

Assigned Doses by Week - one simulation

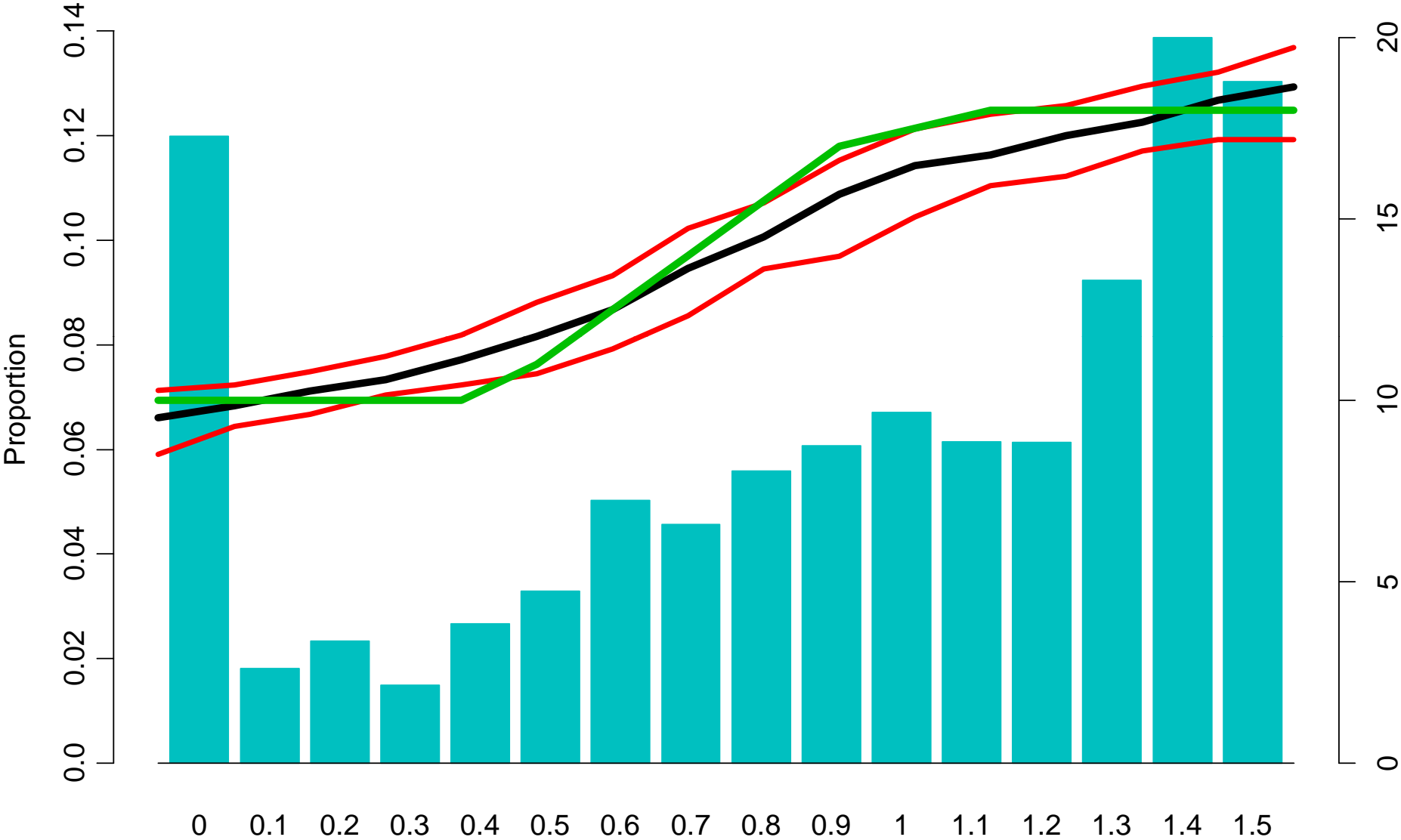


Estimated functions



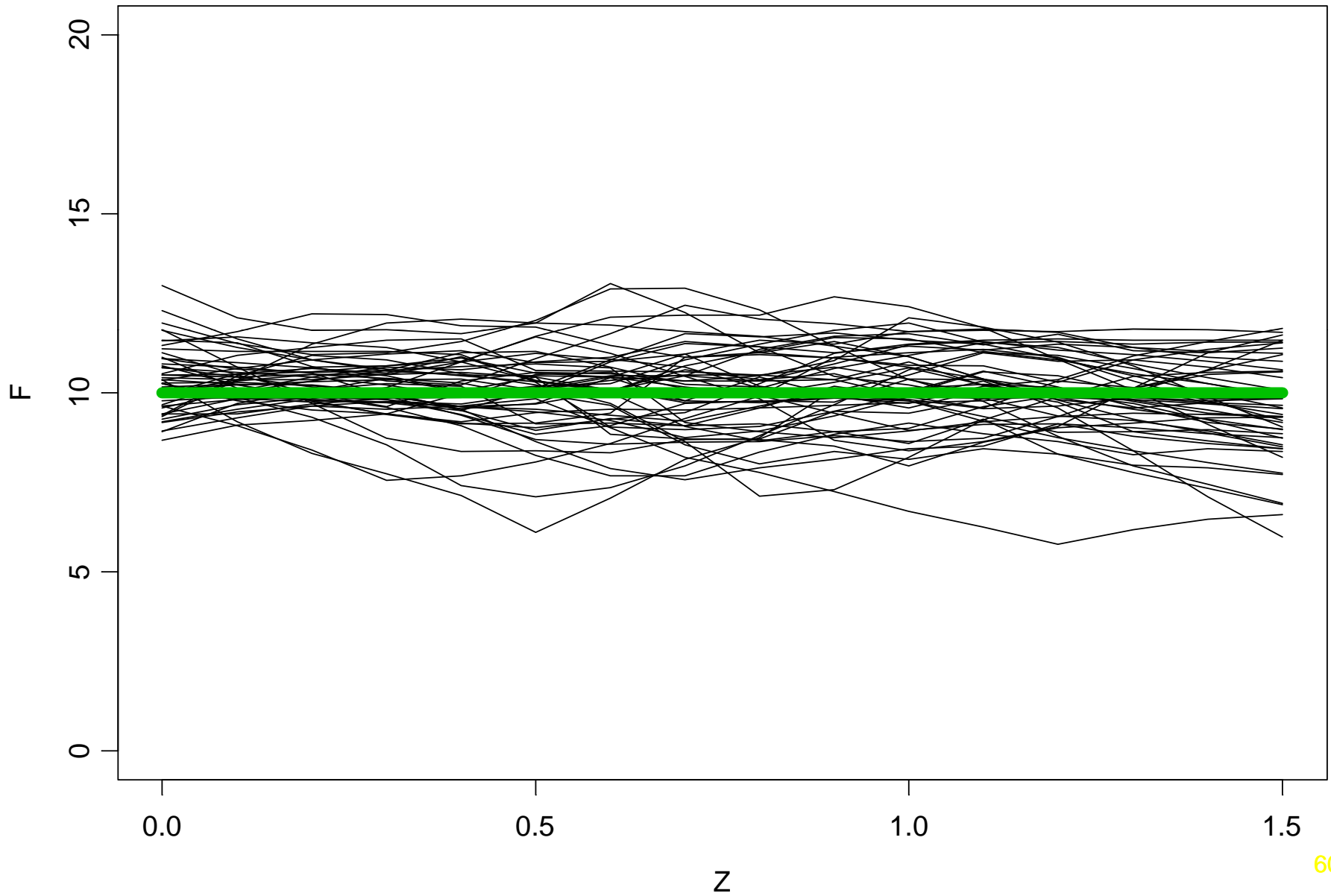
d:/data/build13/run11/

Doses assigned across all simulations



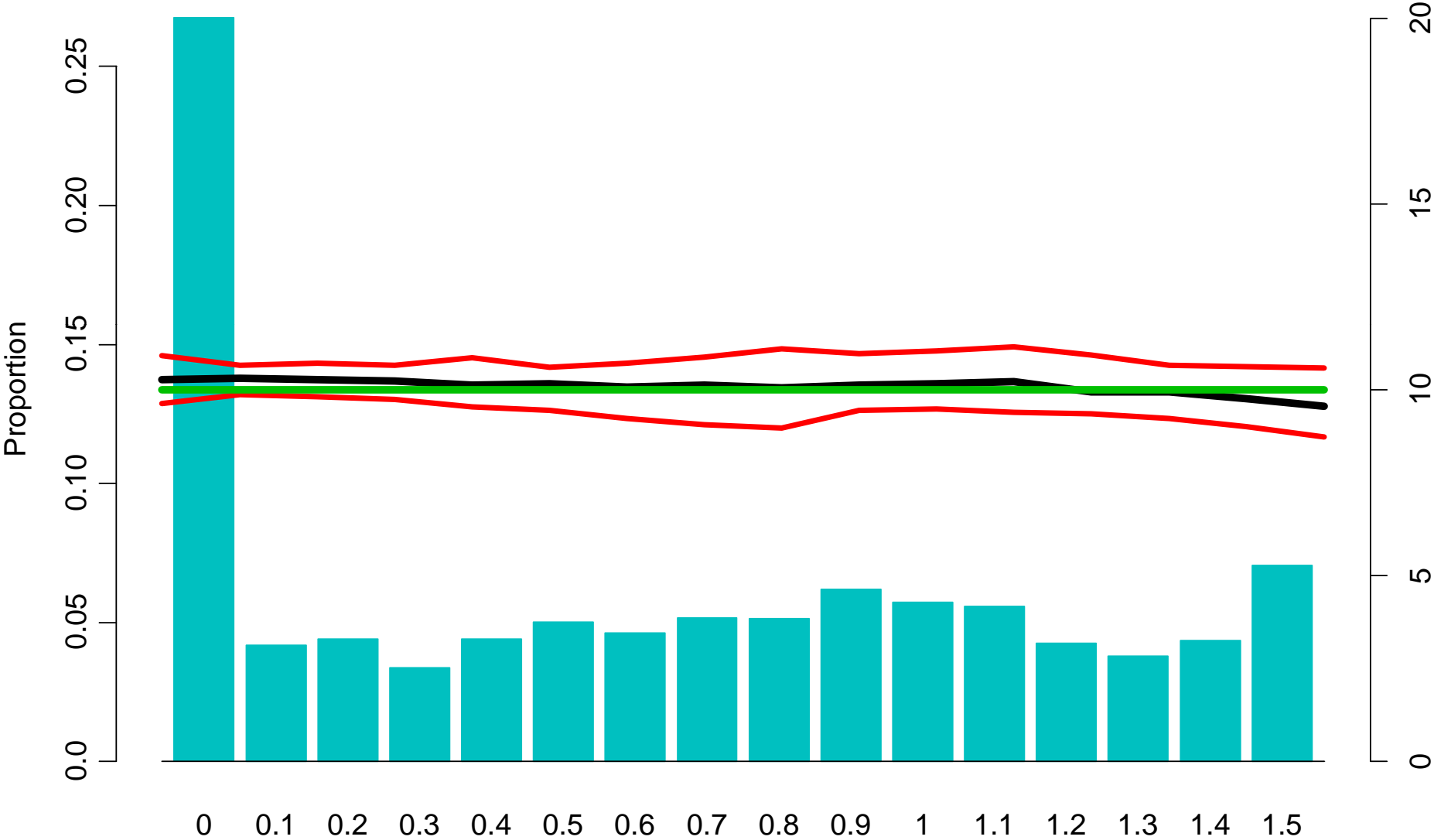
ASSIGNED DOSES
Black: median; Red: upper & lower quartiles; Green: Nominal

Estimated functions (no dose effect)



d:/data/build13/run12/

Doses assigned across all simulations



ASSIGNED DOSES
Black: median; Red: upper & lower quartiles; Green: Nominal

Deploying the design: ASTIN stroke study

ASTIN Study

- Does UK-279,276 improve recovery in ischemic stroke? Dose effect?
- Placebo and 15 doses
- Primary efficacy endpoint:
Scandinavian Stroke Scale, Mean change from baseline to day 90
- Patients with
 - Age \geq 50 years
 - Acute ischemic stroke $<$ 6 hours

ASTIN Study

- **Executive Steering Committee**
- **Independent Data Monitoring Committee**
- **Independent statistician preparing reports for IDMC**
- **Computer system run by Tessella Ltd (UK)**

NEW SUBJECT DETAILS

61561 **Pfizer** UK-279,276
Subject Initials and Date of Birth **PSL1011940** 076

Investigator **DR SANDERS** Screening Number **S10099003**

Using a black pen, carefully fill out the information below. Completely black out the appropriate ellipses, and write in black ink above the ellipses. If you make an error, please use a new form.

