Assessing the Safety of Rosiglitazone for the Treatment of Type II Diabetes A Bayesian Approach

THE LONDON SCHOOL OF ECONOMICS AND

POLITICAL SCIENCE

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How to assess a drug?

Subject ID	Treatment Group	Nausea	Dyspesia	Haemoglobin
233	1	1	1	-3.2
342	2	0	1	-4.7
213	1	1	0	1.4

Figure 1: Sample clinical trial dataset. Decisions about drugs are based on summaries of datasets like this.

How to Assess a Drug?

Regulatory Timeline:

- 1999 Rosiglitazone gets US approval
- 2000 Rosiglitazone gets European approval
- 2007 New evidence for risks arises [Nissen and Wolski, 2007]
- 2010 European regulators revert their recommendation
- 2011 US regulators partially revert their recommendation
- 2013 US regulators undo reversion

How to Assess a Drug?

Average Treatment Effects

Treatment Group	Nausea	Dyspesia	Hemoglobin
1	12%	8%	-0.5
2	13%	10%	-4.1

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- 2 Characterize the distribution of effects
- Need to combine effects into a single value s, termed preference $score^*$
- 4 $P(s^T > s^C | Y)$, generalize from the trial to the population

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Multi-criteria Decision Analysis, in a single slide

- "Standardize" each measured variable to a common scale [worst measurement,..., best measurement] \rightarrow [0,..., 100]
- 2 Choose appropriate weights w_i for each effect
- Construct the weighted sum $S = \sum_{j} w_{j} \cdot c_{j}$, called *preference score*
- 4 Compare the preference scores s^C and s^T for the control and treatment groups, respectively

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Issues/Goals

[Phillips et al., 2013] proposed using MCDA for drug assessment.

Issues:

- assumes independence of effects (correlation of mixed data)
- individual variability

Goals:

- Calculate $P(s_{N+1}^T > s_{N+1}^C | y)$ taking into account
- address above issues

For binary data:

$$\begin{cases} Y_{ij} \sim \text{Bernoulli}(\eta_j), \ i = 1, \dots, N, \ Y_{ij} \text{ independent, for fixed } j \\ h_j(\eta_j) = \mu_j + Z_{ij}, \end{cases}$$

$$(h_j(\eta_j) = \mu_j + Z_{ij})$$

For continuous variables:

$$Y_{ij} = \mu_j + Z_{ij}, \ i = 1, \dots, N.$$

where the distribution of
$$Z$$
 is assumed to be

$$Z_{i:}$$
 are independent $\forall i$

 $Z_{i\cdot} \sim \mathcal{N}_I(0_I, \Sigma),$

(3)

Model Objectives

The aim is to sample from

$$\pi(\mu, \Sigma, Z|Y) \propto f(Y|Z, \mu, \Sigma)\pi(Z|\Sigma)\pi(\mu)\pi(\Sigma)$$
 (4)

(for control and treatment groups) so that we can in turn sample from the score posterior.

We can then

- 1 compute $P(s^T > s^C | y)$ as before
- 2 compute $P(s_{N+1}^T > s_{N+1}^C | y)$ for a future individual N+1 based on

$$\pi(Z_{N+1}|y) = \int \int \pi(Z_{N+1:}|\mu, \Sigma) \pi(\mu, \Sigma|y) \ d\mu \ d\Sigma$$

3 compute $P(s_{N+1}^T > s_{N+1}^C | y, \widehat{\mu}, \widehat{\Sigma})$ based on Bayes or ML estimators of $\widehat{\mu}, \widehat{\Sigma}$.

How to Assess Rosiglitazone?

Using a clinical trial dataset

- 150 subjects in Control
- 152 subjects in Treatment (Rosiglitazone)

Testing for 6 effects:

- \bullet diarrhea (8.9)
- nausea (17.8)
- dyspepsia (1.8)
- \bullet edema (0.5)
- hemoglobin levels (59.2)
- glucose levels (11.8)

MCMC Samples

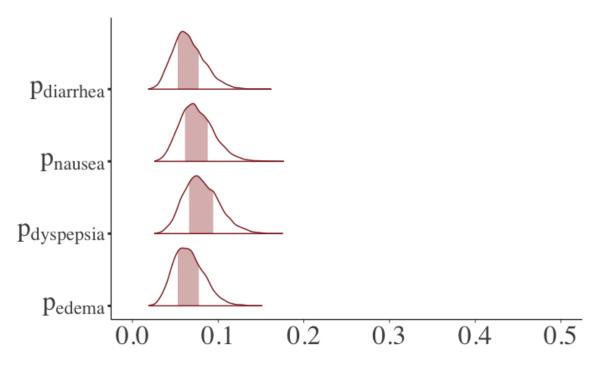


Figure 2: Probability of experiencing an effects for 4 binary effects.

