COVARIATE SELECTION USING PENALISED REGRESSION ANALYSIS IN MILL MUD DATA ANALYSIS



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ABSTRACT

Cane yield data from an experiment designed to investigate the application of mill mud-ash treatments were analysed together with soils apparent electrical conductivity (ECa) measured at four soil depths. The apparent soil electrical conductivity measurements are highly dependent on one another and were used as covariates in the analysis. The variance inflation factor (VIF) indicates that there is multi-collinearity amongst these covariates. A penalised regression approach using lasso with SBC as the tuning method was used to select soil ECa covariates that are most relevant to cane yield. A linear mixed model was then fitted to the data with treatment and selected covariates from the lasso regression as fixed effects and replicate as the random effect. There was significant difference in treatment effects for cane yield (t/ha).

INTRODUCTION

Mill mud is the material remaining after cane juice is clarified and filtered in combination with other beneficial by-products like ash from sugarcane mills. It is a valuable soil conditioner and an important source of plant nutrients. Sub-surface application of mill mud is being trialled to improve productivity in the Herbert region of Queensland, Australia. For precise and variable application rate of mill mud to farm plots soil apparent electrical conductivity (ECa) mapping is needed. Soil ECa mapping allows farm block fertility management zones to be defined and create variable nutrient application rates. Due to changes in technology, soil ECa measurements are moving from a two-level soil depth to four-level soil depth measurements. These measurements are highly correlated. Fitting all the covariates in the model could lead to biased coefficient estimation, high standard errors and may result in incorrect conclusions about the relationship between outcome and predictor variables. Therefore, an appropriate methodology to account for collinearity while fitting these covariates in the analytical model should be used.

OBJECTIVE

- Aim is to assess the ECa covariates for collinearity, explore penalised regression (LASSO) for mitigating the collinearity and for feature selection.
- Compare mill mud treatment means for cane yield tonnes/ha to determine the best treatment.

METHOD

- Data from a sugarcane trial from Herbert region, Queensland, Australia.
- Seven treatments were evaluated:
 - Control = No mill mud/ash applied
 - ABa50 = Ash Banded 50 t/ha
- ABa100 = Ash Banded 100 t/ha
- ABr200 = Ash Broadcast 200 t/ha
- MBa50 = Mill Mud Banded 50 t/ha
- MBa100 = Mill Mud Banded 100 t/ha
- MBr200 = Mill Mud Broadcast 200 t/ha
- Covariates: apparent ECa measured at the soil depths of 0.5, 1.0, 1.6 and 3.2m.
- Covariates were accessed for collinearity by their variance inflation factor (VIF).
- *If > 10 there is collinearity*
- Penalised regression approach using LASSO (Tibshirani, 1996), with SBC as the tuning method was applied to the data for variable selection, using PROC GLMSELECT of SAS Analytic software (SAS Institute, 2013).

 $\beta^{\text{LASSO}}(\lambda) = \text{arg imin}_{-\beta}(||Y - X\beta||^2 + \lambda ||\beta||/_1)$

where λ is a tuning parameter and $\|.\|/_1$ stands for the vector $/_1$ -norm.

- This approach placed a constraint on the size of the regression coefficients in the model by shrinking them toward zero.
- Variables with a regression coefficient equal to zero after the shrinkage process were excluded from the model.
- Covariates with a non-zero coefficients were the most strongly associated with the response variable and were kept in the model.
- The selected variables were then used to fit a mixed model to access the effect of treatment on cane yield.
- Treatments were considered as fixed, replicate as random effects and the selected soil EC as covariates.

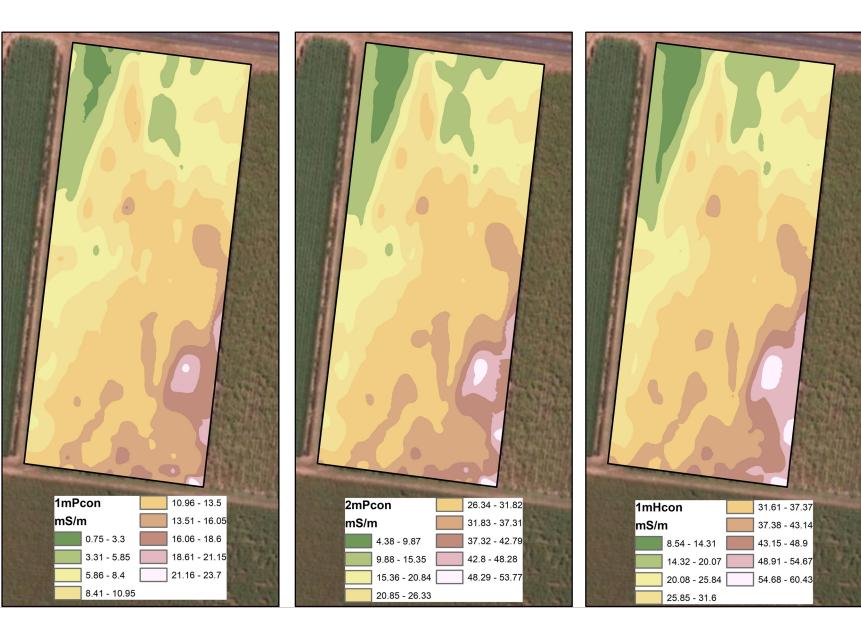


Fig1. Example of soil apparent electricity conductivity (ECa) measurement at soil depth 0.6m (1mPcon), 1.2m (2mPcon) and 1.5m (1mHcon)

