

Causal Inference - session 2

2023.07.25 김정희

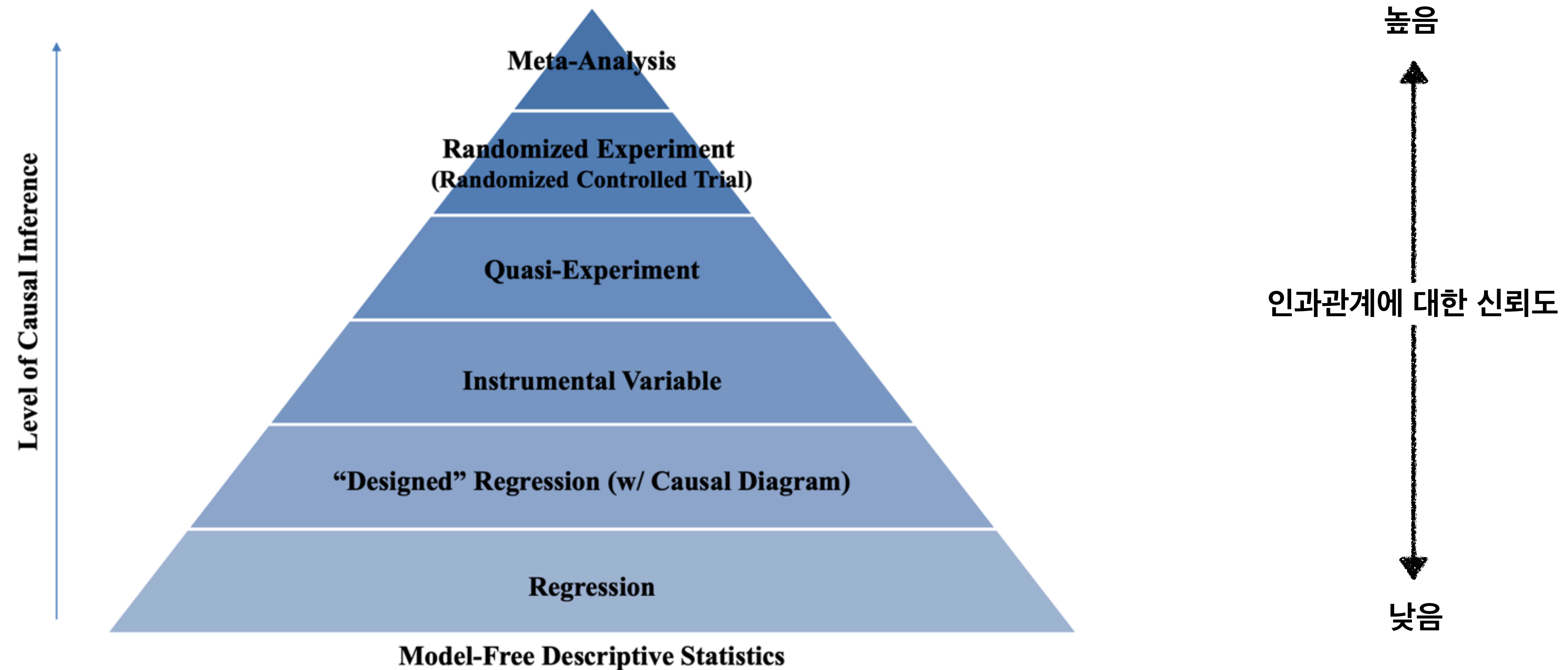
00. Causal Inference

Ceteris Paribus

