

# Causation and difference-in-differences

# Raising the minimum wage

What happens if you raise the minimum wage?

Economic theory says there should be fewer jobs

New Jersey in 1992

\$4.25 → \$5.05

# Before vs. after

Average fast food jobs in NJ

Before: 20.44

After: 21.03

$\Delta$ : 0.59

Is this the causal effect?

# Treatment vs. control

Average fast food jobs in states

$PA_{\text{after}}$ : 21.17

$NJ_{\text{after}}$ : 21.03

$\Delta$ : -0.14

Is this the causal effect?

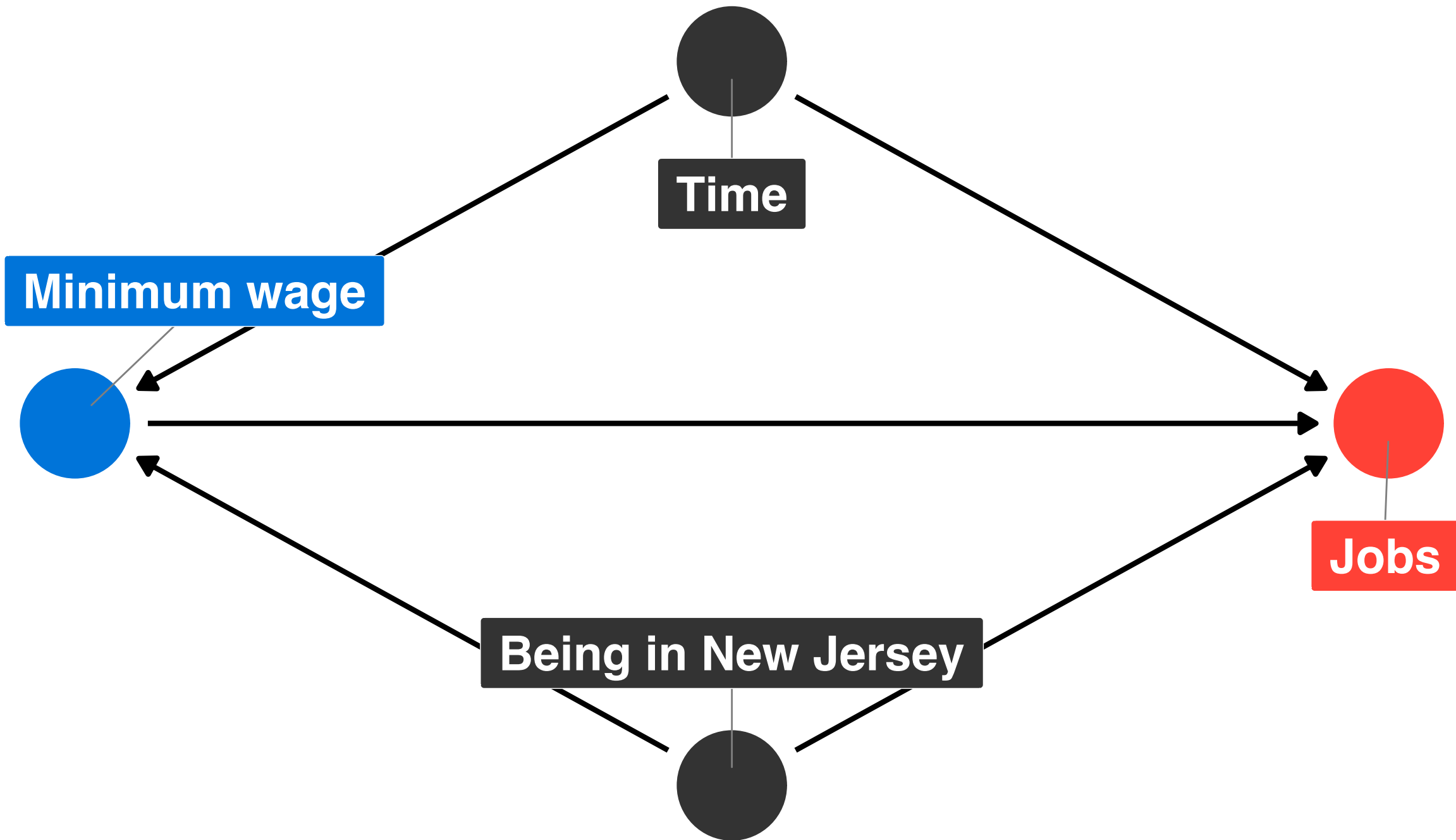
# Problems

**Comparing only before/after**

**Impossible to know if growth happened because of treatment or just naturally**

**Comparing only treatment/control**

**Impossible to know if any changes happened because of natural growth**



	<b>Pre mean</b>	<b>Post mean</b>
<b>Treatment</b>	<b>A</b> <b>(not yet treated)</b>	<b>B</b> <b>(treated)</b>
<b>Control</b>	<b>C</b> <b>(never treated)</b>	<b>D</b> <b>(never treated)</b>

	Pre mean	Post mean	$\Delta$ (post-pre)
Treatment	A (not yet treated)	B (treated)	B-A
Control	C (never treated)	D (never treated)	D-C

**Growth!**



	Pre mean	Post mean
Treatment	A (not yet treated)	B (treated)
Control	C (never treated)	D (never treated)
$\Delta$ (trtmt-ctrl)	A-C	B-D

