

# Non-Monotone Missingness: The Minimal Variance of the Linear Estimate

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## Introduction

The goal of this project is to get an optimal (or near optimal) estimator of a general estimator when the data has a non-monotone missing structure. So far we have reduced the problem to estimating the following:

$$\hat{\theta} = \frac{\delta_{11}}{\pi_{11}}g(Z) + \lambda_0(\delta)\alpha_0(X) + \lambda_1(\delta)\alpha_1(X, Y_1) + \lambda_2(\delta)\alpha_2(X, Y_2). \quad (1)$$

For notation we have variables  $Z = (X, Y_1, Y_2)$  that are observed in the segment noted in Table 1.

Table 1: This table matches the segment with its associated observed variables.

| Segment  | Variables     |
|----------|---------------|
| $A_{00}$ | $X$           |
| $A_{10}$ | $X, Y_1$      |
| $A_{01}$ | $X, Y_2$      |
| $A_{11}$ | $X, Y_1, Y_2$ |

For each segment  $A_{d_1 d_2}$  the probability of an observation being in a specific segment is known to be  $\pi_{d_1 d_2}$ , and the associated random variable that indicates whether observation  $i$  is in  $A_{d_1 d_2}$  is  $\delta_{d_1 d_2 i}$ . For brevity, it will sometimes be convenient to drop the subscript  $i$ . To get an optimal estimator, we need to choose values of  $\lambda$  and  $\alpha$  to minimize the variance of  $\hat{\theta}$ .

## Simplifications

To simplify this model consider the case in which each  $\lambda$  and  $\alpha$  are linear functions. This means that we have the following:  $\lambda_0 = \lambda_0^{(0)} + \lambda_1^{(0)}\delta_{00} + \lambda_2^{(0)}\delta_{10} + \lambda_3^{(0)}\delta_{01} + \lambda_4^{(0)}\delta_{11}$ ,  $\lambda_1 = \lambda_0^{(1)} + \lambda_2^{(1)}\delta_{10} + \lambda_3^{(1)}\delta_{11}$ , and  $\lambda_2 = \lambda_0^{(2)} + \lambda_3^{(2)}\delta_{01} + \lambda_4^{(2)}\delta_{11}$ . Also,  $\alpha_0 = \alpha_0^{(0)} + \alpha_1^{(0)}x$ ,  $\alpha_1 = \alpha_0^{(1)} + \alpha_1^{(1)}x + \alpha_2^{(1)}y_1$ , and  $\alpha_2 = \alpha_0^{(2)} + \alpha_1^{(1)}x + \alpha_3^{(1)}y_2$ . This means that we can express the estimator in Equation 1 as the following matrix equation:

$$\hat{\theta} = n^{-1} \mathbf{1}'_n \left( \frac{\delta_{11}}{\pi_{11}} g(Z) + (\boldsymbol{\alpha} \boldsymbol{\lambda}' \boldsymbol{\delta})' \mathbf{Z} \right).$$

where

$$\boldsymbol{\alpha} = \begin{bmatrix} \alpha_0^{(0)} & \alpha_0^{(1)} & \alpha_0^{(2)} \\ \alpha_1^{(0)} & \alpha_1^{(1)} & \alpha_1^{(2)} \\ 0 & \alpha_2^{(1)} & 0 \\ 0 & 0 & \alpha_3^{(2)} \end{bmatrix}, \boldsymbol{\lambda} = \begin{bmatrix} \lambda_0^{(0)} & \lambda_0^{(1)} & \lambda_0^{(2)} \\ \lambda_1^{(0)} & 0 & 0 \\ \lambda_2^{(0)} & \lambda_2^{(1)} & 0 \\ \lambda_3^{(0)} & 0 & \lambda_3^{(2)} \\ \lambda_4^{(0)} & \lambda_4^{(1)} & \lambda_4^{(2)} \end{bmatrix}, \boldsymbol{\delta} = \begin{bmatrix} 1 \\ \delta_{00} \\ \delta_{10} \\ \delta_{01} \\ \delta_{11} \end{bmatrix} \mathbf{Z}_i = \begin{bmatrix} 1 \\ X_i \\ Y_{1i} \\ Y_{2i} \end{bmatrix}$$

This is a linear estimator and the goal is to find the optimal  $\boldsymbol{\alpha}$  and  $\boldsymbol{\lambda}$  that,

$$\text{minimize } \text{Var}(\hat{\theta}) \text{ such that } \boldsymbol{\lambda}' E[\boldsymbol{\delta}] = 0.$$

