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1. $\begin{pmatrix} \varepsilon_i \\ T_i \end{pmatrix} \sim N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}\right)$ \leftarrow joint distribution of $\begin{pmatrix} \varepsilon_i \\ T_i \end{pmatrix}$ unconditional of X_i

From this, $T_i \perp \varepsilon_i$ because $\text{cov}(T_i, \varepsilon_i) = 0$ & this is a joint distribution independent of X_i . This necessarily also implies that $T_i \perp \varepsilon_i | X_i$ because they are independent $\text{cov}(T_i, \varepsilon_i) = 0$ w/ values of X

2. $\text{cov}(T_i, X_i) = \text{cov}(T_i, aT_i + u_i) = a \text{cov}(T_i, T_i) + \text{cov}(T_i, u_i) = a \text{cov}(T_i, X_i) = \rho_2$ from 0

$$\therefore \rho_2 = 0$$

3.
$$\begin{aligned} E(Y_i | T_i) &= E(\gamma T_i + \beta X_i + \varepsilon_i | T_i) = \gamma E(T_i | T_i) + \beta E(X_i | T_i) + E(\varepsilon_i | T_i) \\ &= \gamma T_i + \beta E(aT_i + u_i | T_i) = \gamma T_i + a\beta E(T_i | T_i) + \beta E(u_i | T_i) \\ &= \gamma T_i + a\beta T_i = (\gamma + a\beta) T_i = (\gamma + \rho_2 \beta) T_i \end{aligned}$$

$$\Rightarrow \phi = \gamma + \rho_2 \beta$$

4. In order for $E(\gamma + \rho_2 \beta) = \gamma$, we need $\rho_2 = 0$ because $\beta > 0$. This implies T_i & X_i will be uncorrelated \Rightarrow If T_i & X_i are uncorrelated, the coefficient of T_i from $E(Y_i | T_i)$ is an unbiased estimate of γ .

5.
$$\begin{aligned} \rho_1 &= \text{cov}(X_i, \varepsilon_i) = \text{cov}(X_i, bT_i + cX_i + u_i) \\ &= b \text{cov}(X_i, T_i) + c \text{cov}(X_i, X_i) + \text{cov}(X_i, u_i) \\ &= b\rho_2 + c \\ \Rightarrow \rho_1 &= b\rho_2 + c \end{aligned}$$

$$\begin{aligned} 0 &= \text{cov}(T_i, \varepsilon_i) = \text{cov}(T_i, bT_i + cX_i + u_i) \\ &= b \text{cov}(T_i, T_i) + c \text{cov}(T_i, X_i) + \text{cov}(T_i, u_i) \\ &= b + c\rho_2 \\ \Rightarrow 0 &= b + c\rho_2 \end{aligned}$$

$$\begin{aligned}
 6. \quad E(Y_i | X_i = x, T_i = t) &= E(\gamma T_i + \beta X_i + \varepsilon_i | X_i = x, T_i = t) \\
 &= \gamma E(T_i | X_i = x, T_i = t) + \beta E(X_i | X_i = x, T_i = t) + E(\varepsilon_i | X_i = x, T_i = t) \\
 &= \gamma t + \beta x + \frac{\rho_1}{1-\rho_2^2} x - \frac{\rho_1 \rho_2}{1-\rho_2^2} t \\
 &= \left(\gamma - \frac{\rho_1 \rho_2}{1-\rho_2^2}\right) t + \left(\beta + \frac{\rho_1}{1-\rho_2^2}\right) x \\
 \Rightarrow \quad \boxed{\gamma} &= \gamma - \frac{\rho_1 \rho_2}{1-\rho_2^2} \quad \& \quad \boxed{\beta} = \beta + \frac{\rho_1}{1-\rho_2^2}
 \end{aligned}$$

7. $\boxed{\gamma}$ will be unbiased if $\frac{\rho_1 \rho_2}{1-\rho_2^2} = 0$, which happens if $\rho_1 = 0$ and/or $\rho_2 = 0$ and $\rho_2 \neq \pm 1$

This is essentially saying that if X_i is uncorrelated with ε_i and/or T_i while also not being perfectly correlated with ε_i , $\boxed{\gamma}$ is an unbiased estimator of γ

8. I would say that's ridiculous. This is possible when ρ_1 is large & ρ_2 is small.