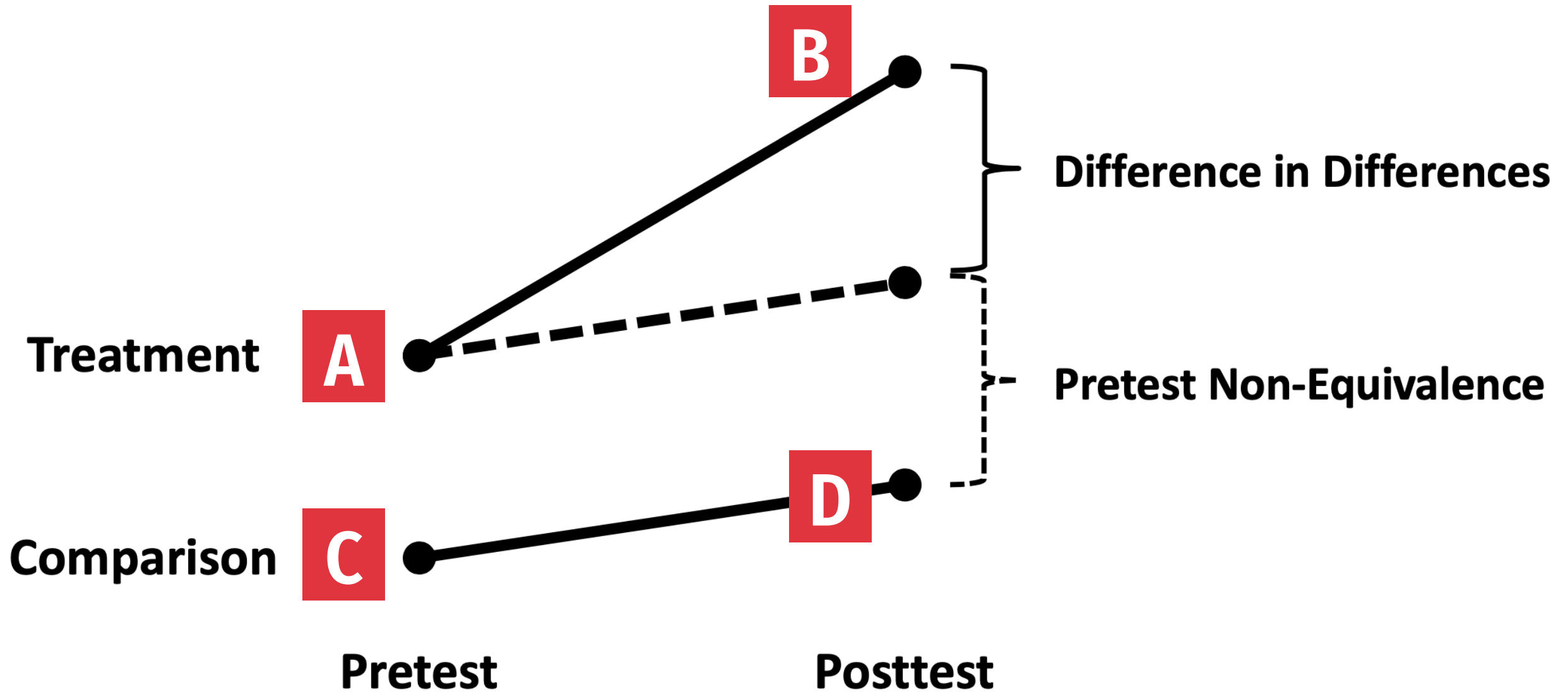
**Within-group effects**

	Pre mean	Post mean	$\Delta$ (post-pre)
Treatment	A (not yet treated)	B (treated)	B-A
Control	C (never treated)	D (never treated)	D-C
$\Delta$ (trtmt-ctrl)	A-C	B-D	(B-A) - (D-C)

**Growth of treatment –  
growth of control (DiD!)**

$$\text{DD} = \left( \bar{x}_{\text{treatment, post}} - \bar{x}_{\text{treatment, pre}} \right) - \left( \bar{x}_{\text{control, post}} - \bar{x}_{\text{control, pre}} \right)$$

	Pre mean	Post mean	$\Delta$ (post-pre)
NJ	A <b>20.44</b>	B <b>21.03</b>	B-A <b>0.59</b>
PA	C <b>23.33</b>	D <b>21.17</b>	D-C <b>-2.16</b>
$\Delta$ (trtmt-ctrl)	A-C <b>-2.89</b>	B-D <b>-0.14</b>	<b>(0.59) - (-2.16) = 2.76</b>