MCMC Samples

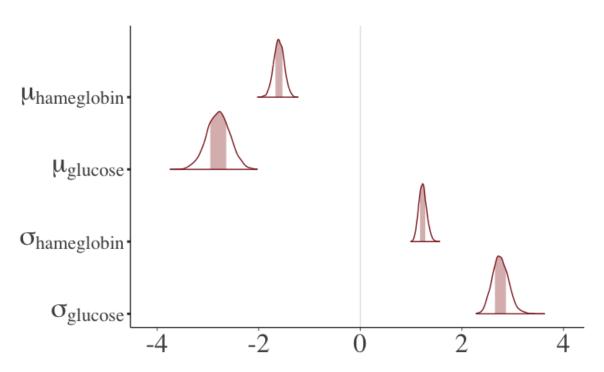


Figure 3: Mean and standard deviations for the 2 continuous effects.

MCMC Samples

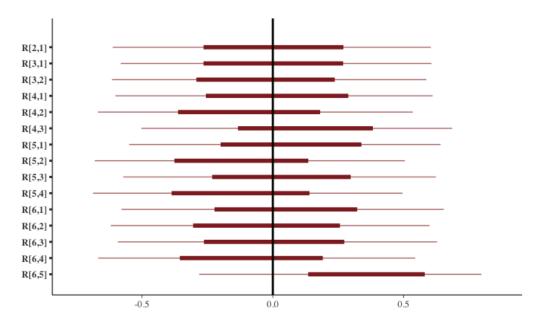


Figure 4: Pairwise correlation coefficient for all 6 effects.

Drug Preference Score

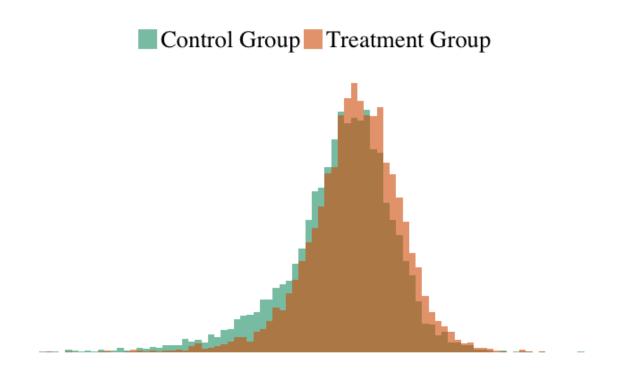
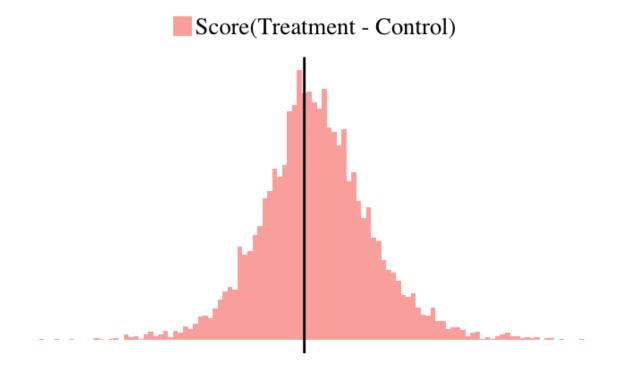


Figure 5: Final Score distribution s_{N+1}^T and s_{N+1}^C for a new individual taking the treatment (Rosiglitazone), and the placebo respectively.

Final Assessment

$$P(s_{N+1}^{\text{RSG}} > s_{N+1}^{C}) = 62 \%$$



Thank you!

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 github.com/bayesways
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Imi work package 5: Report 2:b:ii benefit - risk wave 2 case study report: Rosiglitazone.