

$$J = 1, \dots, J+1 \quad t = 1, \dots, T_0, \dots, T$$

$Y_{it}^N$  = not treated outcome

$Y_{it}^I$  = treated outcome

↑ event/policy

$$\alpha_{1t} = Y_{1t}^I - \underbrace{Y_{1t}^N}_{\text{estimate! don't observe}} \quad \text{for } t = T_0+1, \dots, T$$

Find units w/ similar trends and characteristics!

we have  $Y_{it}$ ,  $t \leq T_0$  and  $\frac{U_i}{r \times 1} \rightarrow$  other covariates, summary statistics from pretreatment period.

$$\underline{K} = (k_1, \dots, k_{T_0})'$$

each  $k$  is a different linear combination of pretreatment outcomes

$$\bar{Y}_i^k = \sum_{s=1}^{T_0} k_s Y_{is}$$

$M \leq T_0$  possible different linear combinations of pretreatment outcomes

looking for optimal weights  $\underline{w} = (w_2, \dots, w_{J+1})'$

for now  $w_j \geq 0 \quad \sum w_j = 1$

select  $\underline{w}^*$  such that

$$\bar{Y}_1^{k_1} \approx \sum_{j=2}^{J+1} w_j^* \bar{Y}_j^{k_1} \quad \text{for each linear combination } k$$

$$U_1 \approx \sum_{j=2}^{J+1} w_j^* U_j$$

$$\text{Then } \hat{\alpha}_{1t} = Y_{1t} - \sum_{j=2}^{J+1} w_j^* Y_{jt}$$

Putting into practice:

$$\text{let } \underline{X}_1 = \left( \underline{U}_1', \bar{Y}_1^{k_1}, \dots, \bar{Y}_1^{k_M} \right)$$

$$\checkmark \quad \begin{bmatrix} \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

$(r+m) \times 1$

$$\tilde{X}_0 = \begin{bmatrix} \vdots & \vdots & \vdots \\ \underline{y}_j^T, \underline{y}_j^{R_1} & \dots & \underline{y}_j^{R_M} \\ \vdots & \vdots & \vdots \end{bmatrix}$$

Need to write  
your own optimization  
routine

$$\min_{\underline{W}} \sqrt{\underbrace{\left( \underline{x}_1 - \tilde{X}_0 \cdot \underline{W} \right)' \tilde{V} \left( \underline{x}_1 - \tilde{X}_0 \cdot \underline{W} \right)}_{(r+m) \times (r+m)}} + \text{LASSO / Ridge / ElasticNet penalties}$$

↪ inverse of variance of the predictors  
(of the matching variables)

① First solve for  $\underline{W}^*(\underline{V})$

$$\textcircled{2} \min_{\tilde{V}} \left( \underline{y}_1 - \tilde{X}_0 \cdot \underline{W}^*(\underline{V}) \right)' \left( \underline{y}_1 - \tilde{X}_0 \cdot \underline{W}^*(\underline{V}) \right) \quad \leftarrow \text{MSPE at } \underline{V}^*$$

③ plug in  $\underline{W}^*(\underline{V}^*)$  now we have optimal weights

Package : SCUL Synthetic Control Using LASSO

$$\min_{\mu, \underline{W}} \sum_{t=1}^{T_0} \left( y_{1t} - \mu - \sum_{j=2}^{J+1} w_j y_{jt} \right)^2 + \lambda \underbrace{\|w_j\|}_{\text{LASSO Penalty}}$$

$$\text{restricts } \underline{x}_1 = (y_{11}, y_{12}, \dots, y_{1T_0})$$

$$\tilde{V} = \underline{I}$$

$$\tilde{X}_0 = \begin{bmatrix} \vdots & \vdots \\ y_{j1}, \dots, y_{jT_0} \\ \vdots & \vdots \end{bmatrix}$$

$$\text{allows } \mu \neq 0, \quad w_j \geq 0, \quad \sum_{j=2}^{J+1} w_j \neq 1$$