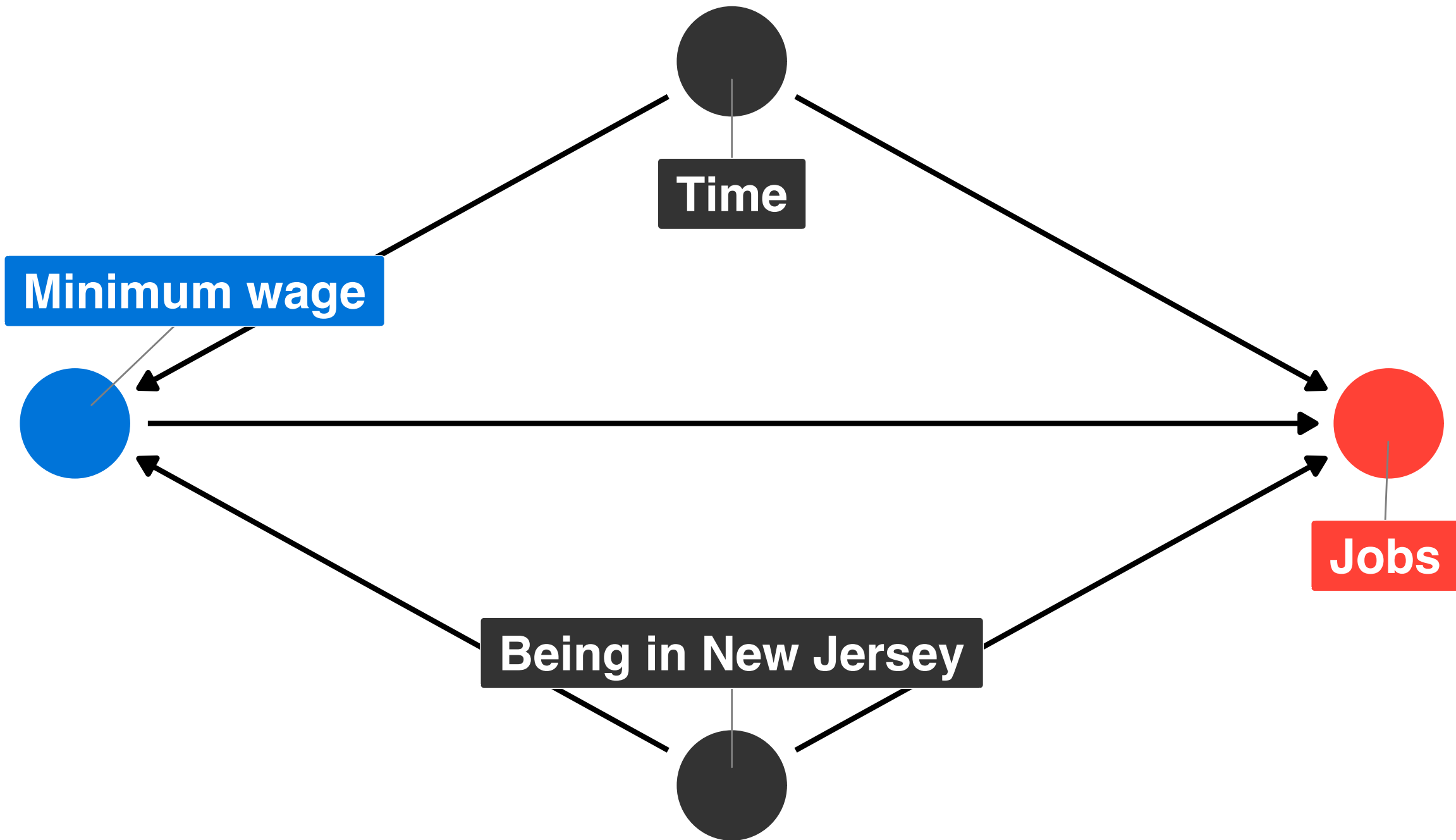


	Pre mean	Post mean	$\Delta$ (post-pre)
Math story	A	B	B-A
Normal story	C	D	D-C
$\Delta$ (trtmt-ctrl)	A-C	B-D	(B-A) - (D-C)

**Finding all the group means is tedious though!**

**What if there are other backdoors to worry about?**

**Regression to the rescue!**





$$Y_{it} = \alpha + \beta \text{Group}_i + \gamma \text{Time}_t + \delta (\text{Group}_i \times \text{Time}_t) + \epsilon_{it}$$

```
model <- lm(outcome ~ group + time + group * time)
```

**Group = 1/TRUE if treatment**

**Time = 1/TRUE if after**

$$Y_{it} = \alpha + \beta \text{ Group}_i + \gamma \text{ Time}_t + \delta (\text{Group}_i \times \text{Time}_t) + \epsilon_{it}$$

```
model <- lm(outcome ~ group + time + group * time)
```

**$\alpha$  = Mean of control, pre-treatment**

**$\beta$  = Increase in outcome across groups**

**$\gamma$  = Increase in outcome across time**

**$\delta$  = Difference in differences!**

$$Y_{it} = \alpha + \beta \text{ Group}_i + \gamma \text{ Time}_t + \delta (\text{Group}_i \times \text{Time}_t) + \epsilon_{it}$$

	Pre mean	Post mean	$\Delta$ (post-pre)
Control	$\alpha$	$\alpha + \gamma$	$\gamma$
Treatment	$\alpha + \beta$	$\alpha + \beta + \gamma + \delta$	$\gamma + \delta$
$\Delta$ (trtmt-ctrl)	$\beta$	$\beta + \delta$	$\delta$