S-centre# -9-Log# of randomized patients in your centre

Date of Visit
Day: 01, Month: JAN, Year: 2000
Screening Number: S10099003

Date of Visit	Screening Number
0	0
1	1
2	2
3	3
4	4
5	5
6	6
7	7
8	8
9	9

Screening number

Scandinavian Stroke Scale

SSS Rating
Range 00-68

27

0	0
1	1
2	2
3	3
4	4
5	5
6	6
7	7
8	8
9	9

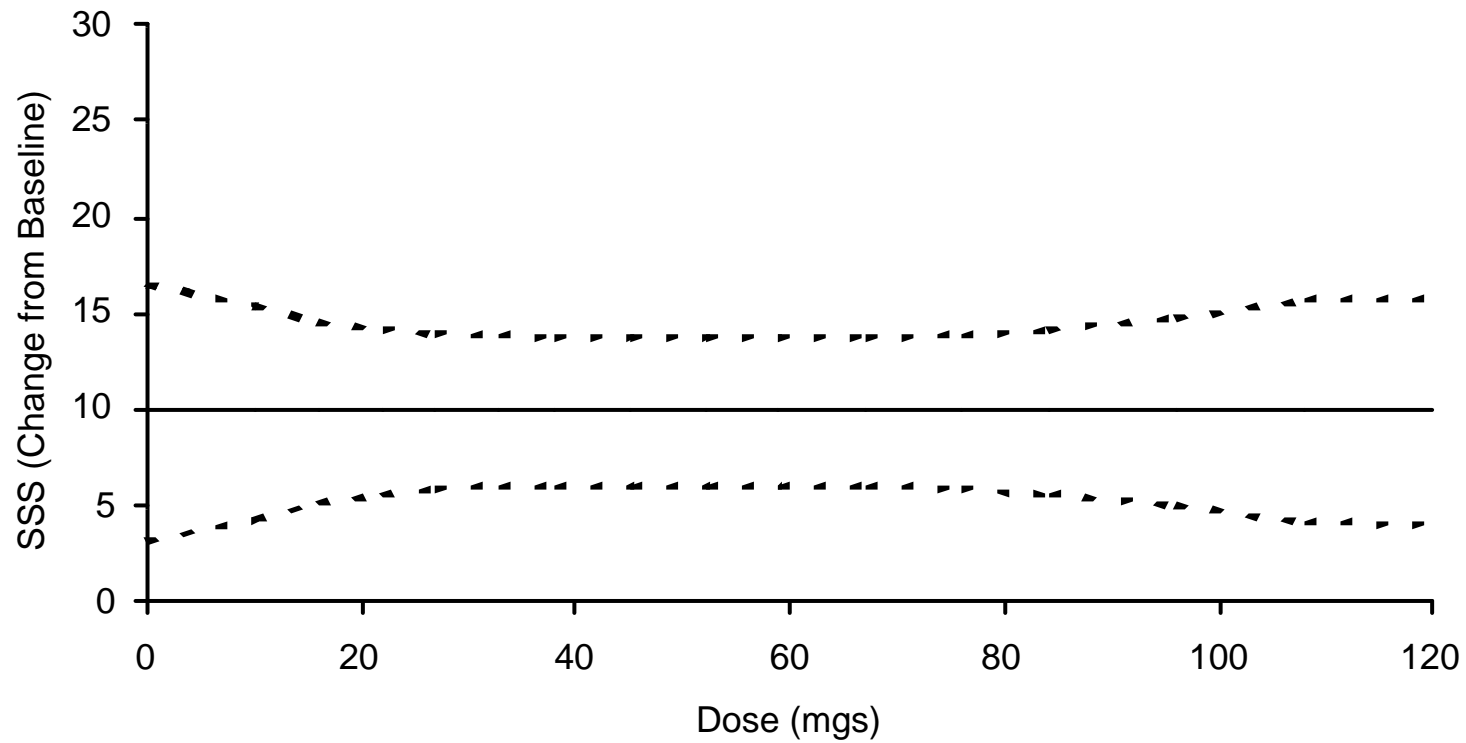
Signature

This form must be signed to be accepted. After you have completed the form, fax it from outside the UK to 44 1235 538 425 or within the UK to 01235 538 425 and wait for a confirmation fax. If you do not receive confirmation within 5 minutes, call outside the UK 44 790 151 0262 or within the UK 0790 151 0262 to report the problem.

