

# **ADAPTIVE DESIGNS**

## **An Overview**

**Nancy Flourney**  
**University of Missouri**

**Fields Institute**  
**September 26, 2003**

# CONVENTIONAL EXPERIMENTAL DESIGNS

- All treatments, doses and/or treatment combinations are fixed prior to the study.
- Patients are randomized to the various treatments.
- The chance that any particular treatment will be selected for any particular patient remains constant throughout the study  
(e.g., randomize half of the subjects to an experimental treatment and half to a control).
- Sample size is fixed

# ADAPTIVE DESIGNS

- Subjects are treated sequentially, or in (two or more) groups.
- A treatment is assigned to each subject.
- The chance that a subject (group) will get a particular treatment changes as information accrues in the study

# REASONS FOR USING ACCRUING EXPERIMENTAL DATA TO CHANGE TREATMENT ALLOCATION PROBABILITIES

→ to improve      power,  
                         efficiency,  
                         safety,  
                         efficiency,  
                         model specification

# WHY THE INTEREST IN ADAPTIVE DESIGNS NOW?

- Theoretical advances
- Computational advances

# ADAPTIVE DESIGNS OUTLINE

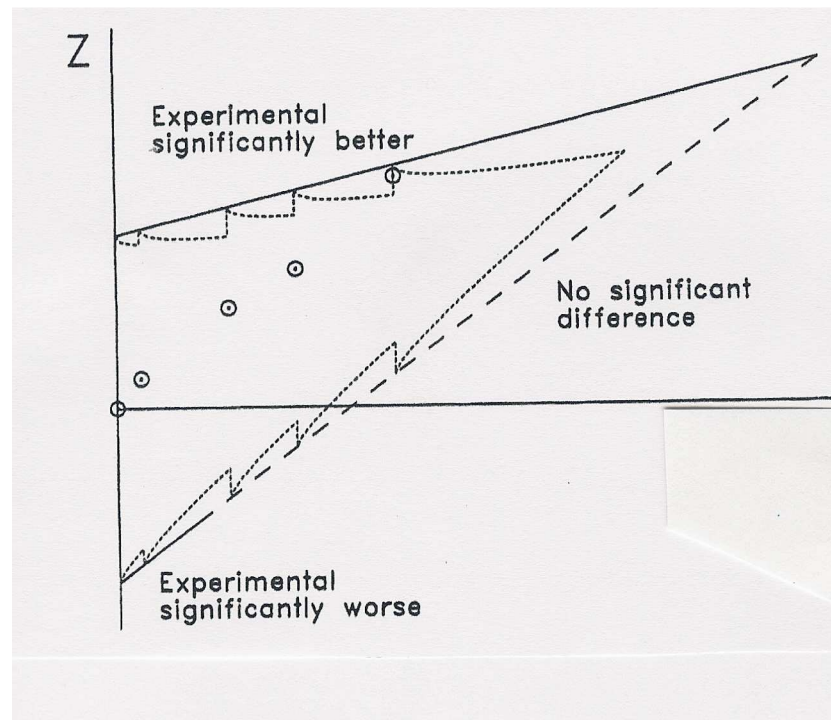
1. CONVENTIONAL VS ADAPTIVE DESIGNS
2. EARLY STOPPING
3. ADAPTING TO BALANCE SUBJECT ALLOCATION BETWEEN TREATMENT GROUPS
4. TWO STAGE DESIGNS
5. BAYESIAN DESIGNS
6. OPTIMAL DESIGNS & APPROXIMATIONS TO THEM
7. AD HOC DESIGNS
  - UP & DOWN DESIGNS FOR TOXICITY ASSESSMENT & PHASE I CLINICAL TRIALS
  - UP & DOWN DESIGNS FOR PHASE I/II TRIALS
  - AN OPTIMIZING URN DESIGN
8. FINAL COMMENTS

# EARLY STOPPING

## Sequential Analysis

- Subjects arrive sequentially.
- A treatment is assigned to each subject
- The probability that a subject gets any particular treatment remains constant.
- Outcomes are assessed sequentially to determine if the better treatment can be identified with the desired confidence, and the study terminated.

# A Triangular Early Stopping Rule





# ADAPTING TO BALANCE SAMPLE SIZES BETWEEN TREATMENTS WHILE PRESERVING THE BENEFITS OF RANDOMIZATION

- A biased coin design (Efron, 1971)
- **Talk 1.** Biased Coin vs Ehrenfest Urn: an analysis of randomness, balance and power  
*by Yung-Pin Chen*

Balance maximizes power  
*when groups have equal variances*

This is often a reasonable assumption  
when comparing means.