# Our turn

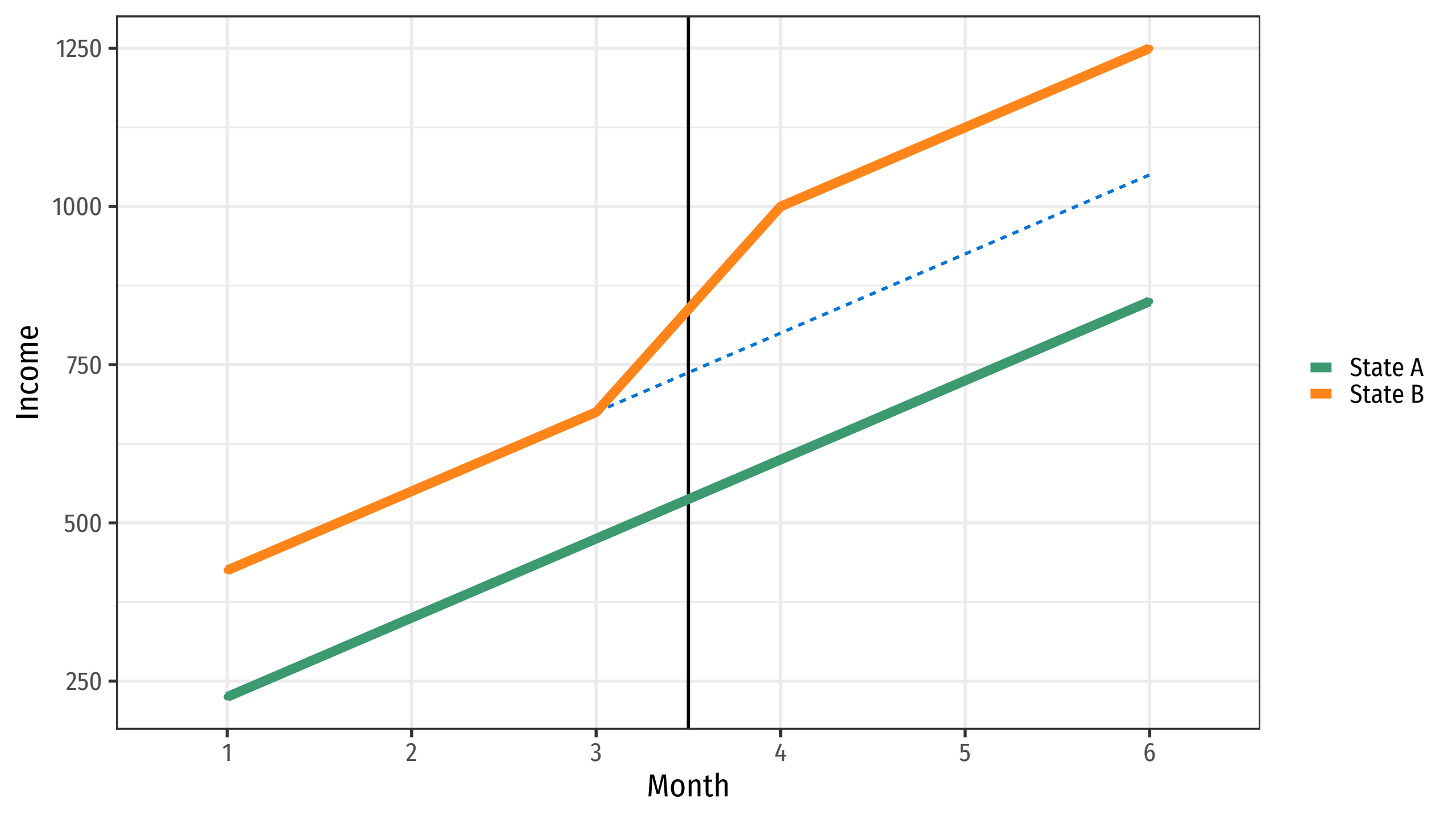
**Let's calculate diff-in-diff  
estimates with R!**