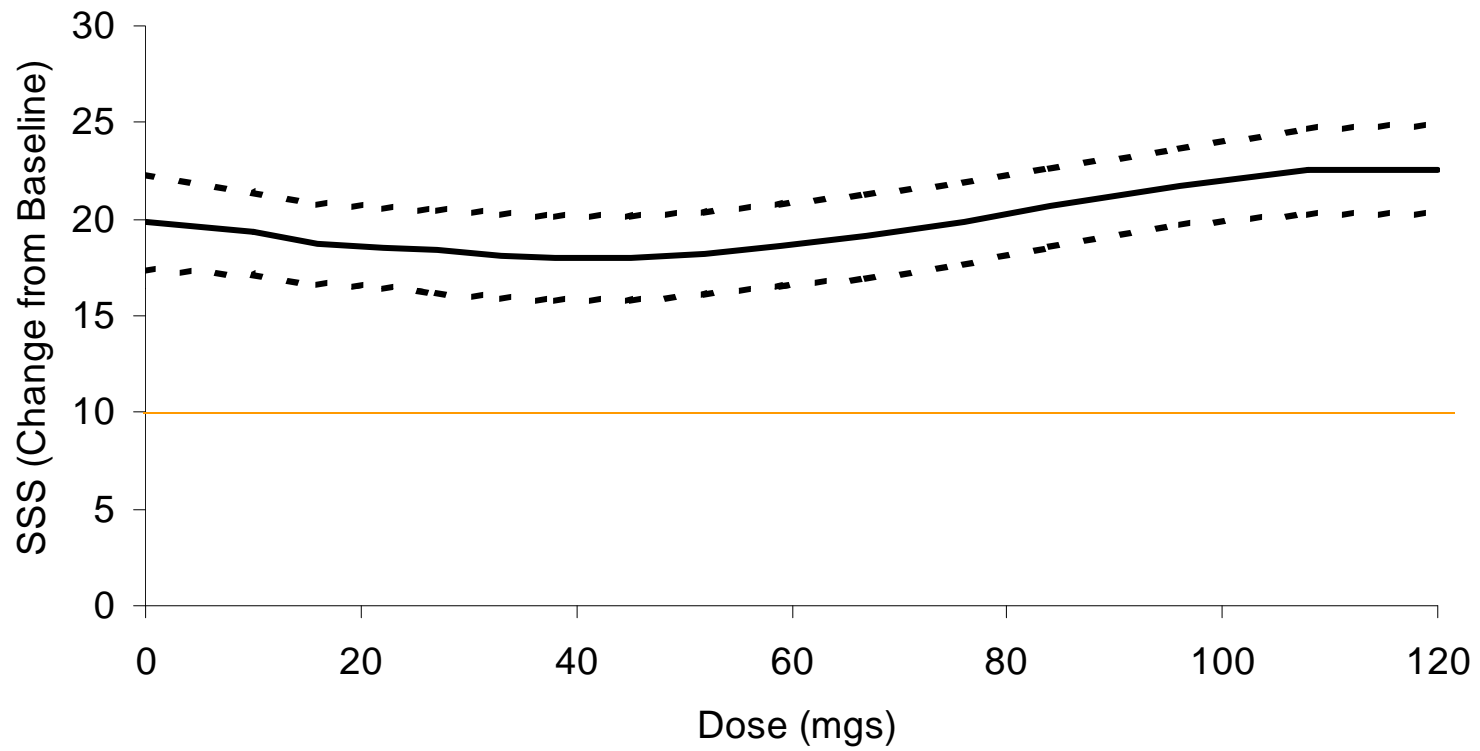
[Signature]
SIGNATURE

ASTIN Results

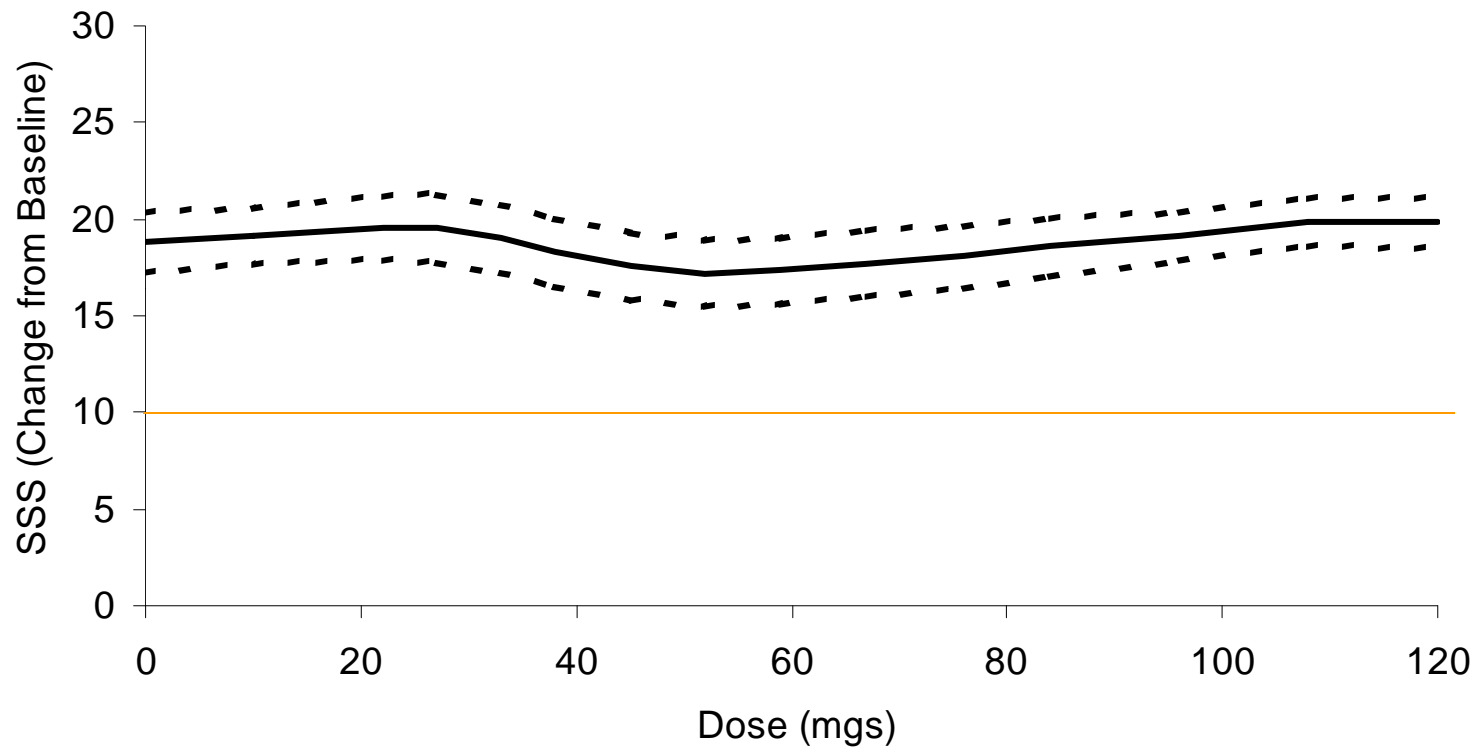
- **Trial stopped for futility; 500 with week 13 SSS**
- **966 patients randomized/treated**
- **93% ischemic stroke**
 - **21% cotreated with tPA**
 - **Mean baseline severity ~ 28 points Scand Stroke Scale**
 - **Demographics comparable across treatment arms**