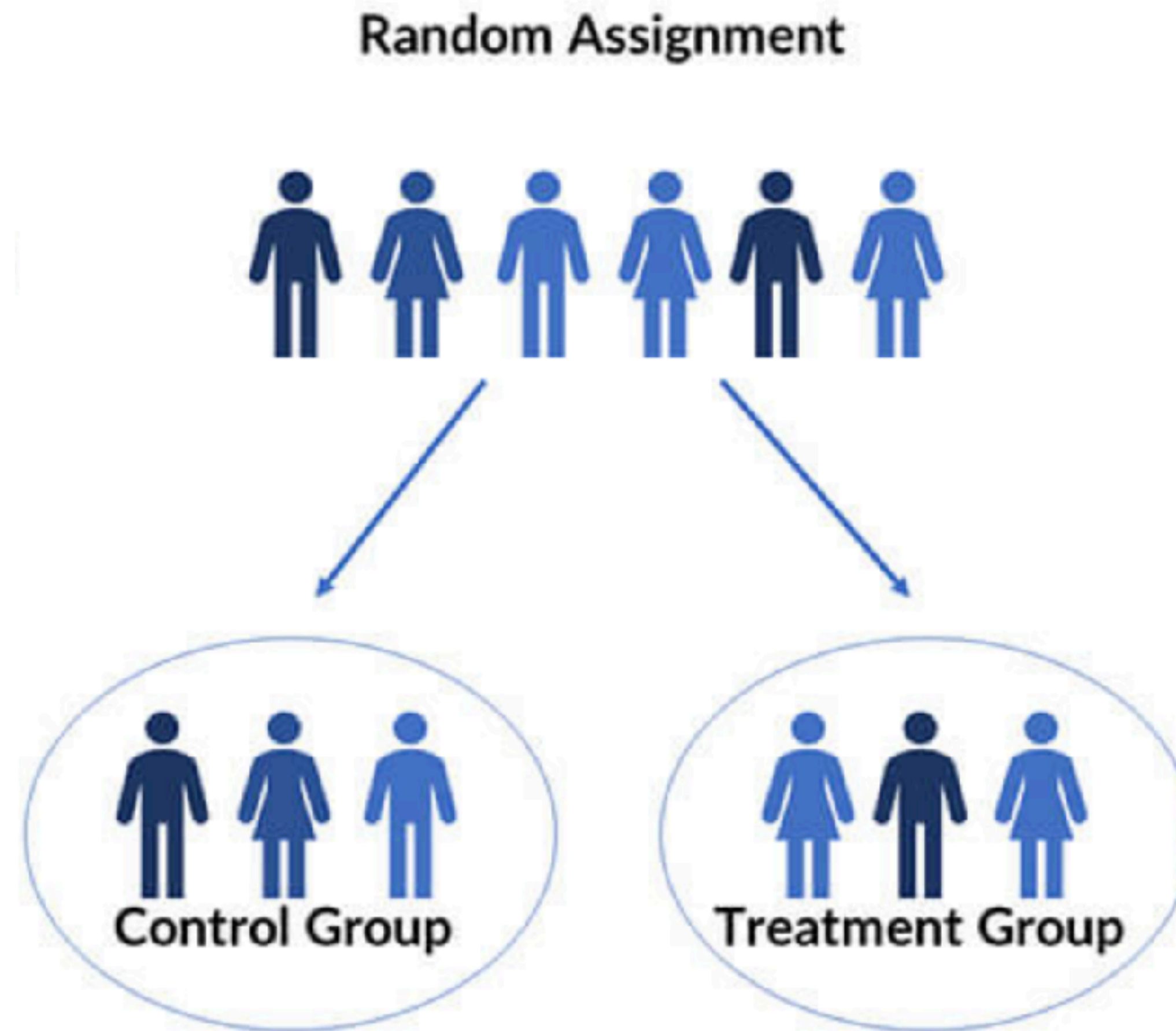
- ‘다른 모든 조건이 동일하다면’
- ‘Ceteris Paribus’를 만족하는 Control Group을 찾을 수 있게 연구 디자인을 고안한다면, 인과관계 추론을 성공적으로 할 수 있을것임.