## Simplifying the Variance

We know that

$$\begin{aligned} \text{Var} \left( \frac{\delta_{11}}{\pi_{11}} g(Z) + (\boldsymbol{\alpha} \boldsymbol{\lambda}' \boldsymbol{\delta})' \mathbf{Z} \right) &= \text{Var} \left( \frac{\delta_{11}}{\pi_{11}} g(Z) \right) + \text{Var}((\boldsymbol{\alpha} \boldsymbol{\lambda}' \boldsymbol{\delta})' \mathbf{Z}) + 2\text{Cov} \left( \frac{\delta_{11}}{\pi_{11}} g(Z), (\boldsymbol{\alpha} \boldsymbol{\lambda}' \boldsymbol{\delta})' \mathbf{Z} \right) \\ &:= V_1 + V_2 + V_3 \end{aligned}$$

Hence, we have for  $\Pi := E[\boldsymbol{\delta}]$ ,

$$\begin{aligned}
V_1 &= \text{Var} \left( \frac{\delta_{11}}{\pi_{11}} g(Z) \right) \\
&= \text{Var} \left( E \left[ \frac{\delta_{11}}{\pi_{11}} g(Z) \mid Z \right] \right) + E \left[ \text{Var} \left( \frac{\delta_{11}}{\pi_{11}} g(Z) \mid Z \right) \right] \\
&= E[g^2(Z)] - E[g(Z)]^2 + \frac{1}{\pi_{11}} E[g^2(Z)] - E[g^2(Z)] \\
&= \frac{1}{\pi_{11}} E[g^2(Z)] - E[g(Z)]^2
\end{aligned}$$

$$\begin{aligned}
V_2 &= \text{Var}((\alpha\lambda'\delta)'Z) \\
&= \text{Var}(E[(\alpha\lambda'\delta)'Z \mid Z]) + E[\text{Var}((\alpha\lambda'\delta)'Z \mid Z)] \\
&= \Pi'\lambda\alpha'\text{Var}(Z)\alpha\lambda'\Pi + E[Z'\alpha\lambda'\text{Var}(\delta)\lambda\alpha'Z] \\
&= \Pi'\lambda\alpha'(E[ZZ'] - E[Z]E[Z'])\alpha\lambda'\Pi + E[Z'\alpha\lambda'(E[\delta\delta'] - \Pi\Pi')\lambda\alpha'Z] \\
&= \Pi'\lambda\alpha'E[ZZ']\alpha\lambda'\Pi - \Pi'\lambda\alpha'E[Z]E[Z']\alpha\lambda'\Pi + E[Z'\alpha\lambda'E[\delta\delta']\lambda\alpha'Z] - E[Z'\alpha\lambda'\Pi\Pi'\lambda\alpha'Z] \\
&= \Pi'\lambda\alpha'E[ZZ']\alpha\lambda'\Pi - \Pi'\lambda\alpha'E[Z]E[Z']\alpha\lambda'\Pi + E[Z'\alpha\lambda'E[\delta\delta']\lambda\alpha'Z] - \Pi'\lambda\alpha'E[ZZ']\alpha\lambda'\Pi \\
&= \Pi'\lambda\alpha'E[Z]E[Z']\alpha\lambda'\Pi + E[Z'\alpha\lambda'E[\delta\delta']\lambda\alpha'Z] \\
&= E[Z'\alpha\lambda'E[\delta\delta']\lambda\alpha'Z]
\end{aligned}$$

The third to last equality holds because a  $1 \times 1$  matrix is always symmetric and the last equality holds because  $\Pi'\lambda = (\lambda'\Pi)' = 0'$ . We can expand this expression but we get the horribly ugly result:

$$\begin{aligned}
&X^2 (\alpha_1^{(0)})^2 (\lambda_0^{(0)})^2 + 2X^2 (\alpha_1^{(0)})^2 \lambda_0^{(0)} \lambda_1^{(0)} \pi_{00} + 2X^2 (\alpha_1^{(0)})^2 \lambda_0^{(0)} \lambda_2^{(0)} \pi_{10} + 2X^2 (\alpha_1^{(0)})^2 \lambda_0^{(0)} \lambda_3^{(0)} \pi_{01} + \\
&2X^2 (\alpha_1^{(0)})^2 \lambda_0^{(0)} \lambda_4^{(0)} \pi_{11} + X^2 (\alpha_1^{(0)})^2 (\lambda_1^{(0)})^2 \pi_{00} + X^2 (\alpha_1^{(0)})^2 (\lambda_2^{(0)})^2 \pi_{10} + X^2 (\alpha_1^{(0)})^2 (\lambda_3^{(0)})^2 \pi_{01} + \\
&X^2 (\alpha_1^{(0)})^2 (\lambda_4^{(0)})^2 \pi_{11} + 2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(0)} \lambda_0^{(1)} \pi_{10} + 2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(0)} \lambda_0^{(1)} + 2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(0)} \lambda_4^{(1)} \pi_{11} + \\
&2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(1)} \lambda_1^{(0)} \pi_{00} + 4X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(1)} \lambda_2^{(0)} \pi_{10} + 2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(1)} \lambda_3^{(0)} \pi_{01} + 2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_0^{(1)} \lambda_4^{(0)} \pi_{11} + \\
&2X^2 \alpha_1^{(0)} \alpha_1^{(1)} \lambda_4^{(0)} \lambda_4^{(1)} \pi_{11} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(0)} \lambda_0^{(2)} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(0)} \lambda_3^{(2)} \pi_{01} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(0)} \lambda_4^{(2)} \pi_{11} + \\
&2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(2)} \lambda_1^{(0)} \pi_{00} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(2)} \lambda_2^{(0)} \pi_{10} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(2)} \lambda_3^{(0)} \pi_{01} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_0^{(2)} \lambda_4^{(0)} \pi_{11} + \\
&2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_3^{(0)} \lambda_3^{(2)} \pi_{01} + 2X^2 \alpha_1^{(0)} \alpha_1^{(2)} \lambda_4^{(0)} \lambda_4^{(2)} \pi_{11} + 3X^2 (\alpha_1^{(1)})^2 (\lambda_0^{(1)})^2 \pi_{10} + X^2 (\alpha_1^{(1)})^2 (\lambda_0^{(1)})^2 + \\
&2X^2 (\alpha_1^{(1)})^2 \lambda_0^{(1)} \lambda_4^{(1)} \pi_{11} + X^2 (\alpha_1^{(1)})^2 (\lambda_4^{(1)})^2 \pi_{11} + 2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_0^{(1)} \lambda_0^{(2)} \pi_{10} + 2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_0^{(1)} \lambda_0^{(2)} + \\
&2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_0^{(1)} \lambda_3^{(2)} \pi_{01} + 2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_0^{(1)} \lambda_4^{(2)} \pi_{11} + 2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_0^{(1)} \lambda_4^{(1)} \pi_{11} + 2X^2 \alpha_1^{(1)} \alpha_1^{(2)} \lambda_4^{(1)} \lambda_4^{(2)} \pi_{11} + \\
&X^2 (\alpha_1^{(2)})^2 (\lambda_0^{(2)})^2 + 2X^2 (\alpha_1^{(2)})^2 \lambda_0^{(2)} \lambda_3^{(2)} \pi_{01} + 2X^2 (\alpha_1^{(2)})^2 \lambda_0^{(2)} \lambda_4^{(2)} \pi_{11} + X^2 (\alpha_1^{(2)})^2 (\lambda_3^{(2)})^2 \pi_{01} + \\
&X^2 (\alpha_1^{(2)})^2 (\lambda_4^{(2)})^2 \pi_{11} + 2XY_1 \alpha_1^{(0)} \alpha_2^{(1)} \lambda_0^{(0)} \lambda_0^{(1)} \pi_{10} + 2XY_1 \alpha_1^{(0)} \alpha_2^{(1)} \lambda_0^{(0)} \lambda_0^{(1)} + 2XY_1 \alpha_1^{(0)} \alpha_2^{(1)} \lambda_0^{(0)} \lambda_4^{(1)} \pi_{11} +
\end{aligned}$$



