

# **Application of a Bayesian approach to treatment selection in a rare disease sub-population**

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# Overview

- Disease background and study objectives
- Initial frequentist and a Bayesian approach
- Introduce the mathematics of the model and ideas around model calibration
- Questions and discussion

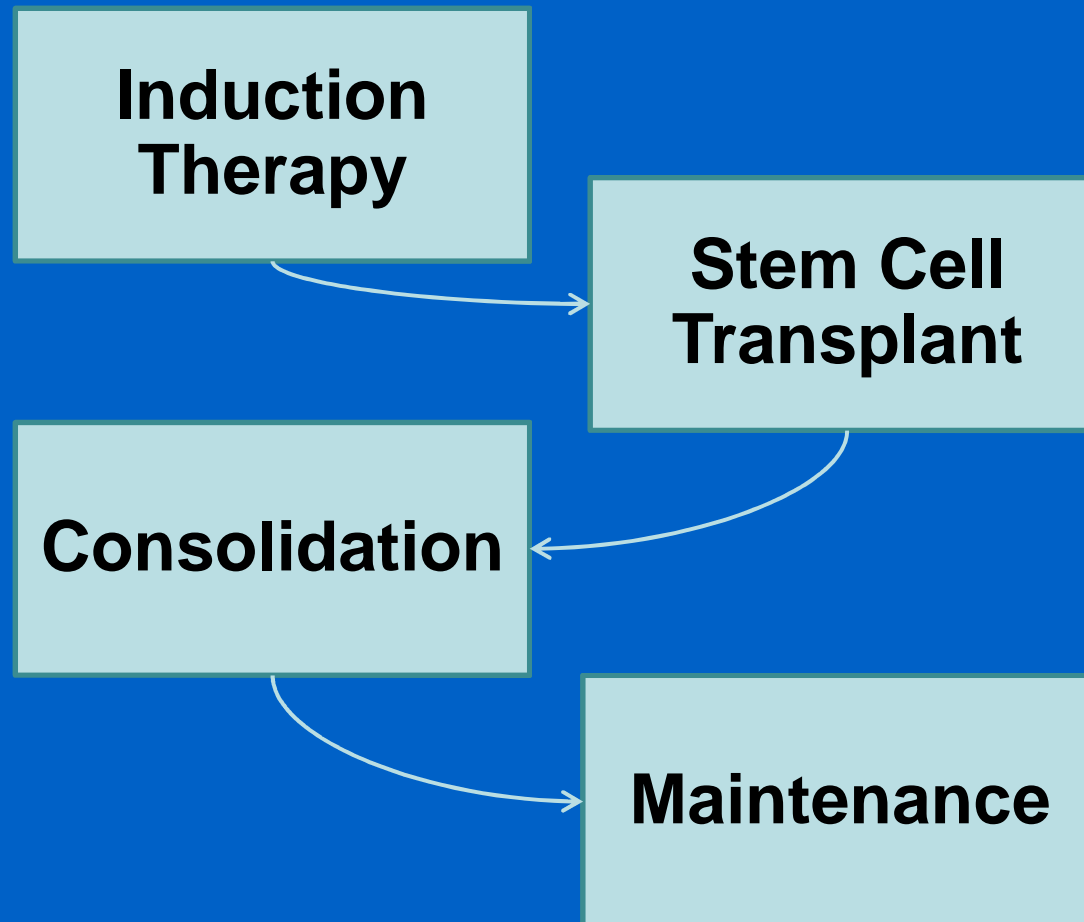
# High-Risk Multiple Myeloma

- Multiple myeloma is a cancer arising from plasma cells, a type of white blood cell which is made in the bone marrow
- Median age of diagnosis is 69, 37% < 65 with OS ~ 10 years
- A group of patients have poor prognosis with OS < 3years
- High risk disease defined by genetic lesions and gene expression profiles associated with poor outcome

# A Rare Population

- Myeloma represents 1.5% of all malignant diseases, and representing 4500 new cases each year in the UK
- High- risk multiple myeloma (HRMM) occurs in approximately 20-30% of MM patients
- <math><6/100,000</math> per year, representing a rare cancer or sub-population