# DiD assumptions

# Assumptions

## Parallel trends

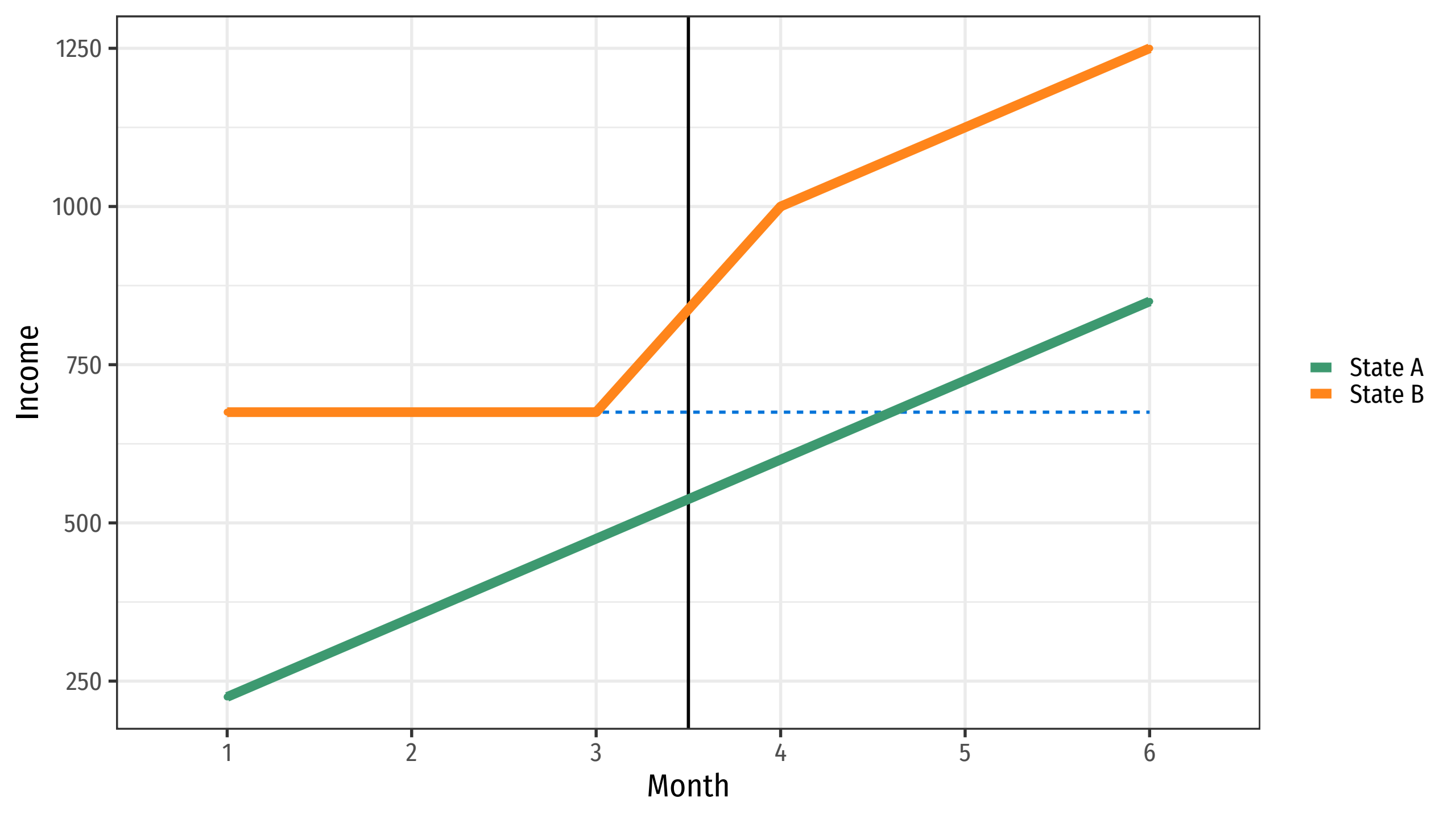
**Treatment and control might have different values at first, but we assume treatment group would have changed like control in absence of treatment**