$$\begin{aligned}
& 2Y_1\alpha_0^{(2)}\alpha_2^{(1)}\lambda_0^{(1)}\lambda_3^{(2)}\pi_{01} + 2Y_1\alpha_0^{(2)}\alpha_2^{(1)}\lambda_0^{(1)}\lambda_4^{(2)}\pi_{11} + 2Y_1\alpha_0^{(2)}\alpha_2^{(1)}\lambda_0^{(2)}\lambda_4^{(1)}\pi_{11} + 2Y_1\alpha_0^{(2)}\alpha_2^{(1)}\lambda_4^{(1)}\lambda_4^{(2)}\pi_{11} + \\
& Y_2^2\left(\alpha_3^{(2)}\right)^2\left(\lambda_0^{(2)}\right)^2 + 2Y_2^2\left(\alpha_3^{(2)}\right)^2\lambda_0^{(2)}\lambda_3^{(2)}\pi_{01} + 2Y_2^2\left(\alpha_3^{(2)}\right)^2\lambda_0^{(2)}\lambda_4^{(2)}\pi_{11} + Y_2^2\left(\alpha_3^{(2)}\right)^2\left(\lambda_3^{(2)}\right)^2\pi_{01} + \\
& Y_2^2\left(\alpha_3^{(2)}\right)^2\left(\lambda_4^{(2)}\right)^2\pi_{11} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(0)}\lambda_0^{(2)} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(0)}\lambda_3^{(2)}\pi_{01} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(0)}\lambda_4^{(2)}\pi_{11} + \\
& 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_1^{(0)}\pi_{00} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_2^{(0)}\pi_{10} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_3^{(0)}\pi_{01} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_4^{(0)}\pi_{11} + \\
& 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_3^{(0)}\lambda_3^{(2)}\pi_{01} + 2Y_2\alpha_0^{(0)}\alpha_3^{(2)}\lambda_4^{(0)}\lambda_4^{(2)}\pi_{11} + 2Y_2\alpha_0^{(1)}\alpha_3^{(2)}\lambda_0^{(1)}\lambda_0^{(2)}\pi_{10} + 2Y_2\alpha_0^{(1)}\alpha_3^{(2)}\lambda_0^{(1)}\lambda_3^{(2)}\pi_{01} + \\
& 2Y_2\alpha_0^{(1)}\alpha_3^{(2)}\lambda_0^{(1)}\lambda_4^{(2)}\pi_{11} + 2Y_2\alpha_0^{(1)}\alpha_3^{(2)}\lambda_0^{(1)}\lambda_4^{(2)}\pi_{11} + 2Y_2\alpha_0^{(1)}\alpha_3^{(2)}\lambda_4^{(1)}\lambda_4^{(2)}\pi_{11} + \\
& 2Y_2\alpha_0^{(2)}\alpha_3^{(2)}\left(\lambda_0^{(2)}\right)^2 + 4Y_2\alpha_0^{(2)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_3^{(2)}\pi_{01} + 4Y_2\alpha_0^{(2)}\alpha_3^{(2)}\lambda_0^{(2)}\lambda_4^{(2)}\pi_{11} + 2Y_2\alpha_0^{(2)}\alpha_3^{(2)}\left(\lambda_3^{(2)}\right)^2\pi_{01} + \\
& 2Y_2\alpha_0^{(2)}\alpha_3^{(2)}\left(\lambda_4^{(2)}\right)^2\pi_{11} + \left(\alpha_0^{(0)}\right)^2\left(\lambda_0^{(0)}\right)^2 + 2\left(\alpha_0^{(0)}\right)^2\lambda_0^{(0)}\lambda_1^{(0)}\pi_{00} + 2\left(\alpha_0^{(0)}\right)^2\lambda_0^{(0)}\lambda_2^{(0)}\pi_{10} + \\
& 2\left(\alpha_0^{(0)}\right)^2\lambda_0^{(0)}\lambda_3^{(0)}\pi_{01} + 2\left(\alpha_0^{(0)}\right)^2\lambda_0^{(0)}\lambda_4^{(0)}\pi_{11} + \left(\alpha_0^{(0)}\right)^2\left(\lambda_1^{(0)}\right)^2\pi_{00} + \left(\alpha_0^{(0)}\right)^2\left(\lambda_2^{(0)}\right)^2\pi_{10} + \\
& \left(\alpha_0^{(0)}\right)^2\left(\lambda_3^{(0)}\right)^2\pi_{01} + \left(\alpha_0^{(0)}\right)^2\left(\lambda_4^{(0)}\right)^2\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(0)}\lambda_0^{(1)}\pi_{10} + 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(0)}\lambda_0^{(1)} + \\
& 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(0)}\lambda_4^{(1)}\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(1)}\lambda_1^{(0)}\pi_{00} + 4\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(1)}\lambda_2^{(0)}\pi_{10} + 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(1)}\lambda_3^{(0)}\pi_{01} + \\
& 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_0^{(1)}\lambda_4^{(0)}\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(1)}\lambda_4^{(0)}\lambda_4^{(1)}\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(0)}\lambda_0^{(2)} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(0)}\lambda_3^{(2)}\pi_{01} + \\
& 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(0)}\lambda_4^{(2)}\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(2)}\lambda_1^{(0)}\pi_{00} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(2)}\lambda_2^{(0)}\pi_{10} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(2)}\lambda_3^{(0)}\pi_{01} + \\
& 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_0^{(2)}\lambda_4^{(0)}\pi_{11} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_3^{(0)}\lambda_3^{(2)}\pi_{01} + 2\alpha_0^{(0)}\alpha_0^{(2)}\lambda_4^{(0)}\lambda_4^{(2)}\pi_{11} + 3\left(\alpha_0^{(1)}\right)^2\left(\lambda_0^{(1)}\right)^2\pi_{10} + \\
& \left(\alpha_0^{(1)}\right)^2\left(\lambda_0^{(1)}\right)^2 + 2\left(\alpha_0^{(1)}\right)^2\lambda_0^{(1)}\lambda_4^{(1)}\pi_{11} + \left(\alpha_0^{(1)}\right)^2\left(\lambda_4^{(1)}\right)^2\pi_{11} + 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_0^{(1)}\lambda_0^{(2)}\pi_{10} + \\
& 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_0^{(1)}\lambda_0^{(2)} + 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_0^{(1)}\lambda_3^{(2)}\pi_{01} + 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_0^{(1)}\lambda_4^{(2)}\pi_{11} + 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_0^{(2)}\lambda_4^{(1)}\pi_{11} + \\
& 2\alpha_0^{(1)}\alpha_0^{(2)}\lambda_4^{(1)}\lambda_4^{(2)}\pi_{11} + \left(\alpha_0^{(2)}\right)^2\left(\lambda_0^{(2)}\right)^2 + 2\left(\alpha_0^{(2)}\right)^2\lambda_0^{(2)}\lambda_3^{(2)}\pi_{01} + 2\left(\alpha_0^{(2)}\right)^2\lambda_0^{(2)}\lambda_4^{(2)}\pi_{11} + \\
& \left(\alpha_0^{(2)}\right)^2\left(\lambda_3^{(2)}\right)^2\pi_{01} + \left(\alpha_0^{(2)}\right)^2\left(\lambda_4^{(2)}\right)^2\pi_{11}
\end{aligned}$$

