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Designing a pilot study using adaptive DP-optimality

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Context - health

- Low molecular weight heparin drugs (e.g. enoxaparin) are used acutely in the treatment of various cardiovascular diseases
 - For example, heart attacks, deep vein thrombosis
- They act by reducing blood clot formation which encourages clots to resolve
- Too much \rightarrow bleeding
- Too little \rightarrow further clots
- A test is needed to assess bleeding/clotting risk so we can get the dose correct

Clotting is complex and complicated



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A clotting time test

- The test is based on the notion that if you reduce the ability of the blood to form a clot then the time to clot (clotting time) will be prolonged.
- The degree of prolongation will determine the therapeutic or toxic potential of the anticoagulant (enoxaparin).
 - A clotting time of 10-20 seconds is normal.
 - A clotting time of 40 to 80 seconds is therapeutic
 - A clotting time of > 80 seconds may result in bleeding
- Note for a patient with a clotting disorder a normal clotting time may result in further clot formation.

Development of a clotting test



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Why the pilot?

- The proof-of-mechanism supported the thesis that we postulated for setting up a test,
- However the curse of heterogeneity resulted in the test not performing well in subsequent individuals
 - Significant variability in clotting time responses between individuals
- A pilot study was needed to determine the experimental conditions under which its performance would be acceptable across a wider range of individuals
- Two factors {Xa, Z} were considered as design variables.

The factors

- Xa is used to cause an *in-vitro* clot.
 - Too little and the clot doesn't occur
 - Too much and the clot occurs too fast and is insensitive to enoxaparin
- Z is used to reduce variability of the clotting time response for the *in-vitro* clot.
 - Too little and it doesn't work
 - Too much and Z causes a clot to occur of its own accord.

The pilot experiment

- A blood sample is taken from a subject
- The sample is split into different vials and 4 known amounts of enoxaparin are added to provide concentrations of {0, 0.25, 0.5, 1.0} U/mL
- The two factors of interest are added (Xa & Z)
- The time for the blood to form a clot is measured (= clotting time)
- Only a single combination of Xa and Z to be used per subject
- We can only study 6 subjects for the pilot

Methods for choosing values of Xa & Z

- Minimax design
 - Choose values that minimize the maximal variance in the experimental cases
- Factorial design
 - Consider a design that factors across all reasonable values
- Adaptive design
 - Consider a method to determine the probability of success over any values of Xa, Z by modelling the response surface

The response surface

- The response surface is described by a parametric function with unknown parameter vector (β)
- A surface will be determined for each of the 4 concentrations of the anticoagulant enoxaparin {0, 0.25, 0.5, 1}
- From these 4 surfaces a utility function can be applied that determines the overall success or failure of the test

The adaptive design

- The parameters are estimated using an adaptive design
- This method is linked to D-optimality for parameter estimation.
- It is also important for the pilot study to provide some empirical evidence of the utility of the test and for this purpose we used P-optimality (to optimise the probability of experimental success)
- A utility function was used to incorporate experimental success

Aim

• To conduct a pilot study with 6 subjects to determine values of {Xa, Z} to take forward into a proof-of-concept study

Response surface

$$CT = \left(\frac{f_Z^2}{f_E^5}a\right) f_{Xa}^{\ b}$$

$$f_{Xa} = 1 - \frac{[Xa]}{\left(\frac{f_E}{f_Z} K_{DXa}\right) + [Xa]}$$

The response surface (CT) over values of Xa and Z is determined for each concentration value of enoxaparin {0, 0.25, 0.5, 1}

$$f_E = 1 - \frac{[enoxaparin]}{K_{DE} + [enoxaparin]}$$

$$f_Z = 1 - \frac{[Z]}{K_{DZ} + [Z]}$$

Parameters,
$$\boldsymbol{\beta} = \{a, b, K_{DE}, K_{DZ}\}$$

Formulating a utility from a set of response surfaces

- The response surface will be determined for each of the 4 concentrations of enoxaparin.
- The response surface is characterised by a particular clotting time over a range of values of Xa and Z.
- Values of the clotting time that are desirable are then determined so that a utility can be formed.

High utility = appropriate clotting times

Defining success

- 1. (10s > CT > 30s) | ([enoxaparin] = 0)
- 2. (CT fold prolongation is > 1.3) | ([enoxaparin] = 0.25)
- 3. (CT fold prolongation > 2 fold more than CT
 ([enoxaparin] = 0.25)) | ([enoxaparin] = 1)
- 4. (CT fold prolongation is < 5) | ([enoxaparin] = 1)
- 5. All CT values must be < 200s

Success = 1 when 1-5 are true

Visual assessment of conditions



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DP-optimality

Design variables,
$$\boldsymbol{\xi} = \{Xa, Z\}$$

Parameters, $\boldsymbol{\beta} = \{a, b, K_{DE}, K_{DZ}\}$ $\hat{\boldsymbol{\xi}} = \arg \max_{\boldsymbol{\xi} \in \Xi} \left(\left| M_F(\hat{\boldsymbol{\beta}}, \boldsymbol{\xi}, [enoxaparin]) \right|^{\frac{1}{p}} U \right)$

p = number of parameters

 $U = \begin{cases} 0.9, \text{ when } CT \text{ success} = y \\ 0.1, \text{ when } CT \text{ success} = n \end{cases}$



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A simulation study

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Methods

- The performance of the design were assessed by simulation
- 100 virtual trials (of 6 patients) were simulated and analysed as per the pilot study protocol
- Assessment of the precision and accuracy of parameter values were considered for both D and DP optimal designs

Results

- Design performance for success
 - D-optimal designs \rightarrow success=y on 24% of simulations
 - DP-optimal design \rightarrow success=y on 94% of simulations
- Parameter values were estimated accurately by the 5th subject
- Not all parameter values could be estimated for the first few subjects



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The Actual Study

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Pilot study results Adaptive Design

Iteration	Subjects	Design points		Parameter values			
	(pooled data)	Xa	Ζ	a	K _{DE}	b	K _{DZ}
1	1	15	0.80	31	-	-	-
2	1,2	15	0.80	31	-	-	-
3	1,2,3	15	0.80	42	4.2	-	-
4	1,2,3,4	15	0.80	74	3.2	10	-
5	1,2,3,4,5	15	3.3	52	3.1	11	25
6	1,2,3,4,5,6	15	0.90	60	3.1	11	7.1

Clotting time response surfaces $E_{\text{TatEnovel}} = \{0, 0.25, 0.5, 1\}$



CT at Enox=0.5

CT at Enox=1



Predicted Probability of Success



Discussion

- An adaptive design using DP-optimality provided a useful method for designing the pilot study
- Fewer subjects were required than for a factorial design
- We have assumed that the parametric function provides a good description of the response surface this was based on our understanding of the clotting system
- We have ranges of {Xa, Z} that are likely to provide a successful proof-of-concept study (just completed)

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