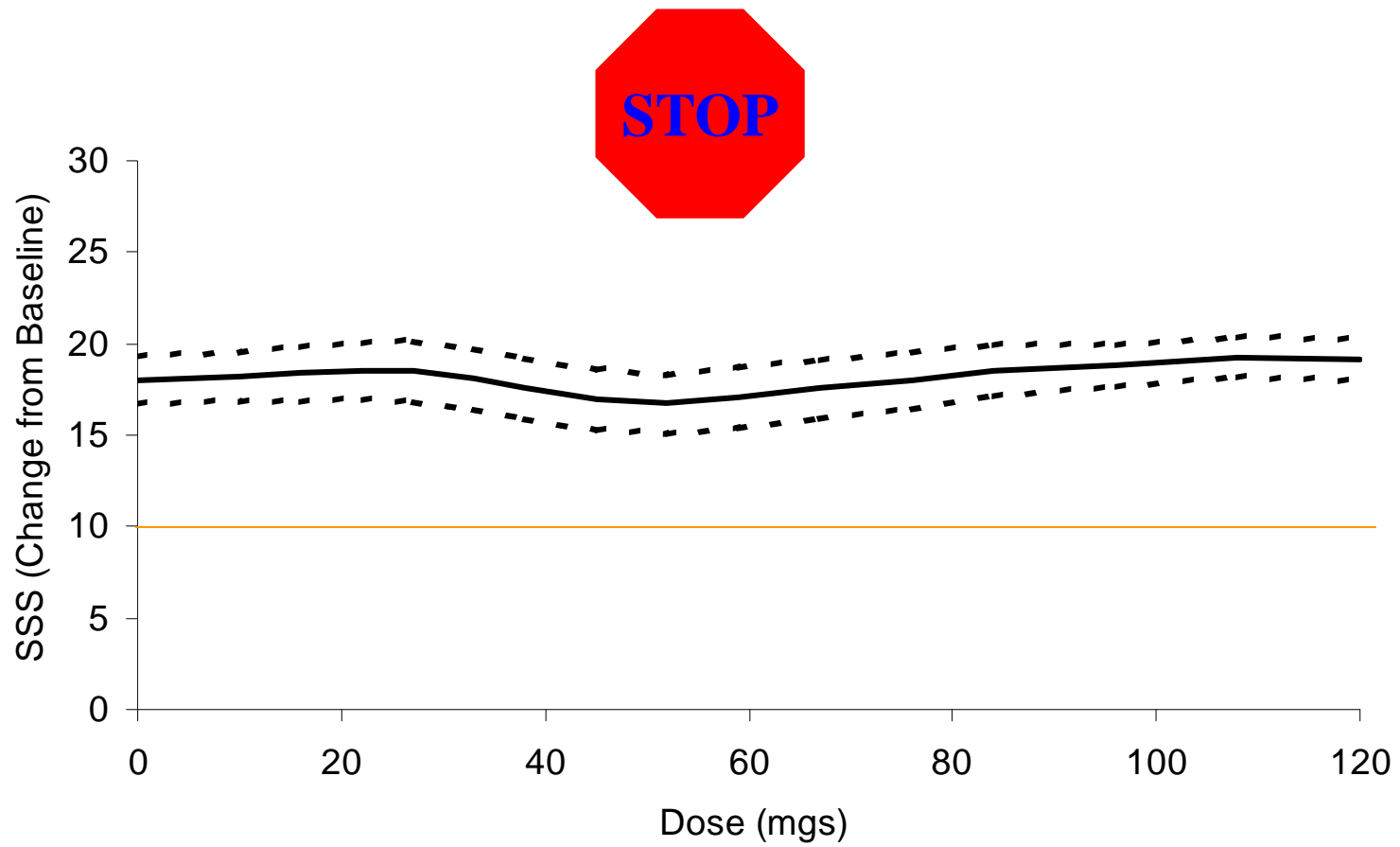
Week - 0



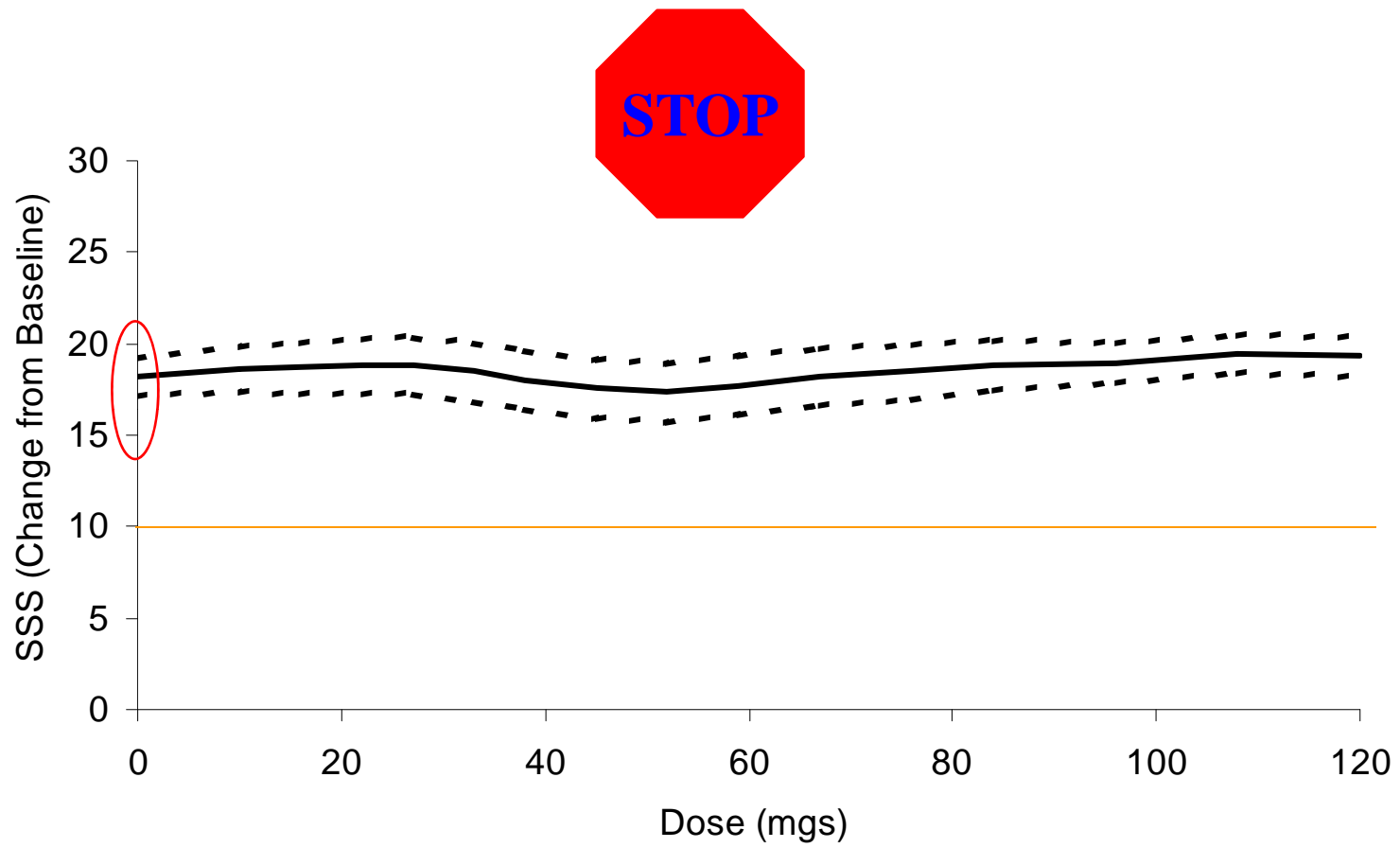
Week - 20



Week - 40

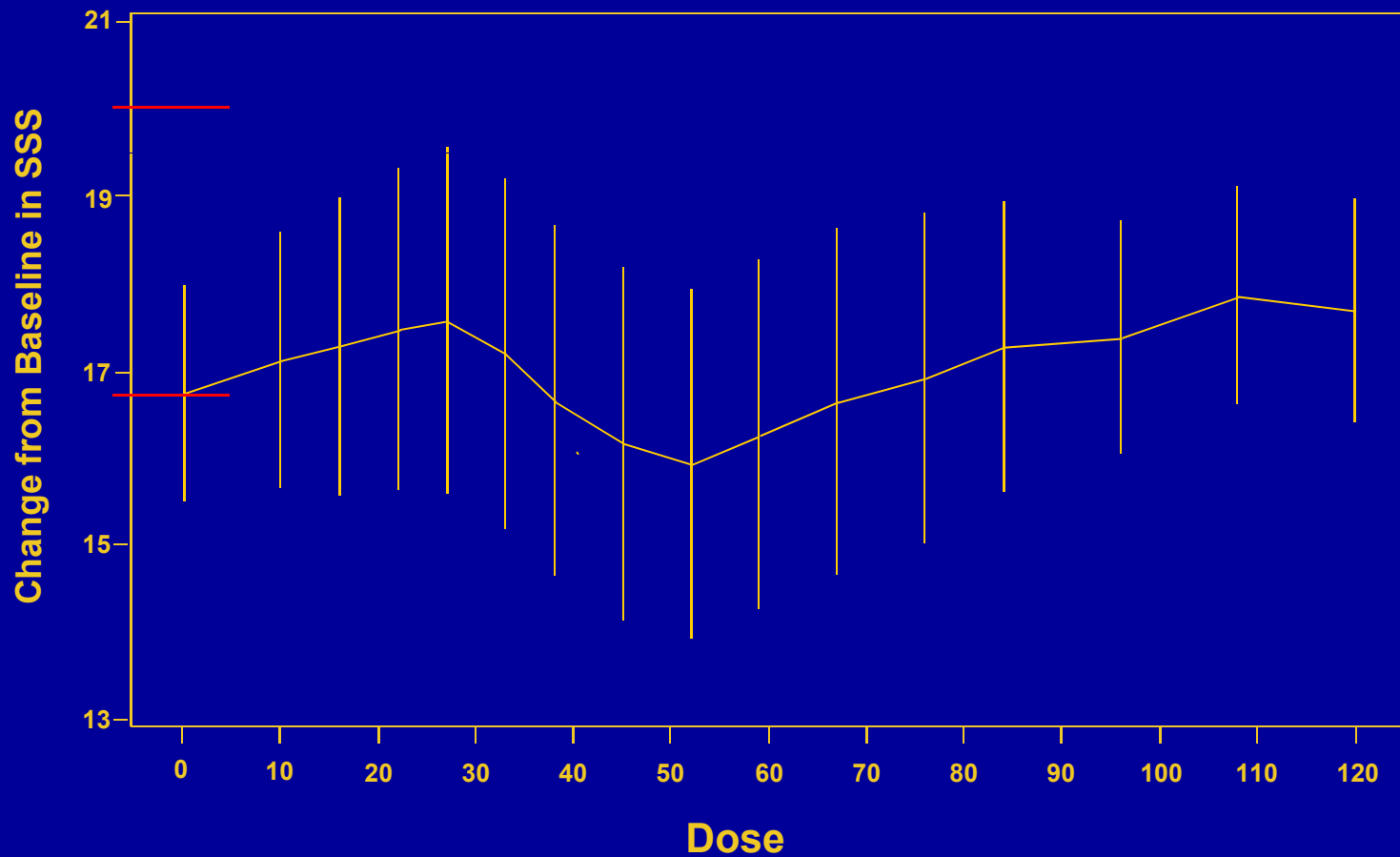


Week - 48

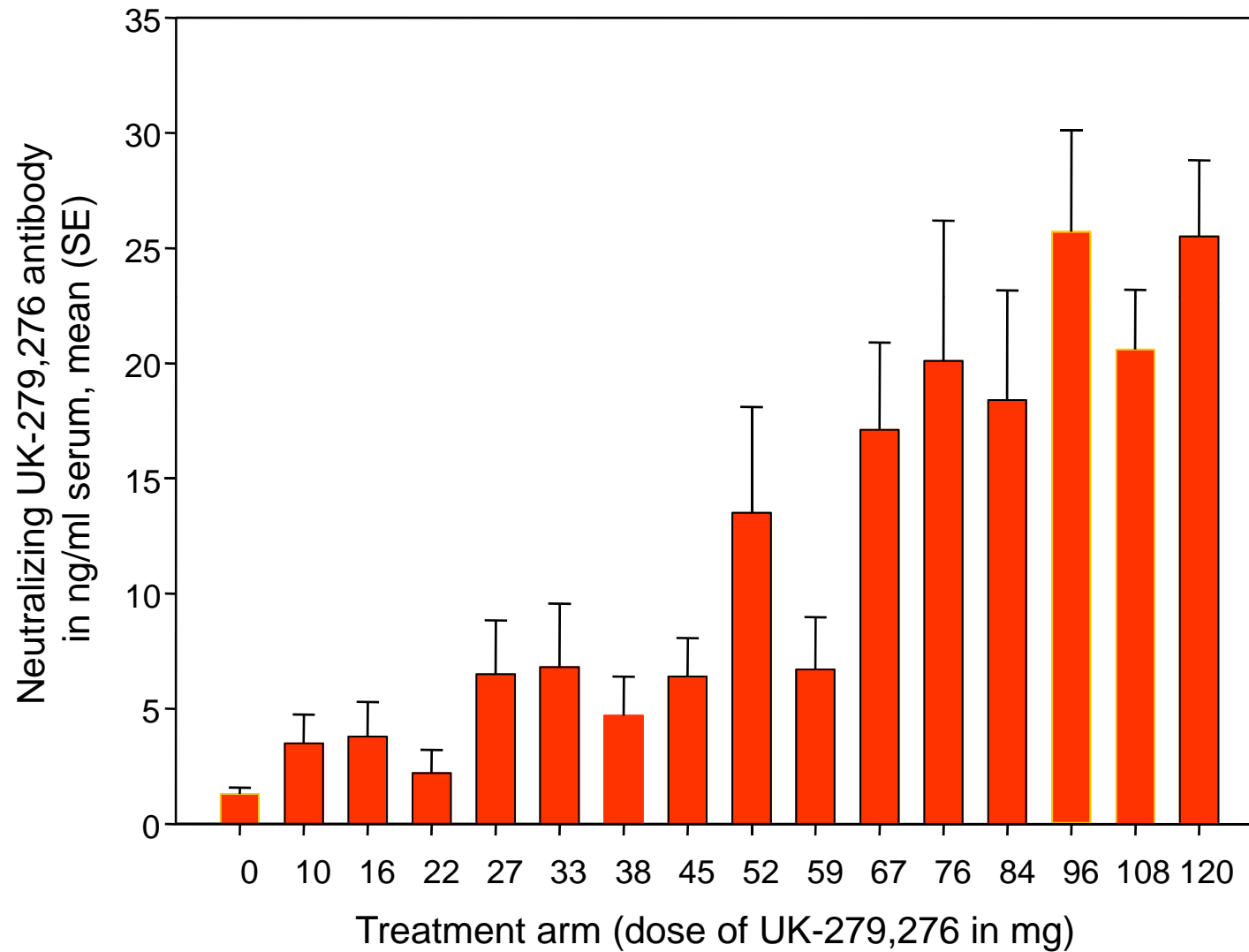


Week - 66

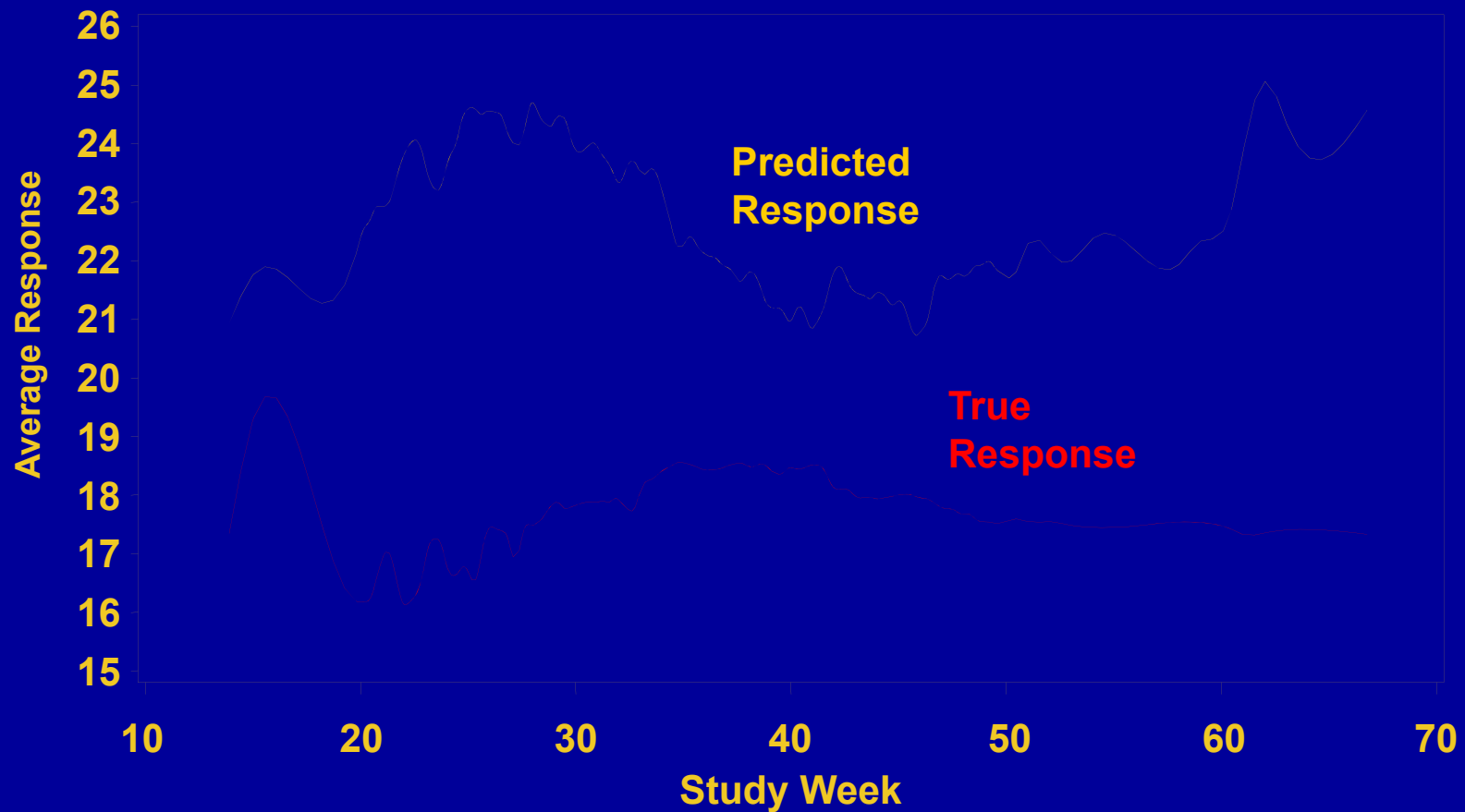
SSS Response Curve – Primary Analysis