00. Causal Inference

Causal Hierarchy of Research Design for Causal Inference

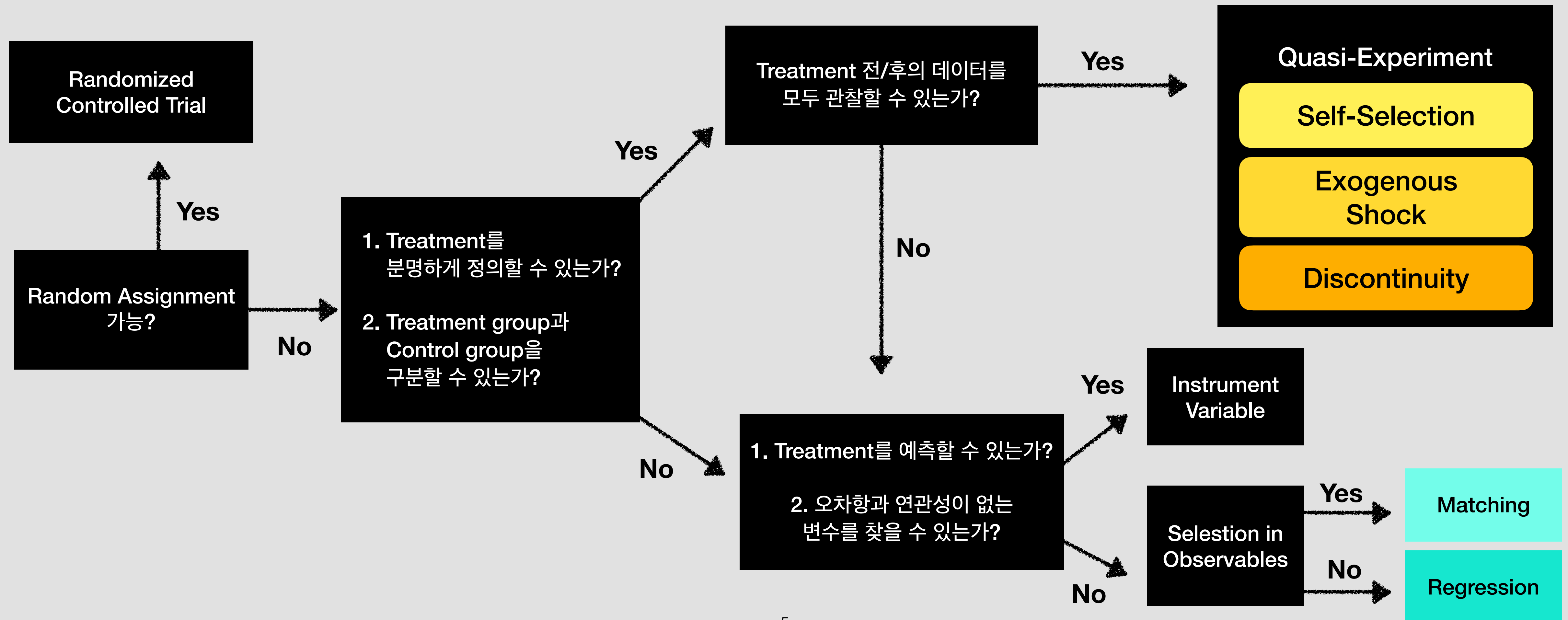


01. Random Assignment



02. Quasi-Experiment

2-1) Research Design



03. Difference-in-Differences

3-1) DID estimator

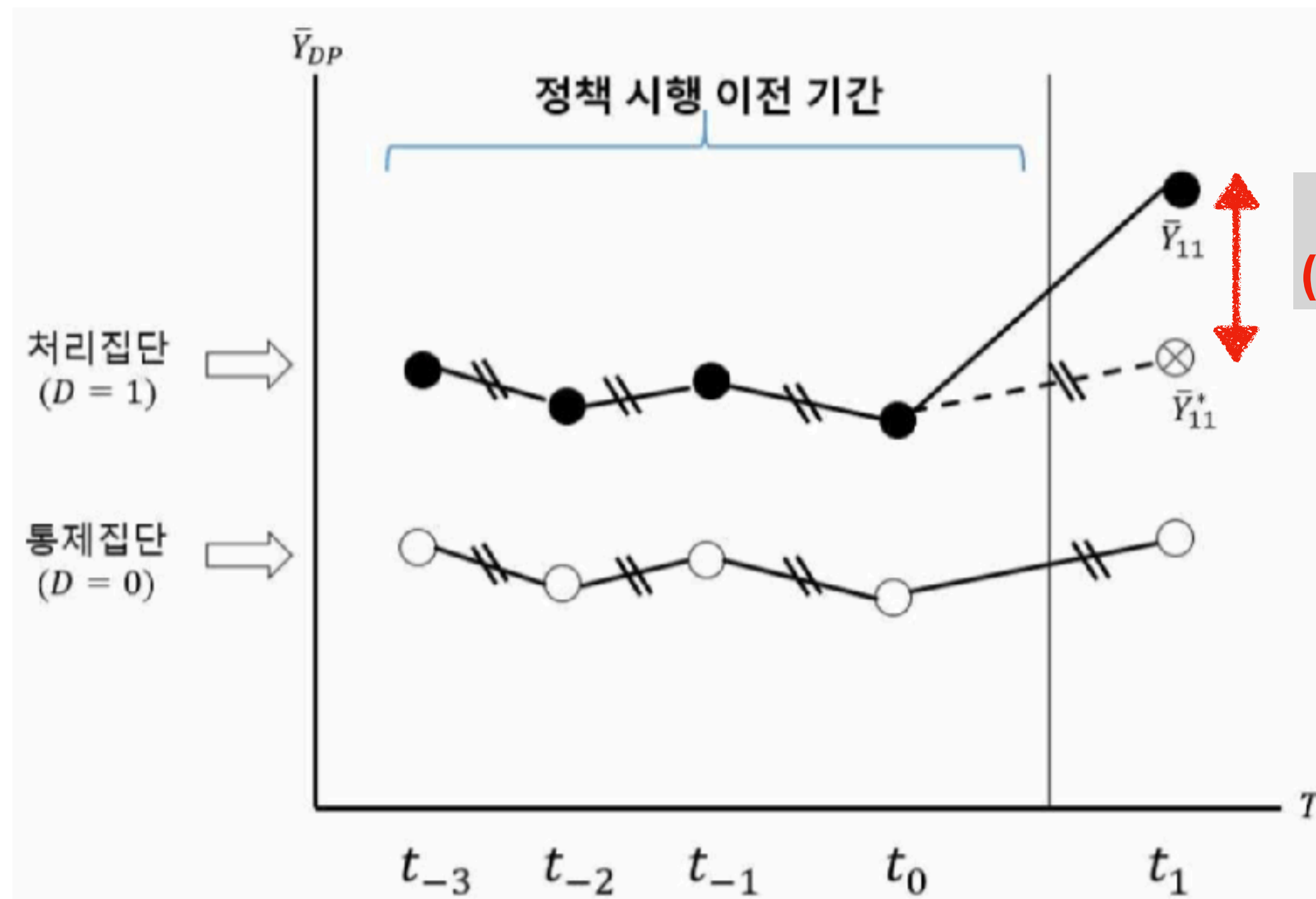
	Before Treatment	After Treatment
Treatment Group	Tb	Ta
Control Group	Cb	Ca