Finally, to understand  $V_3$  we can solve for the third covariance term.

$$\begin{aligned}
& \text{Cov}\left(\frac{\delta_{11}}{\pi_{11}}g(Z), (\alpha\lambda'\delta)'Z\right) \\
&= E\left[\frac{\delta_{11}}{\pi_{11}}g(Z)(\alpha\lambda'\delta)'Z\right] - E\left[\frac{\delta_{11}}{\pi_{11}}g(Z)\right]E[(\alpha\lambda'\delta)'Z] \\
&= E\left[g(Z)\frac{\delta_{11}}{\pi_{11}}\delta'\lambda\alpha'Z\right] - E[g(Z)](\alpha\lambda'\Pi)'E[Z] \\
&= E[g(Z)\begin{bmatrix} 1 & 0 & 0 & 0 & 1 \end{bmatrix}\lambda\alpha'Z] \\
&= (\lambda_0^{(0)} + \lambda_4^{(0)})(\alpha_0^{(0)} + \alpha_1^{(0)}E[g(Z)x]) + (\lambda_0^{(1)} + \lambda_4^{(1)})(\alpha_0^{(1)} + \alpha_1^{(1)}E[g(Z)x] + \alpha_2^{(1)}E[g(Z)y_1]) \\
&\quad + (\lambda_0^{(2)} + \lambda_4^{(2)})(\alpha_0^{(2)} + \alpha_1^{(2)}E[g(Z)x] + \alpha_3^{(2)}E[g(Z)y_2])
\end{aligned}$$

Since we have an understanding of  $\text{Var}(\hat{\theta})$ , we can find the minimum by differentiating with respect to each coefficient in  $\lambda$  and  $\alpha$ .

This is still a work in progress. I need to run a simulation and test it out. I also believe that the terms  $\alpha_0^{(0)}$ ,  $\alpha_0^{(1)}$ , and  $\alpha_0^{(2)}$  hinder identifiability. But I will work on it.