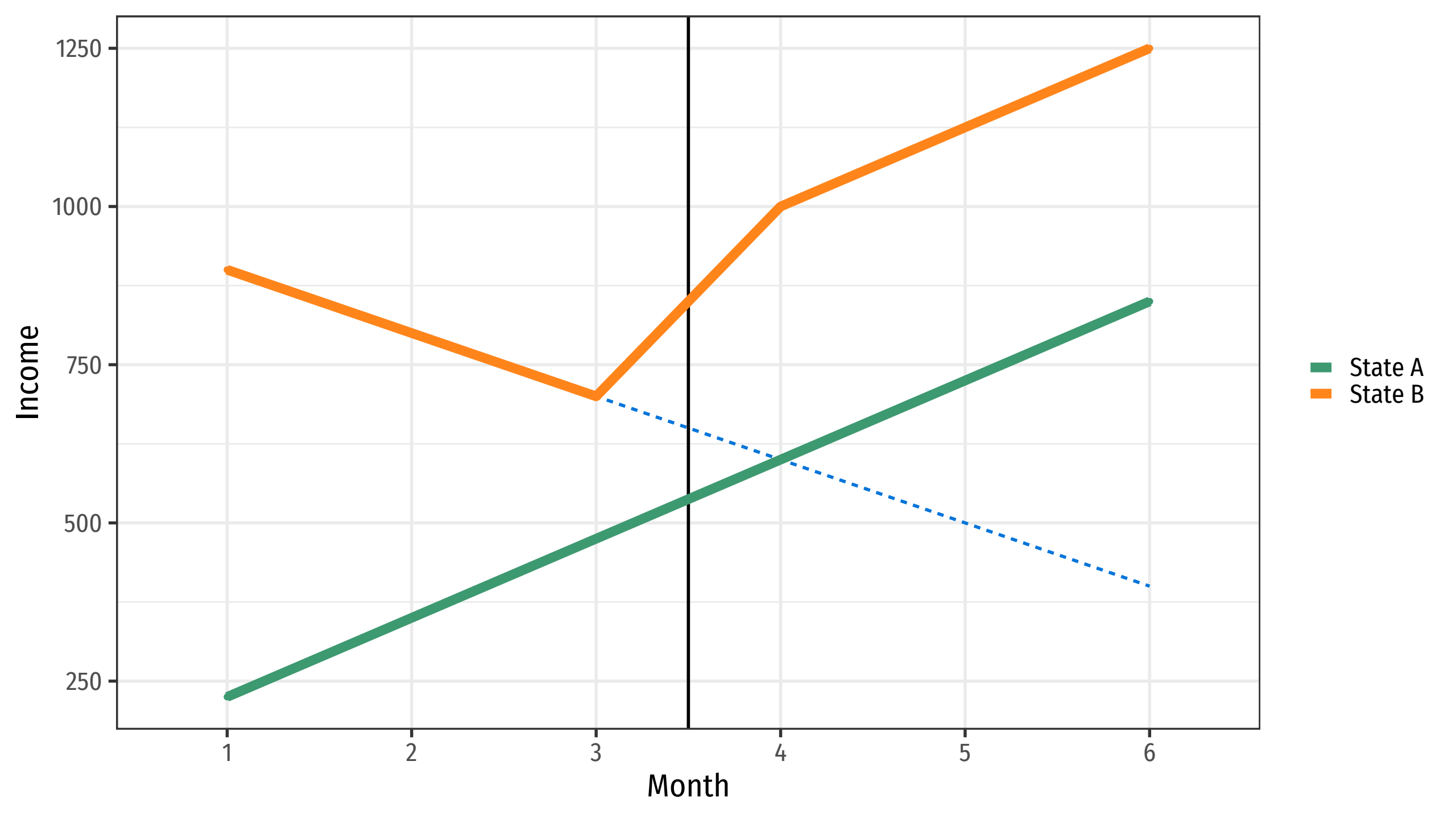
Income

Month

State A  
State B







# Assumptions

## Parallel trends

**Check by pretending the treatment happened earlier.  
If there's an effect, there's an underlying trend.**

