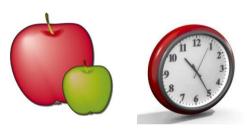
# **Bayesian Clinical Trial Software Overview**

John D. Cook M. D. Anderson Cancer Center

#### **Software outline**



**Basic utilities** 



Safety monitoring

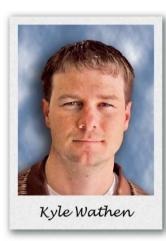


**Randomized trials** 



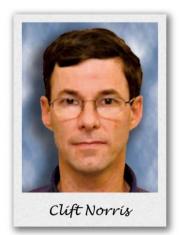
Dose finding

#### **Software developers**

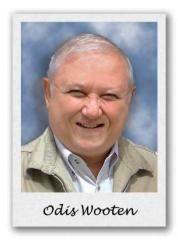












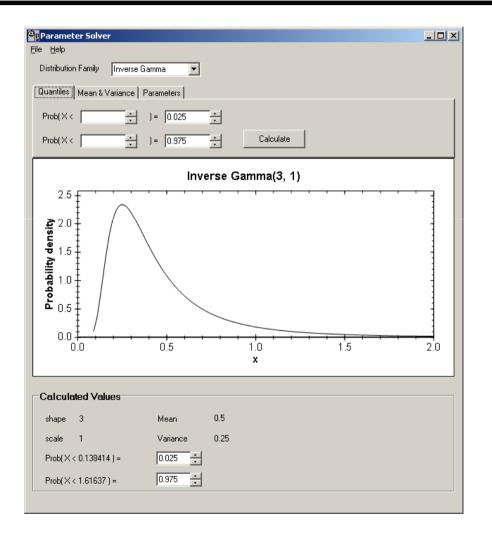
# **Simplest distributions**

- Discrete distribution: binary
- Conjugate prior: beta
- Time distribution: exponential
- Conjugate prior: inverse gamma

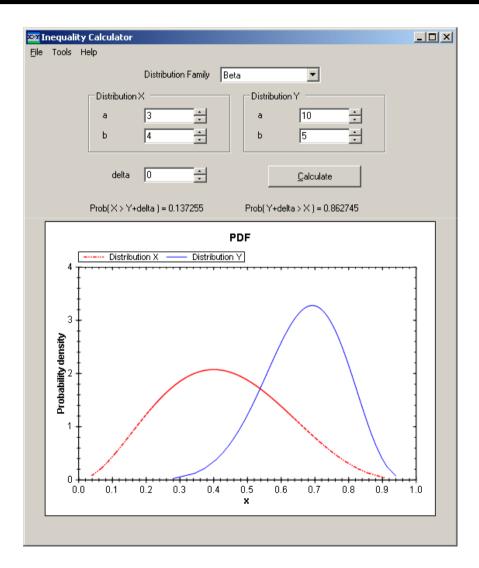
# **Random inequalities**

- P(X > Y)
- $P(X > Y + \delta)$
- Proportion: binomial / beta
- Mean time: exponential / inverse gamma
- Randomization probability
- Stopping rule

#### **Parameter solver**

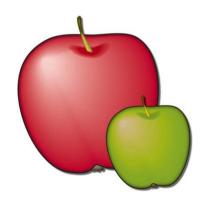


#### **Inequality calculator**



### **Multc Lean software**

- Original software Multc99, developed by Hsi-Guang Sun
- Greatly simplified, added Windows UI
- Added trial duration simulation



# **TTEConduct** software

- Table allows look-ahead: Not just "Should I stop?" but "What would cause the trial to stop?"
- NB: design in months, conduct in days
- Simulation software in development



# **Criticism of exp / IG model**

- Exponential survival time model is usually a poor fit to reality. Hazard not constant.
- Exp / IG model chosen for convenience (conjugate, trivial posterior calculation)
- Nevertheless, model robust in practice (Thall and Wooten)

# What's going on?

- We're not modeling survival per se, we're making a stop/go decision.
- Survival isn't exponential but a continuous mixture of exponentials.

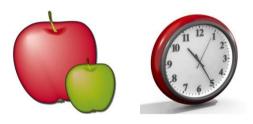
# **Monitoring with Bayes factors**

- Not based on random inequalities
- Alternative must really be alternative
- Better operating characteristics
- Command line software

#### **Software outline**



**Basic utilities** 



Safety monitoring

Randomized trials

Dose finding

### **Stop for futility: Predictive Probability**

- Binary and time-to-event outcomes
- Three decisions: A, B, neither
- If P(neither) is large, stop for futility



### **Adaptive Randomization**

- Randomize, but not equally
- Increase the probability of assigning what appears to be the best treatment
- Compromise between equal randomization
  and myopic optimization

#### **Tuning parameter c**

Let 
$$p_A = P(\pi_A > \pi_B)$$
 and  $p_B = P(\pi_B > \pi_A)$ .

Assign treatment A with probability

$$\rho = \frac{\mathfrak{p}_A^c}{\mathfrak{p}_A^c + \mathfrak{p}_B^c}$$

#### **Special values of c**

- If c = 0,  $\rho = 0.5$ . Equal randomization
- If c = 1,  $\rho = p_A$ . Common choice (proposed in 1933!)
- As  $c \to \infty$ ,  $\rho \to [p_A > p_B]$ Myopic optimization

• See tech report for all values of c

# **Adaptive Randomization software**

- Supports binary and TTE outcomes
- Up to 10 arms
- Equal randomization special case



# **CRMSimulator software**

- Emphasis on ease-of-use, not generality
- Contains features commonly used at MDACC, and no more

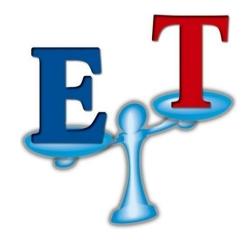


# **CRM misunderstandings**

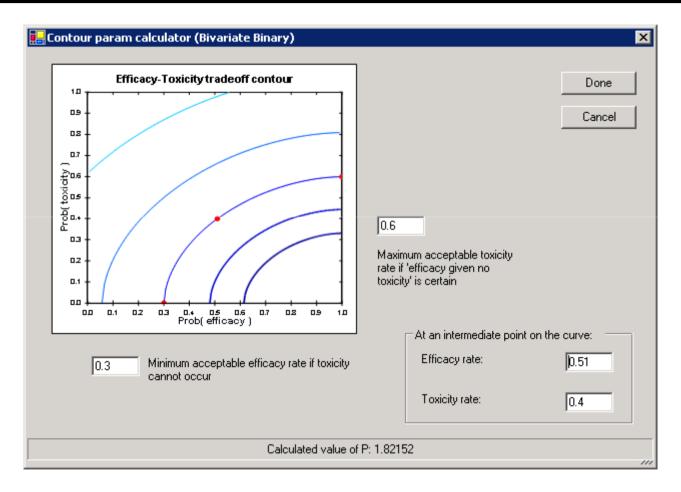
- Not "dose escalation"
- Law of small numbers
- Doesn't fit 3+3 expectations

# **EffTox dose-finding**

- Minimize toxicity, maximize efficacy
- Investigator specifies trade-off
- Uses twice as much data per patient
- Uses dose values, not just dose order



#### Heart of the method



### **Software summary**



**Basic utilities** 



Safety monitoring



**Randomized trials** 



Dose finding

# Links

- https://biostatistics.mdanderson.org/ SoftwareDownload/
- http://www.JohnDCook.com