Let's consider comparing proportions  $p_T$   
and  $p_C$

1. Risk difference  $p_T - p_C$

2. Log odds metric

$$\log (p_T/(1 - p_T)) - \log (p_C/(1 - p_C))$$

3. Log risk ratio

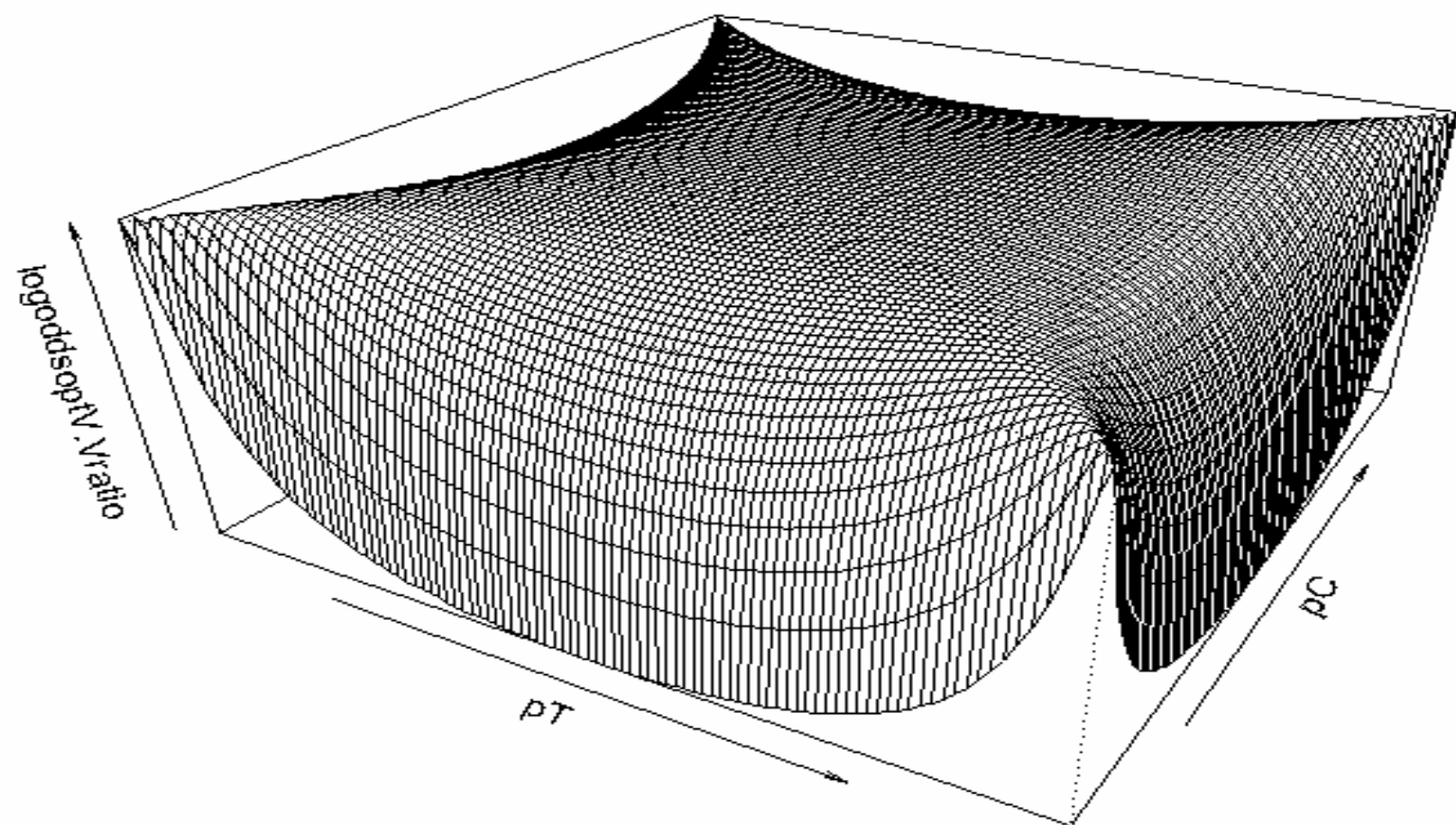
$$\log (p_T) - \log (p_C)$$

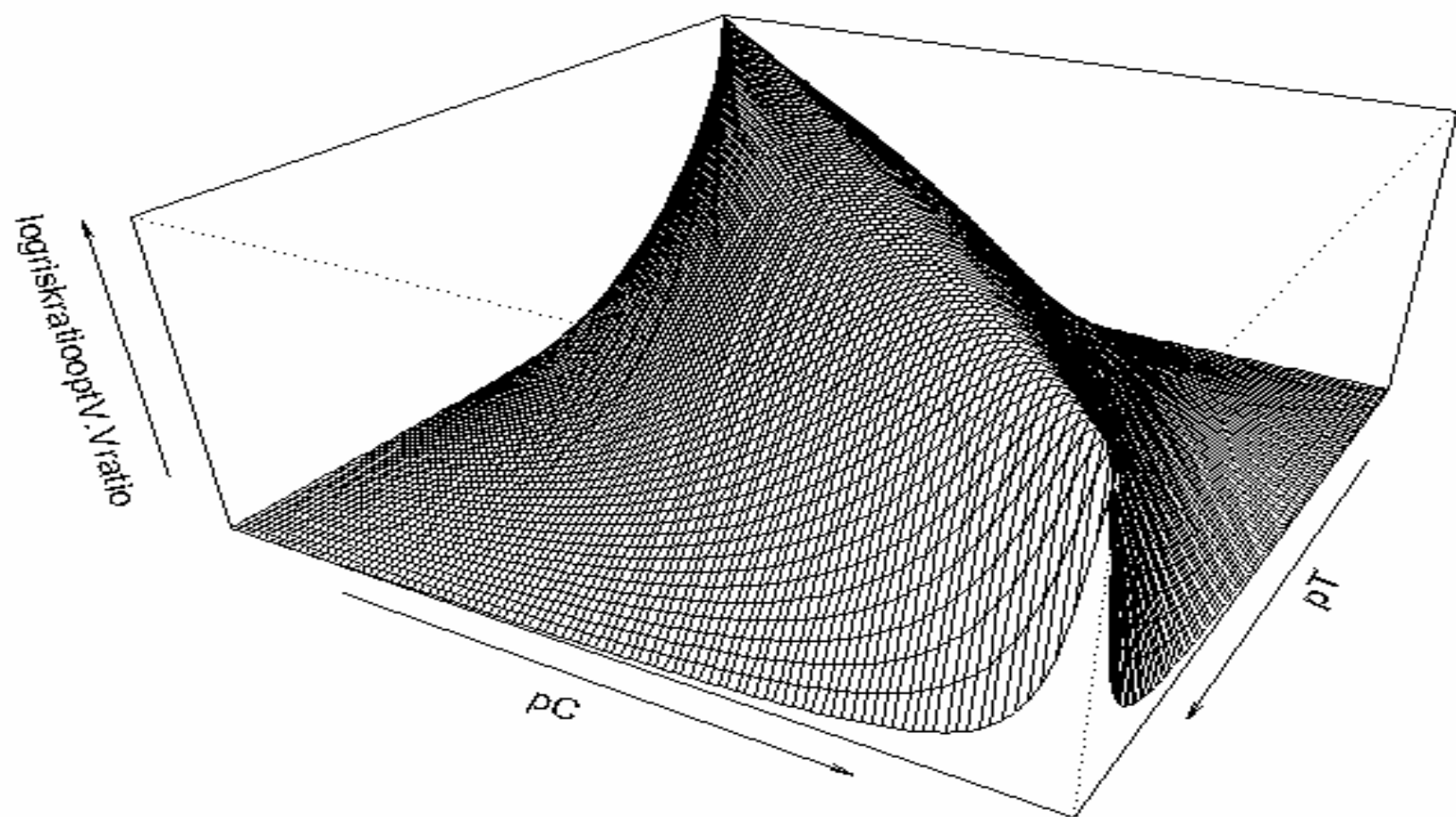
Calculate variance of each metric.

Evaluate each variance at  $n_T/n = 1/2$

Find  $n_T/n$  for which variances are a minimum,  
and evaluate variances at these minima

Plot optimal variance / variance at  $n_T/n = 1/2$





# TWO STAGE DESIGNS

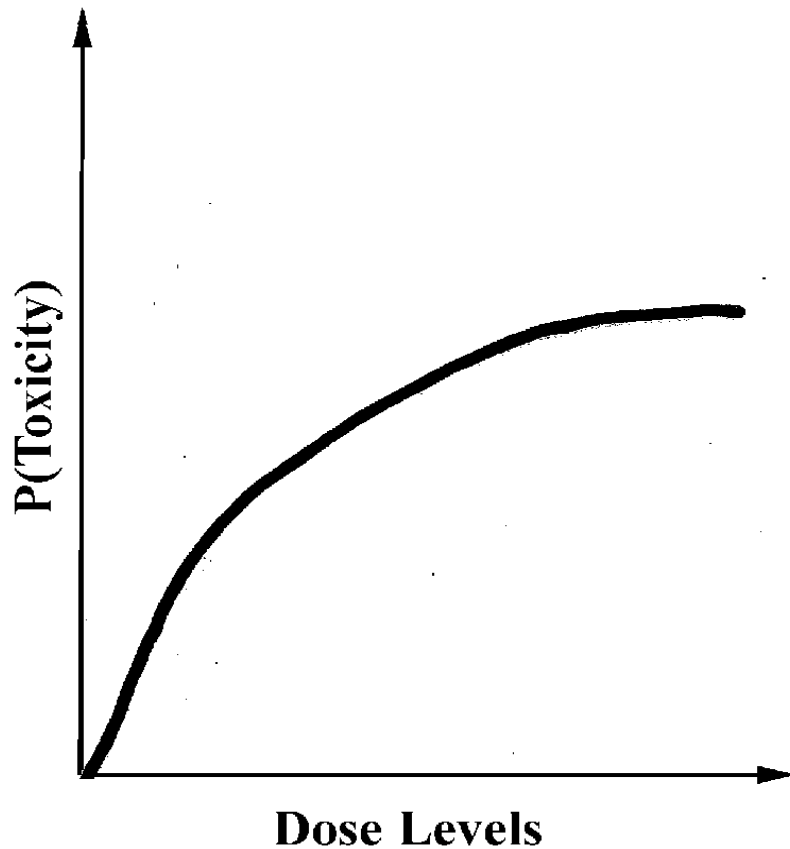
- **Talk 2.** A Calculus for Design of Two-Stage Adaptive Procedures *by Tatsuki Koyama*
- **Talk 3.** ADAPTIVE FACTORIAL EXPERIMENTS FOR MODEL IDENTIFICATION *by Subir Ghosh*