# Treating newly diagnosed Myeloma



# Treating High-Risk disease

- Best practice currently unknown in UK, current therapy is ineffective
- Two approaches considered for induction therapy

**High Intensity**

**Low Intensity**

- Design a trial to assess whether we can improve outcomes for HR patients by selecting the optimum treatment strategy to take forward for further research

# Initial Approach to Trial Design

- A randomised controlled phase II trial selection trial was proposed (3 arms)
- Progression free survival primary endpoint

## Problems

- Would result in sample sizes of up to 450 patients i.e. screening 2500 newly diagnosed patients
- Not comparing like for like –deliverability of treatment is important

# Bayesian Approach

- Reverse philosophy - How many patients can we recruit and whether this amount of data has sufficient value to justify the trial design
- Recruit 120 patients
- Thall and Sung (1998) Designs for single-arm clinical trials with multiple binary outcomes
- Chosen because allows multiple interims for futility



# A Control Arm?

- A 'standard' control arm for newly diagnosed MM patients is difficult to define, particularly while the Mye XI/XI+ trials are ongoing and results awaited.
- The data from Mye XI/XI+, would effectively provide almost concurrent control data
- Efficiencies in the overall sample size
- Only possible as CTRU conducting Mye XI/XI+

# Multiple endpoints

## Final analysis

- Progression free survival at 18 months

## Interim analyses – conducted after every 10 patients reach ASCT + 100 days

- Progression free survival, ASCT + 100 days
- Treatment deliverability
- Minimum residual disease

# Patient Pathways



# Patient Pathways

Group	Scenario				n	
A1	Deliverable (ASCT) = YES	Progression-free at 100 days post-ASCT	Progression-free at 18 months post-randomisation	MRD +ve	14	
A2				MRD -ve	29	
A3				MRD unknown	30	
A4			Progressed or died by 18 months post-randomisation		MRD +ve	15
A5					MRD -ve	18
A6					MRD unknown	19
A7			Progressed/died by 100 days post-ASCT			10
A8	Deliverable (ASCT) = NO	Progression-free at 100 days post-ASCT	Progression-free at 18 months post-randomisation	MRD +ve	15	
A9				MRD -ve	0	
A10				MRD unknown	0	
A11			Progressed or died by 18 months post-randomisation		MRD +ve	13
A12					MRD -ve	0
A13					MRD unknown	0
A14			Progressed/died by 100 days post-ASCT			50

# The Model

- $A_1, \dots, A_{14}$  possible pathways with probabilities  $\theta_1, \dots, \theta_{14}$  and outcomes
- Historic control data  $\theta_S \sim Dir(a_S)$
- Prior for Experimental  $\theta_E \sim Dir(a_E)$
- $X_n = (X_{n,1}, \dots, X_{n,14})$  follows multinomial distribution in  $n$  (patients at analysis) and  $\theta_E$
- $\theta_E | X_n \sim Dir(a_{1E} + X_{n,1}, \dots, a_{14E} + X_{n,14})$

# Compound events

- Objective is to monitor clinically important events
- $C = A_{j_1} \cup \dots \cup A_{j_r}$  e.g. deliverability ( $A_1 \cup \dots \cup A_7$ )
- Can be shown that  $\Pr(C) [\tau]$  follows a beta distribution
- $\tau_S \sim \text{Beta}(a_1 + \dots + a_7, a_7 + \dots + a_{14})$