RESULTS

- Covariates were highly correlated with r between 0.96 to 0.99.
- VIF for all the variables were > 10 (412 to 2029)
- Soil ECa measurement at soil depths of 0.5 and 1.0
 m were the most strongly associated covariates with
 cane sugar yield

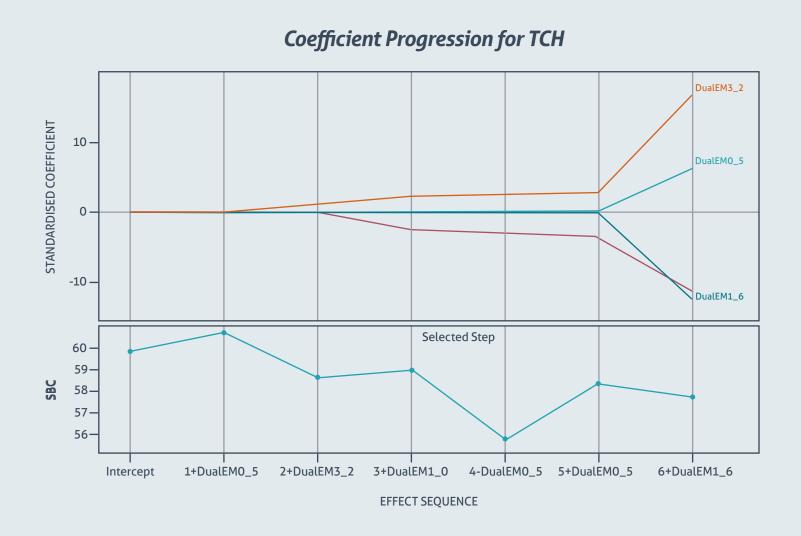


Fig2a. Coefficient progression and model selection step, each curve represents a coefficient as labelled

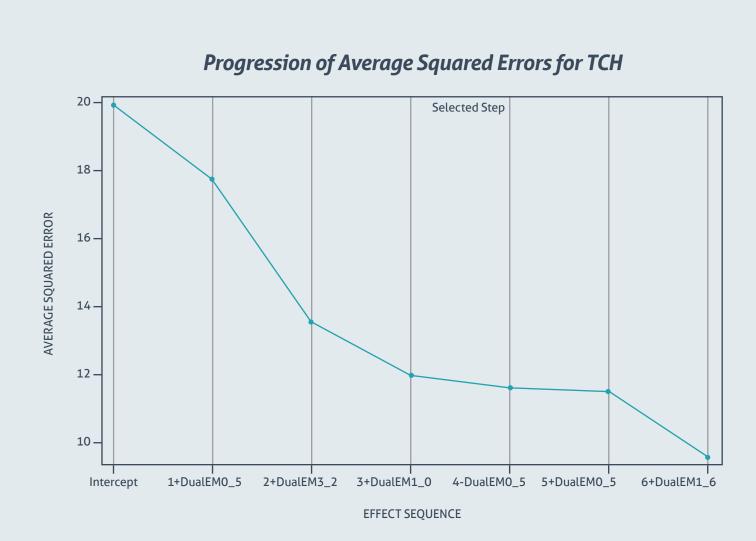


Fig2b. Progression of average square errors for cane yield showing selected covariates

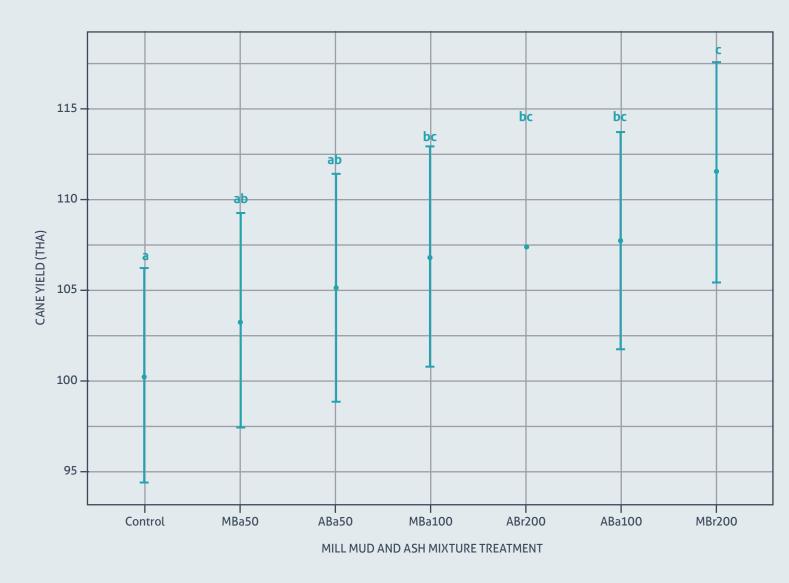


Figure 3. Predicted cane yield t/ha, error bars are confidence interval at 95% level

CONCLUSION

We were able to use LASSO regression to select the most relevant Eac covariates for our response variable, cane yield. The ash/mill mud application had significant impact on cane yield, with greater yield (tonnes/ha) from the mil mud broadcast at 200t/ha (MBr2000) than the banded application of either mill mud or ash at 50t/ha.

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