# RESPONSE - DRIVEN ADAPTIVE DESIGNS

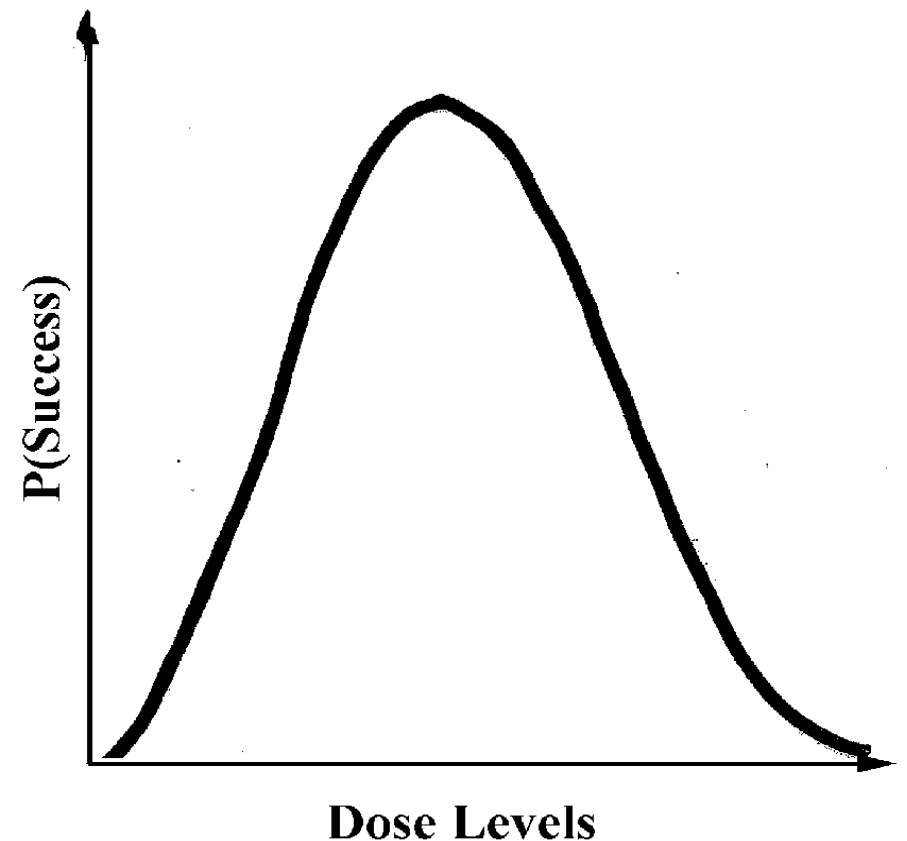
- Bayesian Designs
- Optimal Designs  
(Exact & Asymptotic)
- Ad hoc designs
  - Up-and-Down Designs
  - Urn Designs
  - Stochastic approximation

# MOTIVATING APPLICATIONS IN MEDICINE

**Increasing Response  
Function**



**Unimodal Response  
Function**





# BAYESIAN DESIGNS

- **Talk 4.** Individualized patient dosing in phase I clinical trials  
*by Andre Rogatko*
- **Talk 5.** Flexible Bayesian methods for cancer phase I clinical trials  
*by Mourad Tighiouart*

# OPTIMAL DESIGNS

- **Talk 7.** Optimal allocation in multi-armed clinical trials  
*by Yevgen Tymofyeyev*
- **Talk 8.** Bandit problems and adaptive clinical trials  
*by Xikui Wang*
- **Talk 9.** Optimal few-stage designs for clinical trials  
*by Janis Hardwick*

# AD HOC ADAPTIVE DESIGNS

## AN URN DESIGN

- **Talk 6** Minimized Hellinger distance estimations for randomized play the winner rule *by An-Lin Chen*

## AN UP-AND-DOWN DESIGN

- **Talk 12.** Up-and-down designs for phase I trials; an evaluation of different designs and estimators *by Hon Keung Tony Ng*
- **Also Talks 10, 11 & 13?**

# BIG ISSUES

## Estimation and Inference

- **Talk 6.** Minimum Hellinger distance estimation for randomized play the winner rule  
*by An-Lin Chen*
- **Talk 10.** Nonparametric likelihood for response adaptive randomization with delayed response  
*by Anand Vidyahankar*

# BIG ISSUES

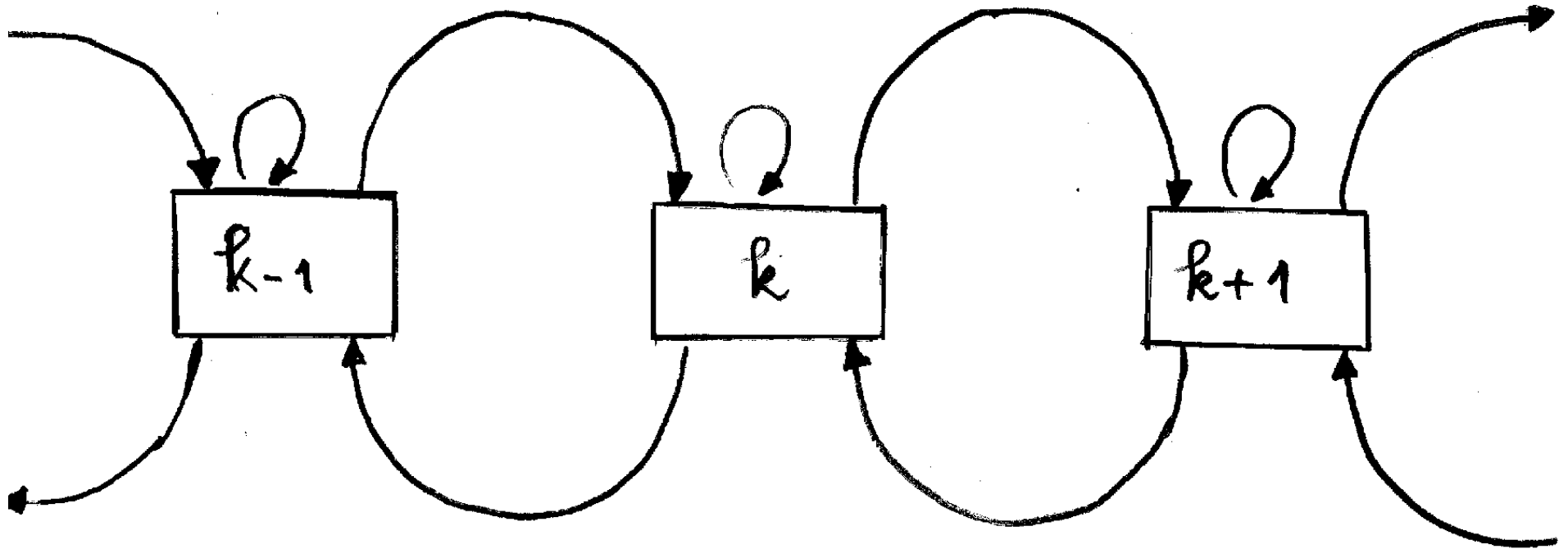
## efficiency and power

- **Talk 1.** Biased coin vs Ehrenfest urn: an analysis of randomness, balance and power *by Yung-Pin Chen*
- **Talk 9.** Optimal few-stage designs for clinical trials *by Janis Hardwick*
- **Talk 11.** Response-adaptive designs: ethics and efficiency of estimation *by Anastasia Ivanova*
- **Talk 13.** Response-adaptive designs: maximizing power and minimizing the expected number of failures *by Feifang Hu*