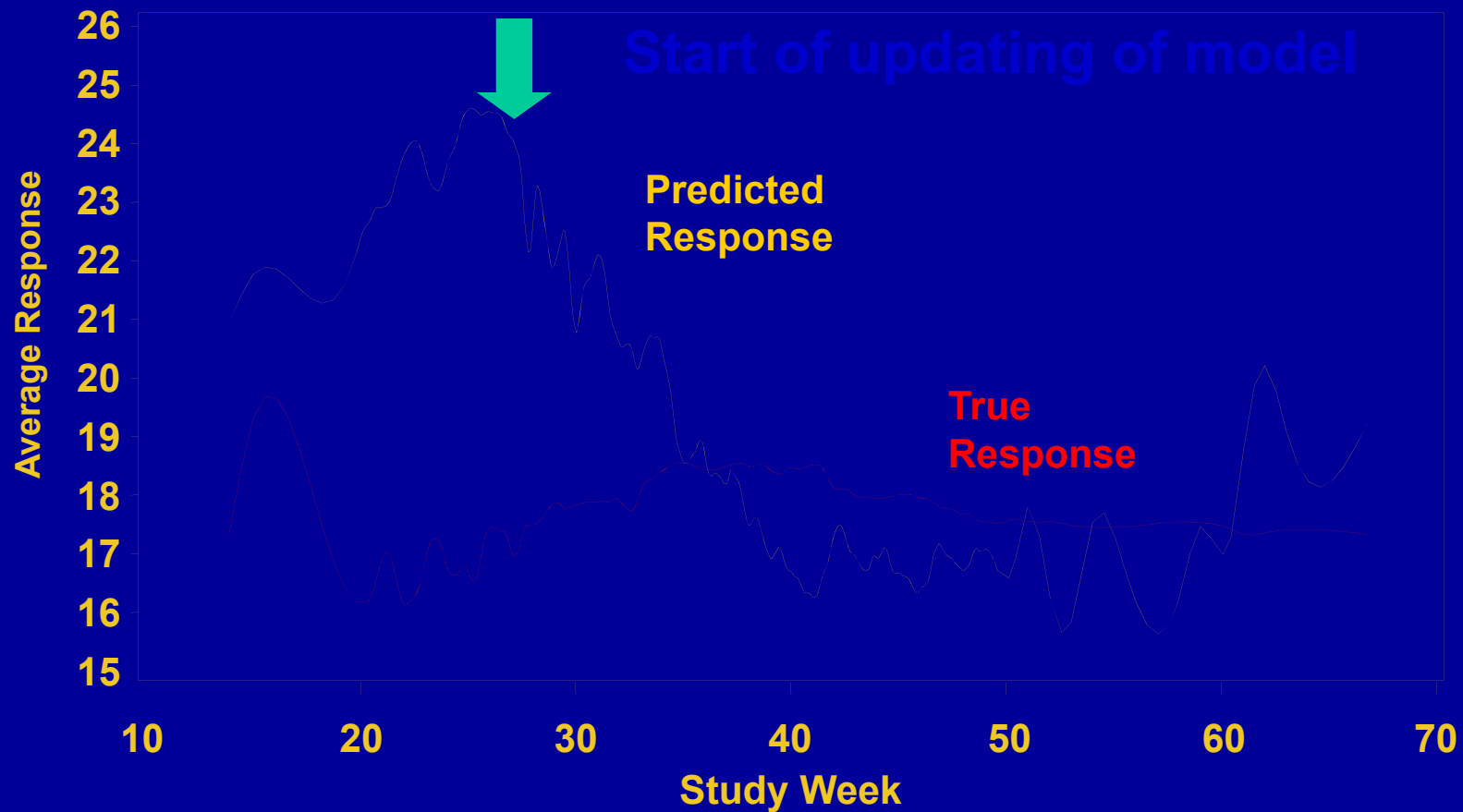
Neutralizing antibodies to UK-279,276



Comparison of Imputed and True Response—Initial longitudinal model



Comparison of Predicted and True Response—Updated longitudinal model



Mike Krams's conclusions

- **Complex sequential design possible**
- **Requires up-front effort**
 - **Simulation**
 - **Validation**
 - **Regulators: US, EMEA, UK, Germany, Sweden, Canada**
- **Saves time & resources**
- **Algorithm perceived as success**
 - **Hunted effectively**
 - **Stopped as soon as allowed, clear-cut decision**
 - **Could have stopped sooner!**

Consequences of Adaptive Approach

- Fundamental change in way we do medical research
- We'll get the dose right
- Better treatment of patients
- More rapid progress
- . . . at less development cost

OUTLINE

- **Bayesian adaptive design?**
- **Predictive probabilities in design**
- **Adaptive dose-finding:
A stroke trial**