$$DID\ estimator = (Ta - Tb) - (Ca - Cb)$$

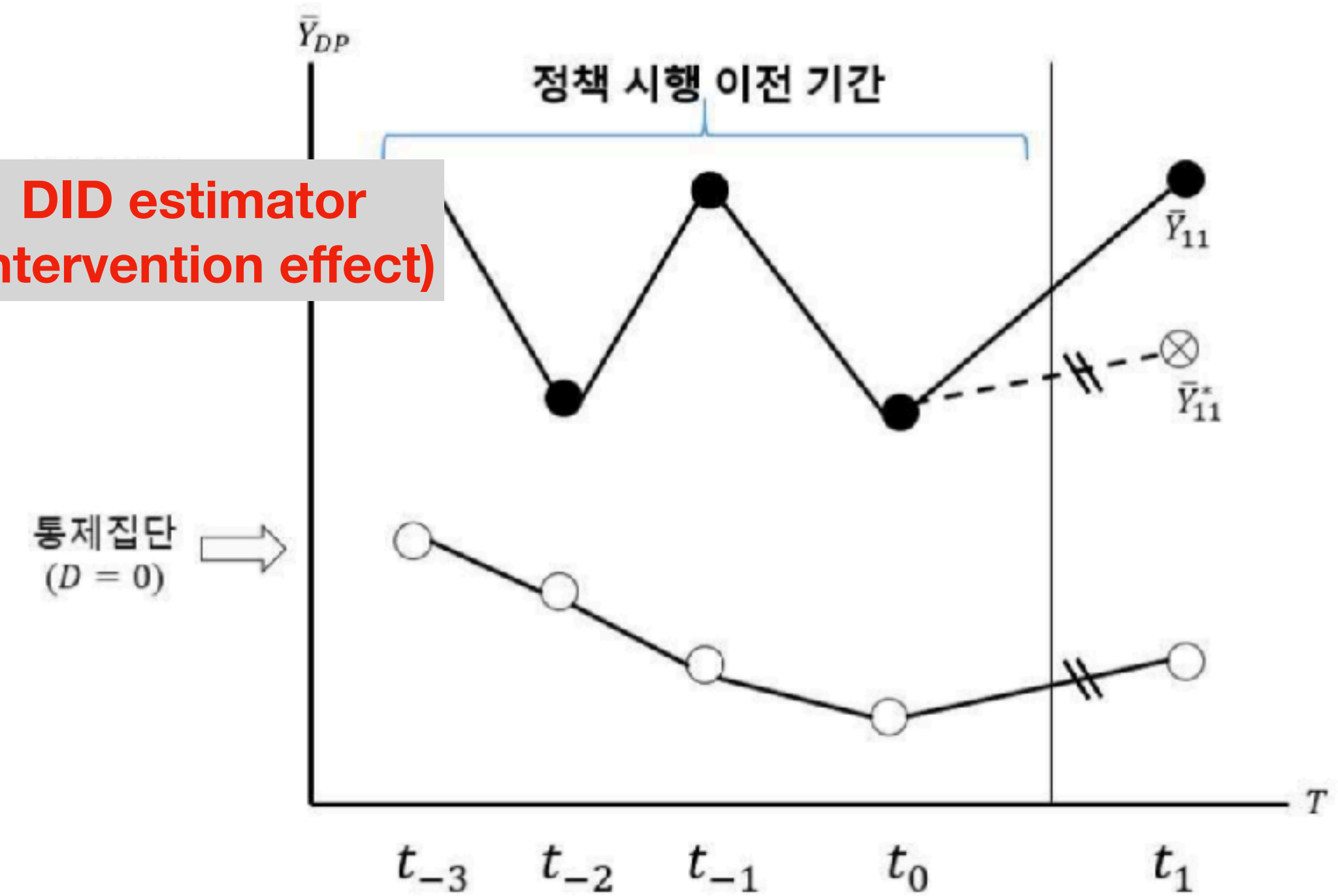
03. Difference-in-Differences

3-2) Parallel trend assumption

〈그림 7〉 DiD 추정량의 평행 추세 가정이 만족할만한 상황



〈그림 8〉 DiD 추정량의 평행 추세 가정이 만족하지 않을만한 상황



03. Difference-in-Differences

??????????

	Before Treatment	After Treatment	
Treatment Group	T_B	T_A (counterfactual: T'_A)	
Control Group	C_B	C_A	DID estimator = $(T_A - T_B) - (C_A - C_B)$