# UP-AND-DOWN DESIGNS

- FOR PRODUCT LABELING
- FOR CONTROLLING TOXICITY
- FOR DOSE-FINDING

# UP AND DOWN DESIGN



# ADVANTAGES OF UP-AND-DOWN DESIGNS

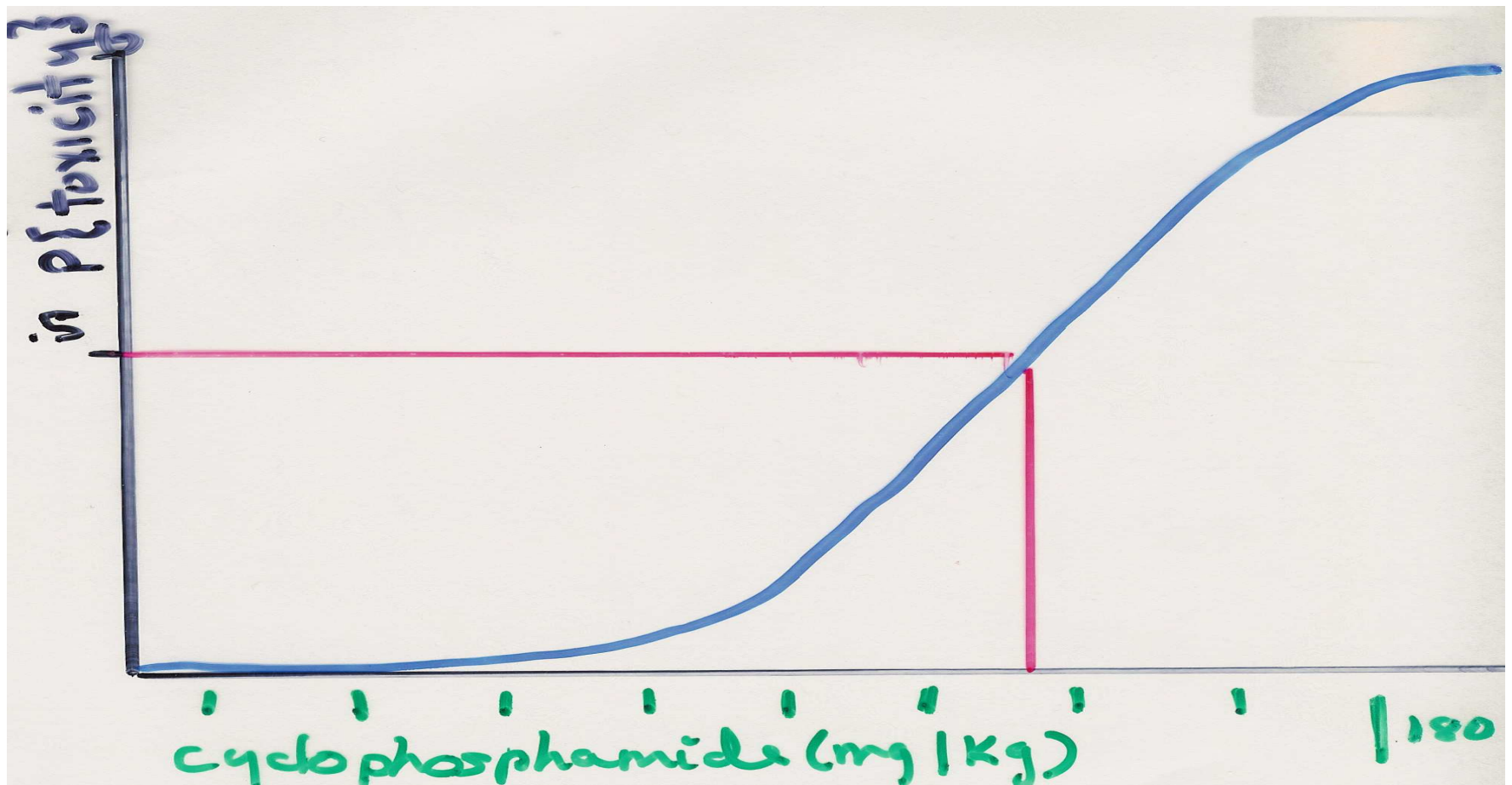
- **Cluster Doses around an unknown Target Dose**
- **Easy to Conceptualize**
- **Easy to Implement**
- **Changing Doses is Done Conservatively**  
(i.e., no large increments between subjects)
- **No Parametric Model**
- **Exact Distribution Theory Available:**
  - Durham, **Flourney**, Haghighi (1995). Up-and-down designs II: Exact treatment moments.  
**IMS Monograph**
  - **Flourney**, Durham, Rosenberger (1995).  
Toxicity in sequential dose-response experiments.  
**Sequential Analysis.**



# UP-AND-DOWN DESIGNS

WITH SEQUENTIAL  
ACCRUAL

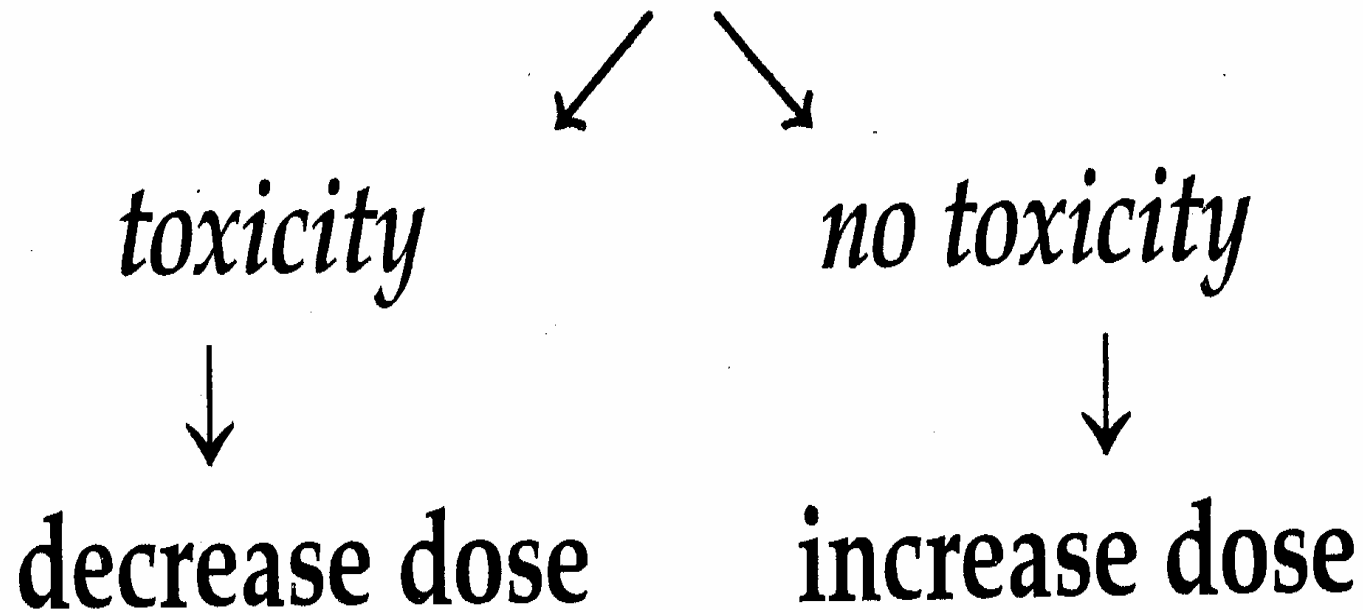
# TARGETING THE LD50



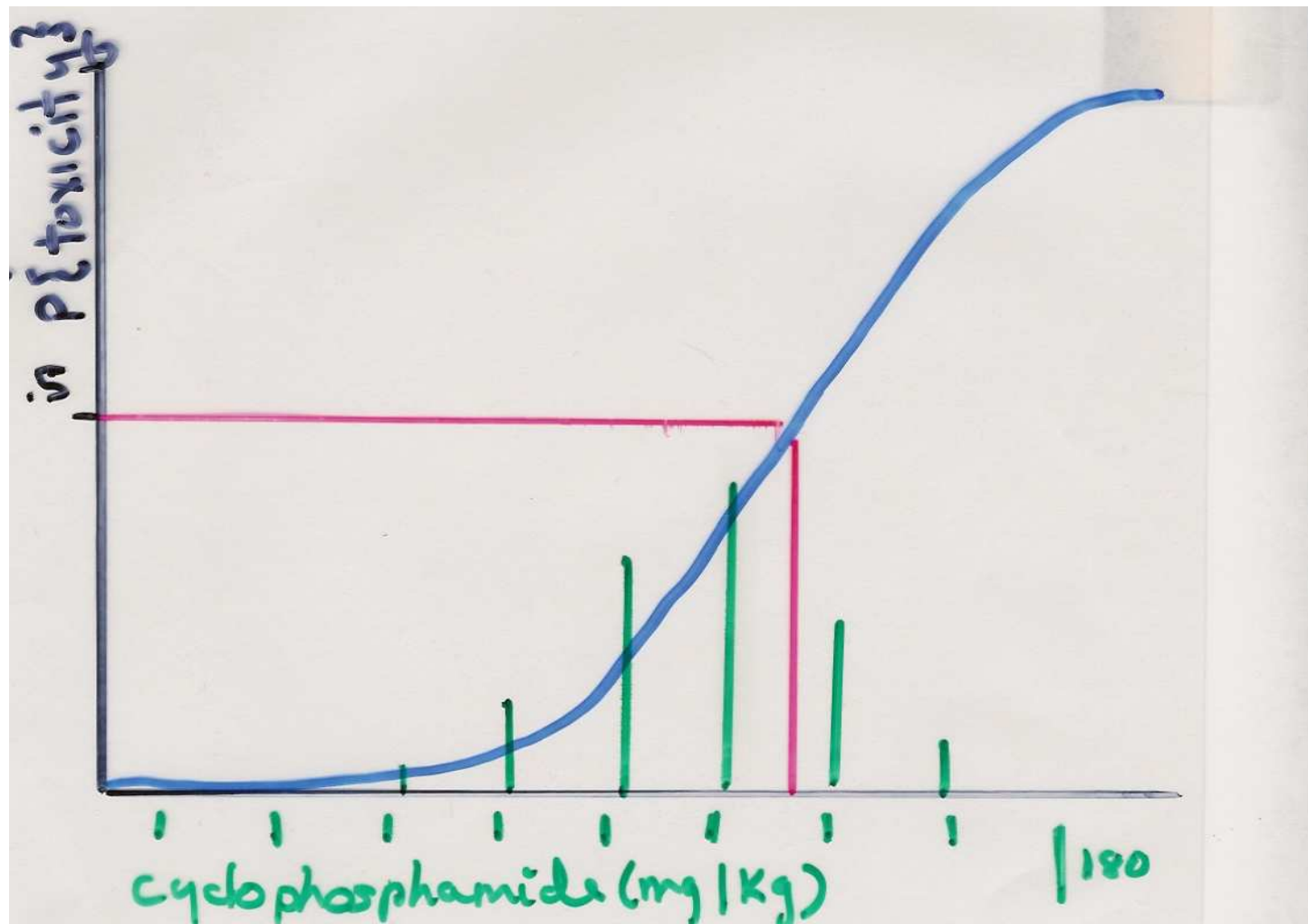
# DIXON AND MOOD (1948)

