

# APTS Design of Experiments - Practical 2

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In this practical session, you will use the `acebayes` package to investigate some aspects of (Bayesian) optimal design. Feel free to also use the practical sessions to understand and reproduce results from the notes.

## 1. Bayesian $D$ -optimal design using `optim`

Adapt the code from the notes to find locally  $D$ -optimal designs with  $n = 2$  runs for the Poisson regression model with linear predictor  $\eta(x) = \beta_0 + \beta_1 x$  with the log link (log-linear regression).

Fix  $\beta_0 = 0$  and find designs for  $\beta_1 = 1, \dots, 6$ , restricting to  $-1 \leq x \leq 1$ . For each design, plot the design points along with the expected response. There is a pattern in the optimal designs, can you identify it?

(Hint: there is a relationship between the first design point and  $1/\beta_1$ ).

The `acebayes` package has a function `aceglm` that can find Bayesian optimal designs for generalised linear models. Read the help file, and use the function to find a Bayesian  $D$ -optimal design when we assume  $\beta_0 = 0$  and a uniform prior distribution for  $\beta_1$  with support  $[1, 6]$ .

**Important:** when using `aceglm` set `N2 = 0` (otherwise it will result in a nasty crash!).

## 2. Bayesian $D$ -optimal design using `acebayes`

In the notes, we examine a compartmental model. A simplified version of this model is given by

$$y_i \sim N(\mu(x_i; \boldsymbol{\theta}), \sigma^2),$$

with

$$\mu(x_i; \boldsymbol{\theta}) = \theta_3 [\exp(-\theta_1 x) - \exp(-\theta_2 x)].$$

Use the `acenlm` function in `acebayes` to find a Bayesian  $D$ -optimal design with  $n = 18$  runs, assuming  $x \in [0, 24]$ . Assume uniform prior distributions for  $\theta_1, \theta_2$ :

$$\theta_1 \sim U(0.01884, 0.09884); \quad \theta_2 \sim U(0.298, 8.298)$$

with  $\theta_3 = 21.8$ ; see Gotwalt, Jones, and Steinberg (2009) and Overstall, Woods, and Adamou (2017). Plot the optimal design and realisations from the model.

# References

Gotwalt, C. M., B. A. Jones, and D. M. Steinberg. 2009. "Fast Computation of Designs Robust to Parameter Uncertainty for Nonlinear Settings." *Technometrics* 51: 88–95.

Overstall, A. M., D. C. Woods, and M. Adamou. 2017. "acebayes: An R Package for Bayesian Optimal Design of Experiments via Approximate Coordinate Exchange." *arXiv:1705.08096*.