### The rheumatoid arthritis drug development model: a case study in Bayesian clinical trial simulation

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### **Problem statement**

What decisions should be made about a Phase IIb and Phase III study for a new Rheumatoid Arthritis treatment?

- Rheumatoid Arthritis
  - A chronic, progressive, inflammatory disease which affects about 0.5% -1% of adults
  - Traditional Disease-Modifying Anti-Rheumatic Drugs lots of them
    - Methotrexate (MTX) most effective
  - Biologic more effective and more costly
    - Etanercept, infliximab, adalimumab (TNF-α), anakinra (IL-1 inhibitor)
  - A new drug we wish to test
- We need to make decisions about the devolvement program
  - Decisions about each study design
    - Sample sizes?
    - Exposure duration?
    - ...
  - Stopping rules for the program
    - Efficacy thresholds? Safety thresholds?

### Overview of the Decision Analysis method What is needed for a Decision Analysis model

Strategies

Collections of decisions that must be made about study design whose effects are simulated

• Sample size, comparator, endpoint, exposure, patient population, stopping rules

Consequences and effects of the decisions, plus other relevant variables, which the model will incorporate

- Treatment efficacy and safety
- Recruitment rates, drop out rates, costs



The final measures of the design, which the model will calculate, and by which we will evaluate candidate strategies

• Probability of success (registration), time LPLV, cost

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Information

# A decision hierarchy identifies issues to be decided and issues already decided or that can be deferred.



- Policy
- Environment
- Decisions already made
- Near- and long-term strategic direction
- Near-term significant resource commitments
- Issues that must be resolved today
- Later significant resource commitments
- Decisions for specialists
- Operational or tactical decisions

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### Decisions

# Rows have no meaning - options from different columns may be combined

Decisions already made									
Both studies		Phase IIb			Phase III				
Average disease duration	rage Stopping rule: safety Comparator criteria		Doses	Stopping rule	Comparator	Dose	Stopping rule	Exposure duration	
8 years	1) SC1 withdrawal > 10% 2) SC2 withdrawal >25% 3) SC3 significantly different from MTX	MTX	L, M1, M2, H	<ol> <li>Fail superiority to MTX</li> <li>Fail non-inferiority to active comparator (indirect comparison)</li> </ol>	MTX + Etanercept	Lowest successful dose in Phase IIb	Fail non- inferiority to active comparator	6 months	

Decisions to make now								
	Phase IIb	Phase III						
End point	Sample size per arm	Exposure duration	Sample size	Non-inferiority margin				
ACR20	40	3 months	150	0.7				
ACR50	60	6 months	200	0.8				
	80		250	0.9				
	100							

- (1) SC = safety criteria
- (2) ACR20, ACR50 binary outcome which indicates a 20% or 50% improvement over a given time period

**Strategies** 



### Effectiveness

- Two data sources
  - Phase 3 trials for biologics (snippet of data below)
  - Early 1 month Phase 2a trial

Drug	Regime	N	1 Month ACR50	3 Months ACR50	6 Months ACR50
Anakinra	Placebo	121	NA	6	10
	30mg day	119	NA	NA	20
	75mg day	116	NA	12	13
	150mg day	116	NA	9	22
Anakinra	MTX	251	NA	15	20
	100mg day	250	NA	33	43
Etanercept	MTX	228	10	61	91
	25mg 2wk+MTX	231	44	95	133
	25mg 2wk	223	35	79	92

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# **Effectiveness Prediction Functions**

- Use Phase III data set to estimate
  - Odds ratios between different treatments at the same time points
    - by a mixed-treatment-comparisons meta-regressions
  - Predict probability of ACR event at 3 or 6 months from 1 or 3 months
    - By logistic regression with random-effects
  - These can be functions of different treatment and disease duration
- Use Phase IIa study to predict the probability of ACR given new treatment compared to MTX



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3 month withdrawal probabilities if given MTX + biologic treatment at dose d

 $\pi_{sc1}(3, d)$  Probability of withdrawing because of SC1

 $\pi_{sc2}(3,d)$  Probability of withdrawing because of SC2

 $\pi_{sc3}(3,d)$  Probability of withdrawing because of SC3

6 month withdrawal probabilities are twice 3 month probabilities



# Probability of withdrawing because of SC1 safety if given MTX + biologic treatment at

 $\pi_{sc1}(3,d) = \lambda(1 - \exp(-d\beta))$ 

dose d after 3 months

- Relative risk of withdrawal if given MTX + {M1}mg compared to MTX + {H}mg
  - So get an (implicit) distribution for eta
- Risk of withdrawal if given MTX + {M1}mg  $\gamma(1 \exp(-\{M1\}\beta)) \sim Beta(2.2,59.7)$ 
  - So get an (implicit) distribution for  $\gamma$



$$\frac{1 - \exp(-\{M1\}\beta)}{1 - \exp(-\{H\}\beta)} \sim Beta(16.1, 8.2)$$



## Elicitation

- Suppose we wish to elicit a distribution for a risk
- The experts judge the percentiles of the risk to be

5 <sup>th</sup>	50 <sup>th</sup>	95 <sup>th</sup>
0.1	0.2	0.35

Find a and b to minimize

$$(F_{a,b}(0.1) - 0.05)^2 + (F_{a,b}(0.2) - 0.5)^2 + (F_{a,b}(0.35) - 0.95)^2$$
  
CFD of a beta distribution

Find a Beta(5.8, 22.3) distribution

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# Study simulation model

How decisions, information and values are linked



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# Clinical Trial Simulation vs Bayesian Clinical Trial Simulation



- **Clinical trial simulation** 
  - Can estimate expected results from complex trials
  - But parameters are fixed
    - Bayesian clinical trial simulation
      - To compute PoS we must also simulate parameters
      - This is done in the same loop and needs no extra simulated trials
      - Average over the unknown parameters



Simulate parameters

Simulate patients

**Determine trial** 

results

Collate results

and outcomes

### Probability of success depends on design Could pick a design that gives maximum PoS

Value



### Study results Dig into where studies are failing



Phase IIb		Phase III		Phase IIb				Phase III			
End point	Sample size	Exposure	Sample size	Non- inferiority margin	PoS	Fail Non- inferiority	Fail Superiority	Fail Safety	PoS (Registration)	Fail Non- inferiority	Fail Safety
ACR20	80	3 months	200	0.8	7.8%	91.8%	42.6%	5.2%	4.7%	1.7%	0.16%
ACR20	80	6 months	200	0.8	6.8%	89.9%	40.1%	2.2%	4.7%	1.5%	0.02%

- The overall probability of successful drug registration is the same in both cases
  - But a 6-month study has a slightly smaller chance progression from Phase 2b to Phase 3
  - This is good as it stops the program before the expensive study



### Impact of larger Phase IIb trials Size of the Phase IIB is key driver of PoS



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Value

### Sensitivity analysis: the Tornado Diagram Not calculated during this work, but are a useful way of assessing which uncertainties have most influence on value



# What does decision analysis bring to trial design?

- Comprehensive approach that evaluates many different combinations
- Considers interactions of options
- Accounts for uncertainty in assumptions
- Evaluation of tradeoffs beyond statistical power