Given a Trial at Dose  $k$

---



# ASYMPTOTIC TREATMENT DISTRIBUTION TARGETING THE LD50



# THEOREM

For any increasing response function, the asymptotic treatment distribution is unimodal with mode  $\hat{\mu}$  less than  $\Delta$  away from the dose for which  $P\{\text{toxicity}\} = .5$

Durham, SD, **Flournoy, N.** (1994). Random walks for quantile estimation. **Statistical Decision Theory and Related Topics V**, 467-476. Springer-Verlag

# **EPA AND OECD “APPROVED”**

**UP AND DOWN DESIGN  
FOR  
PRODUCT LABELING  
ANIMAL STUDIES**

**Up-and-Down Procedure  
for Acute Oral Toxicity  
Updates and Announcements**  
<http://iccvam.niehs.nih.gov/>

**U.S. EPA Announces  
Availability of Revised Final  
Health Effects Test Guidelines:  
[Acute Oral Toxicity](#)  
Revised December 2002**

# EPA & OECD “Approved” Up-and-down procedure

Aim: sequential procedure to estimate  $LD_{50}$

Choose initial dose below guess of  $LD_{50}$

Test one animal

death: decrease dose by factor of  $\sqrt{10}$  for next animal

survival: increase dose by factor of  $\sqrt{10}$  for next animal

Continue until some stopping criterion is met

Estimate  $LD_{50}$  using maximum likelihood (assuming  $\beta$ )

obtain confidence interval from profile likelihood



# Test Guidelines/Acute Toxicity

## Acute Oral Toxicity Up-And-Down-Procedure

- [User Documentation for the AOT425StatPgm Program](#)
- [AOT425StatPgm](#) (This is a self-extracting zip file which will install the program on your computer in two steps...) *Note: This is to confirm that use of the computer program, AOT425StatPgm, developed by Westat for the US EPA , is freely given and there are no licensing restrictions in connection with its use.*
- [AOT Test Data Set](#) (This is a zip file which contains 15 test data sets, a result table, and instructions for their use to verify proper installation of the AOT425StatPgm program)
- [QA Testing for the AOT425StatPgm Program](#) - ( [PDF](#) )
- [Simulation Results for the AOT425StatPgm Program](#) - ( [PDF](#) )
- [Toxicology Guidance: Performance of the Up-and-Down Procedure](#) - ( [PDF](#) )
- [OECD Test Guideline 425: Acute Oral Toxicity - Up-and-Down Procedure](#) - ( [PDF](#) )
- [OPPTS Harmonized Test Guideline 870.1100 Acute Oral Toxicity](#) ( [PDF](#) )