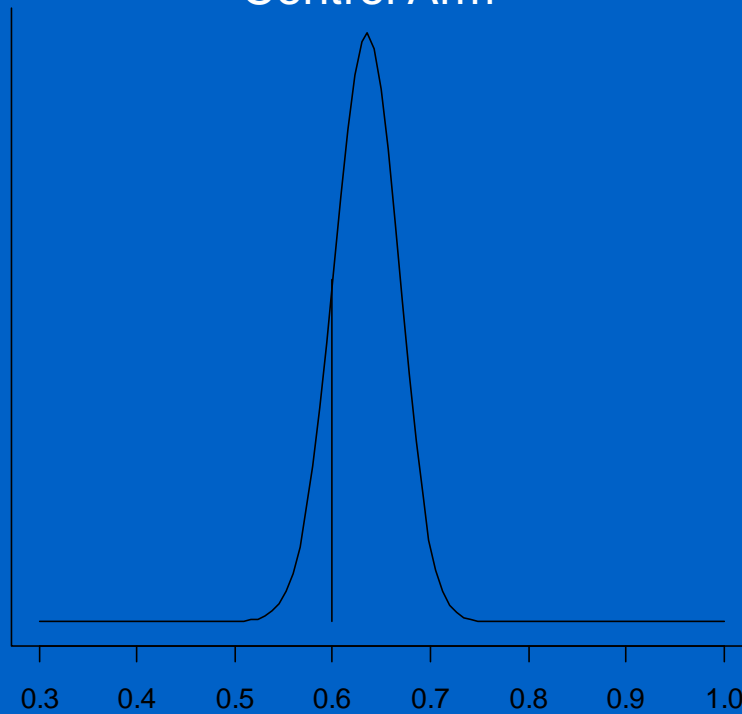
Monitoring Criterion is posterior probability

$$\Pr[\tau_S + \delta < \tau_E | X_n]$$

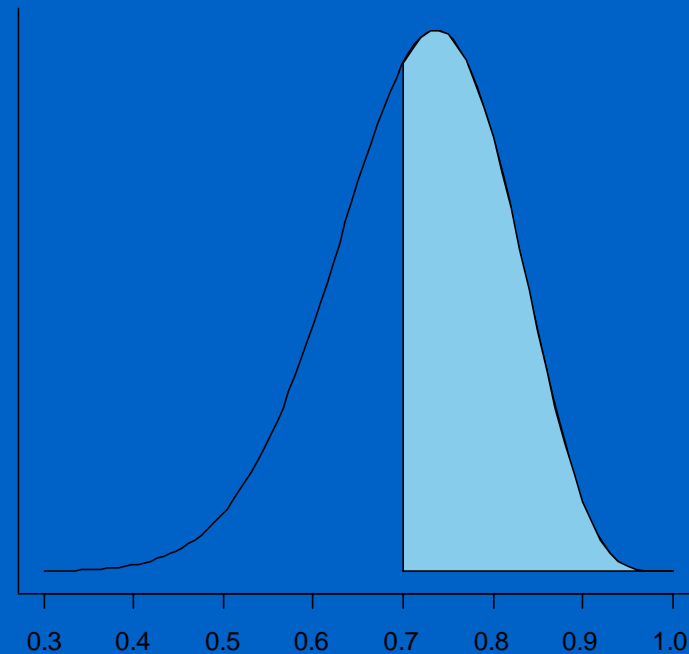
# Monitoring criterion

$$\Pr[\tau_S + \delta < \tau_E | X_n]$$

Control Arm



Experimental Arm



$$\int_0^{1-\delta} \{1 - B_{\tau_E}(p + \delta)\} b_{\tau_S}(p) dp$$

# Stopping criteria - MUK9

## Interim analyses

- $P(\text{Non-deliverability} > \text{control rate} + 20\%) > 0.9$
- $P(\text{Proportion progressed/died at 100 days post-ASCT} > \text{control rate}) > 0.9$
- $P(\text{MRD -ve rate at 100 days post-ASCT} > \text{control rate} + 10\%) < 0.05$

## Final analyses

- $P(\text{Proportion alive and progression-free at 18 months post-registration} > \text{control rate}) < 0.85$



# Simulations

- Simulations used to determine the operating characteristics of the design
- Assess the trials viability and tune the stopping boundaries
- Simulate at fixed value  $\mu_S = \mu_E$  (null) and alternative to be some  $\mu_E$
- Assess the early stopping probability ( $\pi$ )
- Under null  $1 - \pi$  considered type I error
- Under alternative  $\pi$  is type II error

# Myeloma UK Clinical Trial Network



# Discussion Points

- Updating the control arm part way through recruitment
- How to define simulation scenarios to demonstrate good operating characteristics
- The use of a different progression time point at interim and final analyses
- 120 patients is fixed, patients will be enrolled to single arm if one arm is dropped
- Comparison of two arms at final analysis