Source: <http://www.epa.gov/oppfead1/harmonization/>

# STATISTICAL CONCERNS

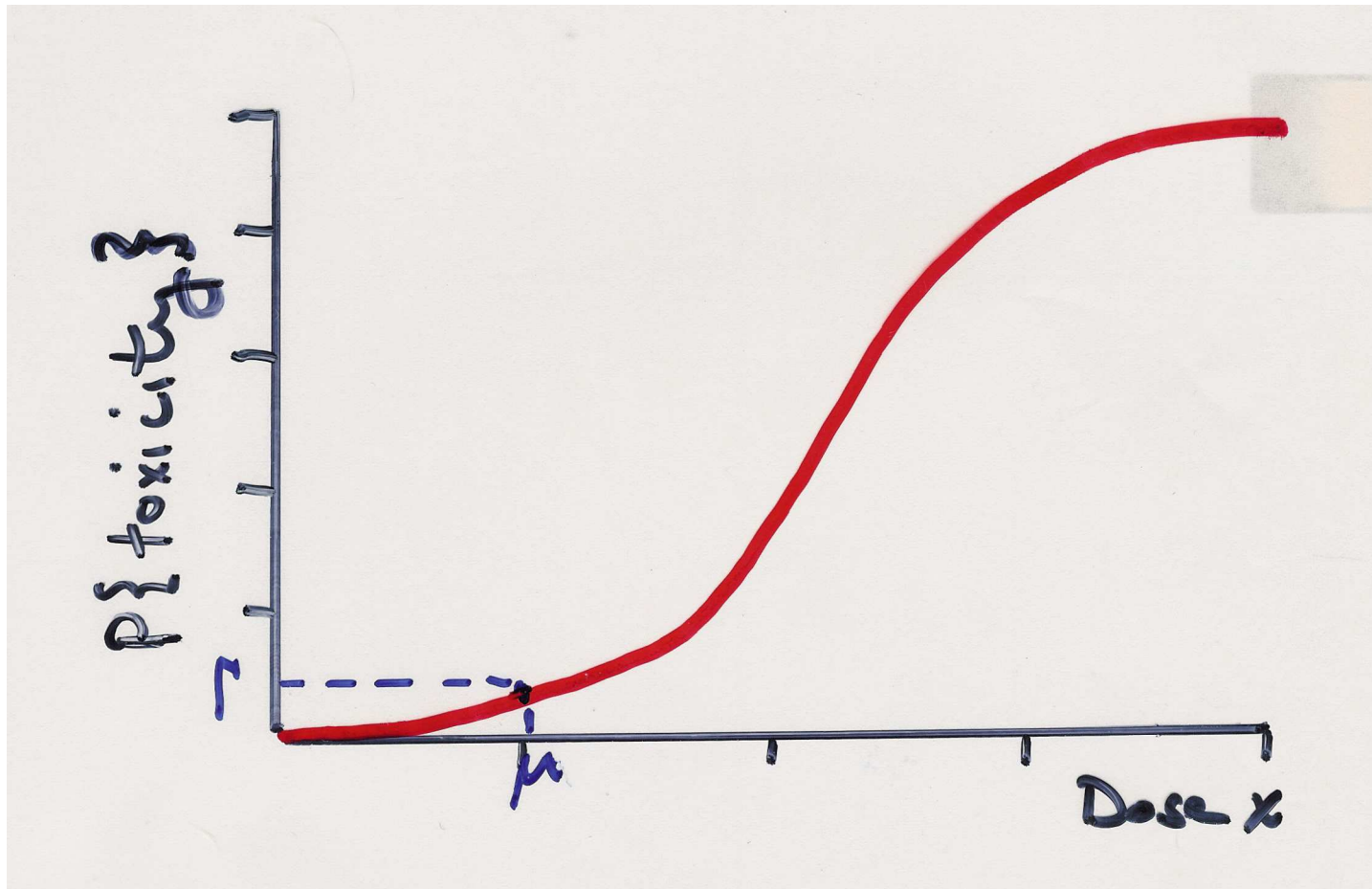
## UDP SHOULD NOT BE USED FOR

- RISK ASSESSMENT

Estimate LD<sub>p</sub>,  $p$  possibly small

- CONFIDENCE INTERVALS  
FOR THE LD<sub>50</sub>

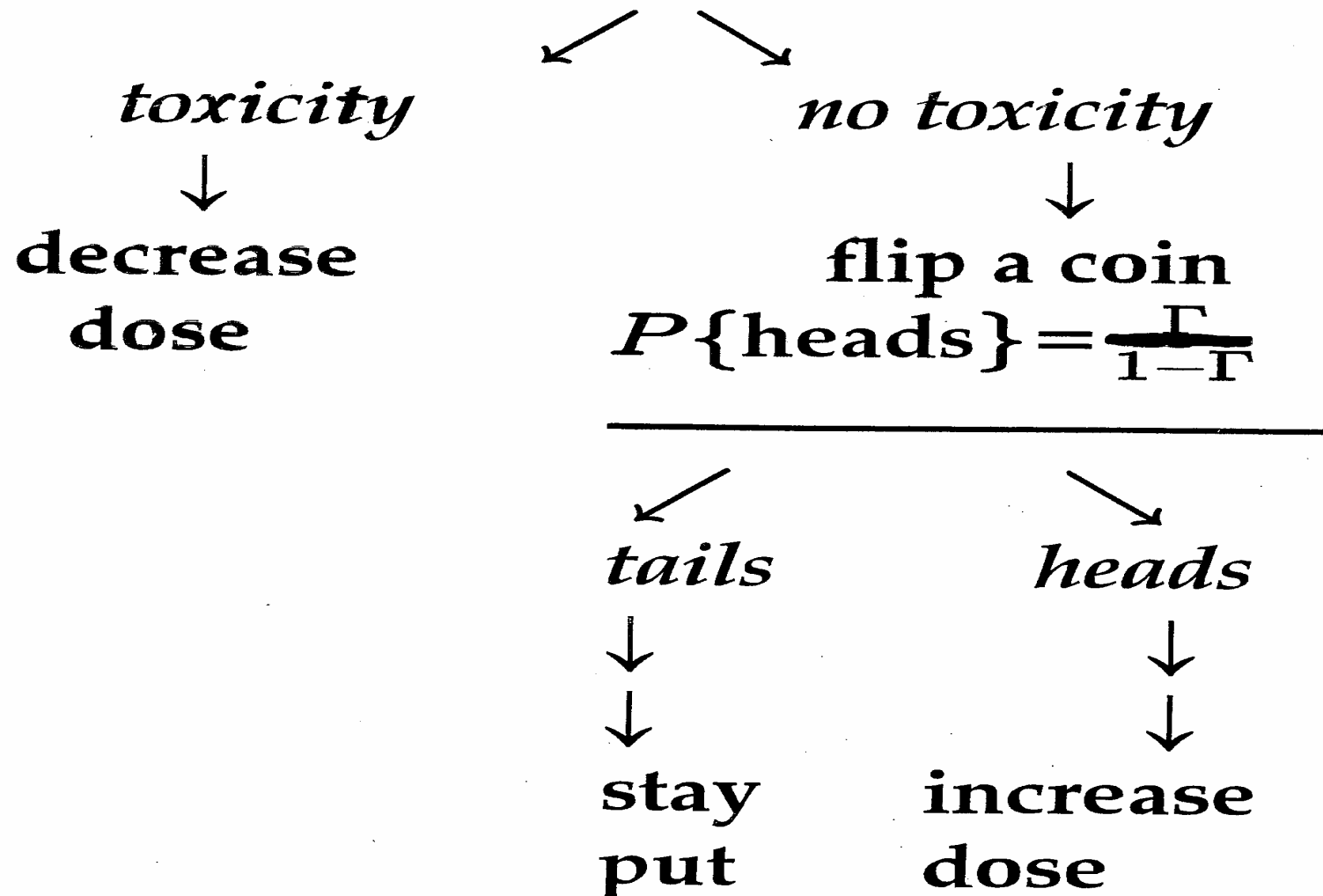
# TARGETING THE LD<sub>50</sub>



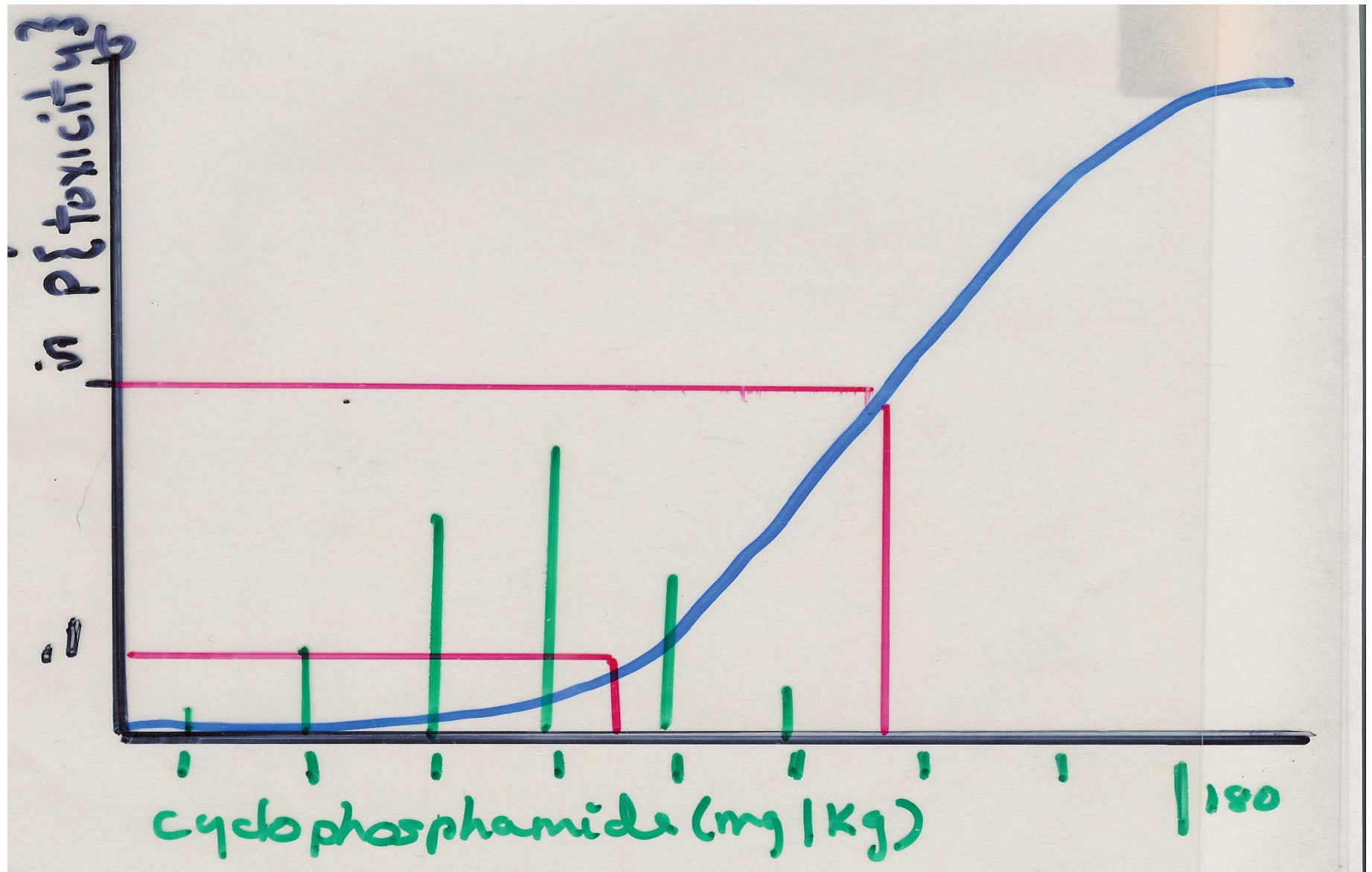
# BIASED COIN UP-AND-DOWN DESIGN FOR $\Gamma \leq .5$

Given a Trial at Dose  $k$

---



# ASYMPTOTIC TREATMENT DISTRIBUTION TARGETING LDF



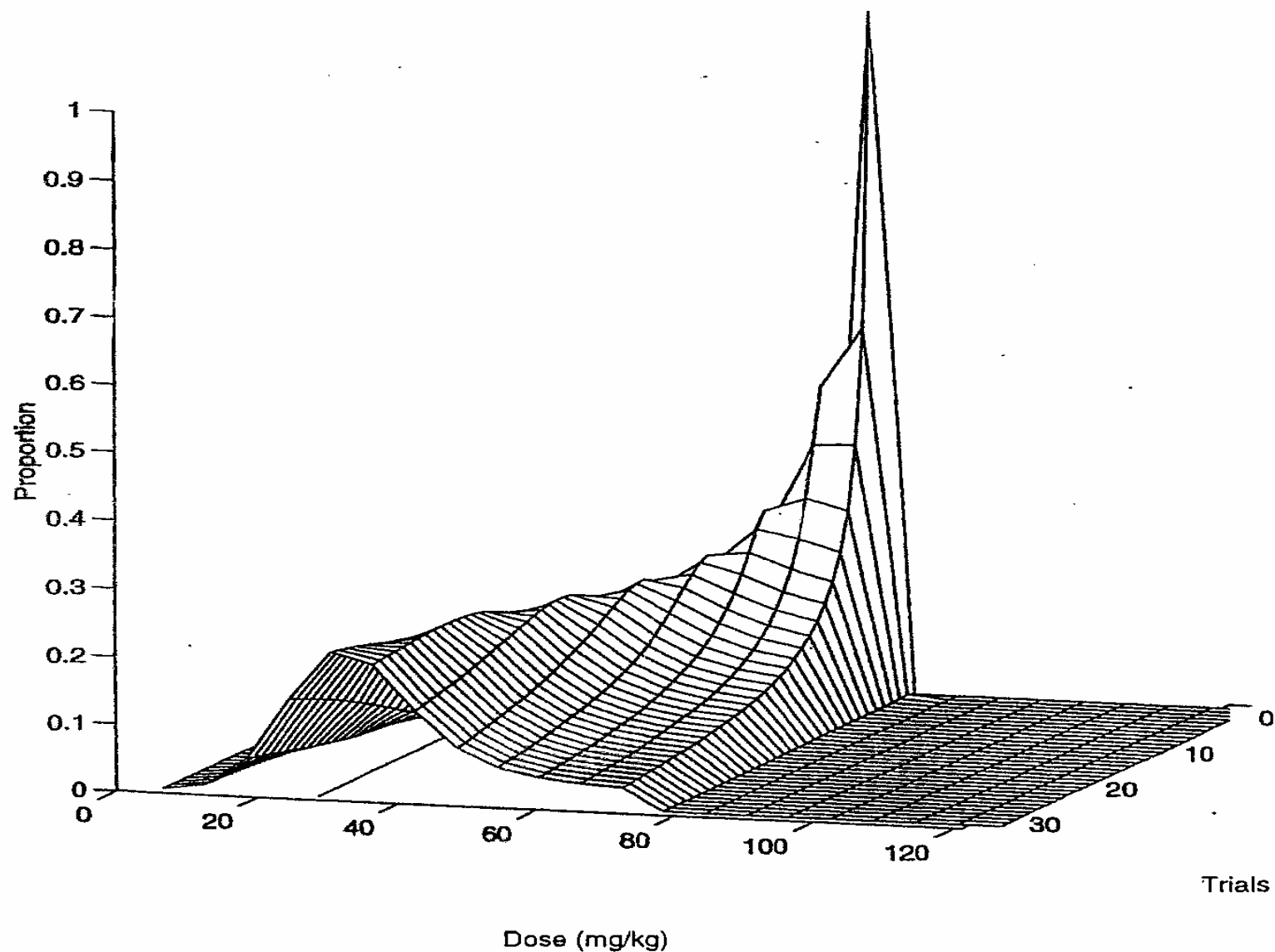
# THEOREMS

If treatments are selected according to the Biased Coin Up and Down Design, and the probability of response increases with dose,

Then, asymptotically,

- Treatments assignments are unimodally distributed around the target dose
- Mode of the treatment distribution is the largest dose  $\leq$  target dose
- $|\text{treatment mode} - \text{target dose}| \leq \text{interval between doses}$

WHAT ABOUT SMALL  
SAMPLE SIZES?



Expected proportions of subjects allocated at each dose level under the empirical response function. We interpolate between the points that occur at  $n = 1, \dots, 34$  and each possible dose level in  $x_1$ .



WE RECOMMEND  
USING SMOOTHED ISOTONIC  
REGRESSION  
TO  
ESTIMATE THE LD $\Gamma$

- Stylianou, M, **Flournoy, N** (2002).  
Dose finding using isotonic regression  
estimates in an up-and-down biased  
coin design. *Biometrics*.

# ISOTONIC REGRESSION is NONPARAMETRIC

If observed proportion of responses increase with dose, you are done.

Going from lowest dose toward the highest dose, whenever the empirical proportion of responses drops, average it with the one before.

# WHY NOT MLE?

- REQUIRES PARAMETRIC MODEL FOR RESPONSE FUNCTION, e.g. logistic
- DESIGN IS NOT GOOD FOR ESTIMATING THE SLOPE PARAMETER OF A PARAMETRIC MODEL
- OFTEN MLE DOES NOT EXIST FOR SMALL SAMPLE SIZES

# SOME OTHER UP-AND-DOWN DESIGNS

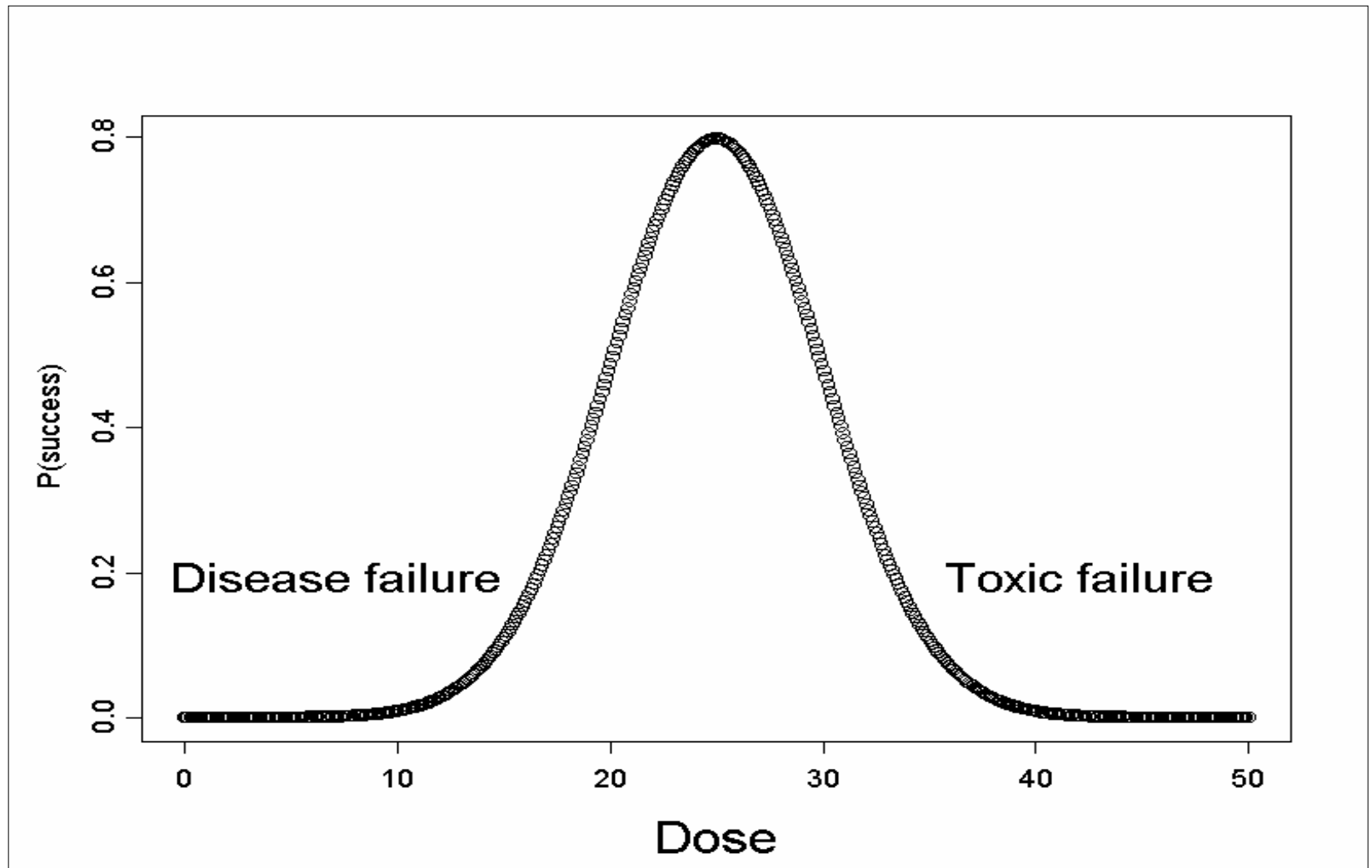
## FOR CONTROLLING TOXICITY

- Group up-and-down Designs (Gezmu & Flournoy)
- r-in-a-row (Gezmu & Flournoy)
- Moving Average (Ivanova, Mohanty & Durham)
- Narayana's (Ivanova, Mohanty & Durham)

Talk 12. comparisons *by Hon Keung Tony Ng*

# Optimizing Up-and-Down Designs

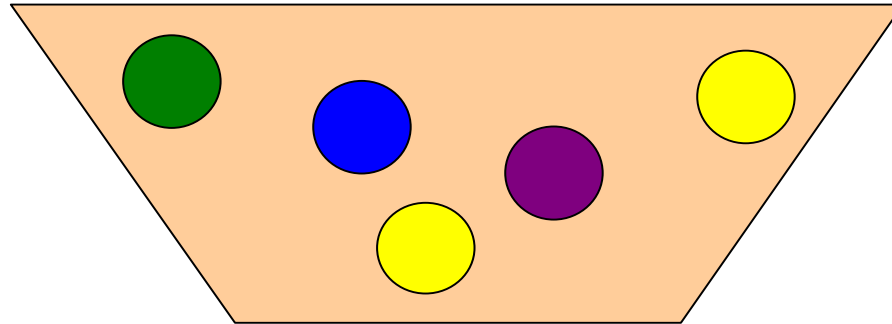
## Cluster Subjects Around Optimal Dose



# OPTIMIZING URN DESIGNS

# PURE BIRTH URN DESIGN

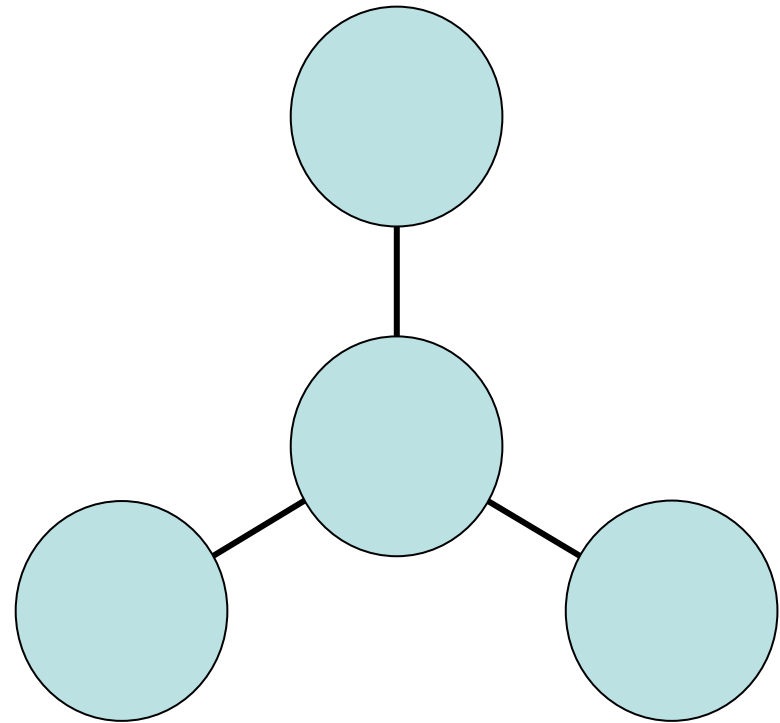
Li, Durham & Flournoy



- Draw a ball and replace it.
- If ball is color  $i$ , give treatment  $i$ .
- If that treatment is successful,  
add another color  $i$  ball.
- If that treatment is a failure, do nothing.

# PURE BIRTH URN DESIGN

- Suppose the best treatment corresponds to the green balls.
- Green balls will proliferate until virtually all the balls are green.





$$\alpha_k \equiv P\{\text{success given treatment } k\}$$

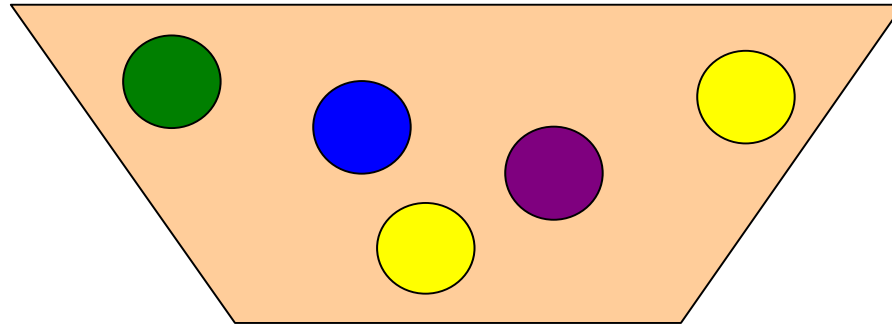
## THEOREM

If  $\max\{\alpha_1, \dots, \alpha_K\} = \alpha_i$ ,  
the proportion of type  $i$  balls  
 $\rightarrow 1$

wp 1 as  $n \rightarrow \infty$

# PURE DEATH URN DESIGN

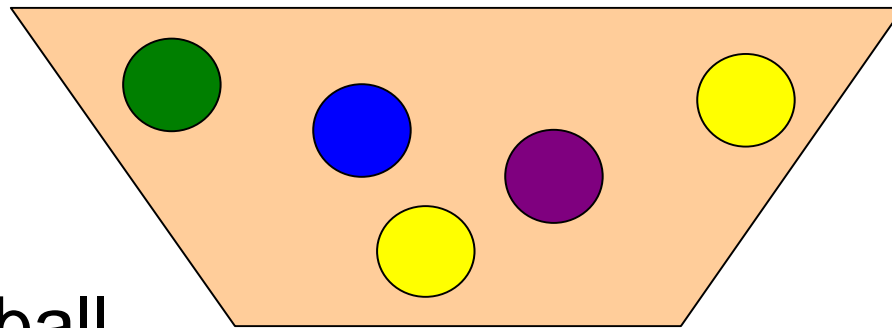
Ivanova



- Draw a ball and replace it.
- If ball is color  $i$ , give treatment  $i$ .
- If that treatment is successful,  
replace the drawn ball.
- If that treatment is a failure, do not  
replace the drawn ball.

# BIRTH AND DEATH URN DESIGN

Ivanova, Rosenberger, Durham & Flournoy



- Draw a ball.
- If ball is color  $i$ , give treatment  $i$ .
- If that treatment is successful,  
add another color  $i$  ball.
- If that treatment is a failure,  
do not replace the drawn ball.

# RANDOMIZED PLAY THE WINNER

Does it have a future?

Talk 6 (others?).

# ADAPTIVE DESIGNS OUTLINE

1. CONVENTIONAL VS ADAPTIVE DESIGNS
2. EARLY STOPPING
3. ADAPTING TO BALANCE SUBJECT ALLOCATION BETWEEN TREATMENT GROUPS
4. TWO STAGE DESIGNS
5. BAYESIAN DESIGNS
6. OPTIMAL DESIGNS & APPROXIMATIONS TO THEM
7. AD HOC DESIGNS
  - UP & DOWN DESIGNS FOR TOXICITY ASSESSMENT & PHASE I CLINICAL TRIALS
  - UP & DOWN DESIGNS FOR PHASE I/II TRIALS
  - AN OPTIMIZING URN DESIGN
8. FINAL COMMENTS

# BIG ISSUES

POWER  
EFFICIENCY  
ESTIMATION  
INFERENCE