Counterfactual outcome in the absence of treatment	T_B	$T_B + (T'_A - T_B)$	Change for the treated in the absence of treatment (not observed)
Counterfactual outcome inferred from the control group	T_B	$T_B + (C_A - C_B)$	Inferred from the control group unaffected by the treatment

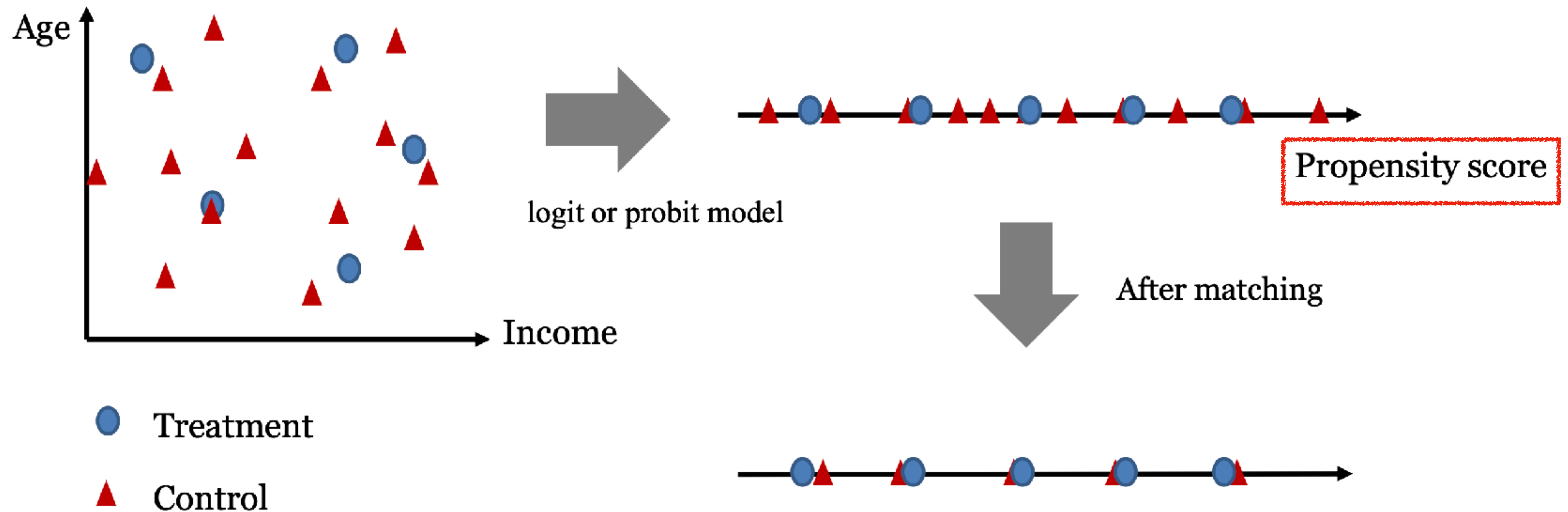
How similar is the control group to the treatment group in the absence of the treatment?

04. Matching Techniques

- Propensity Score Matching (PSM)
- Coarsened exact Matching (CEM)

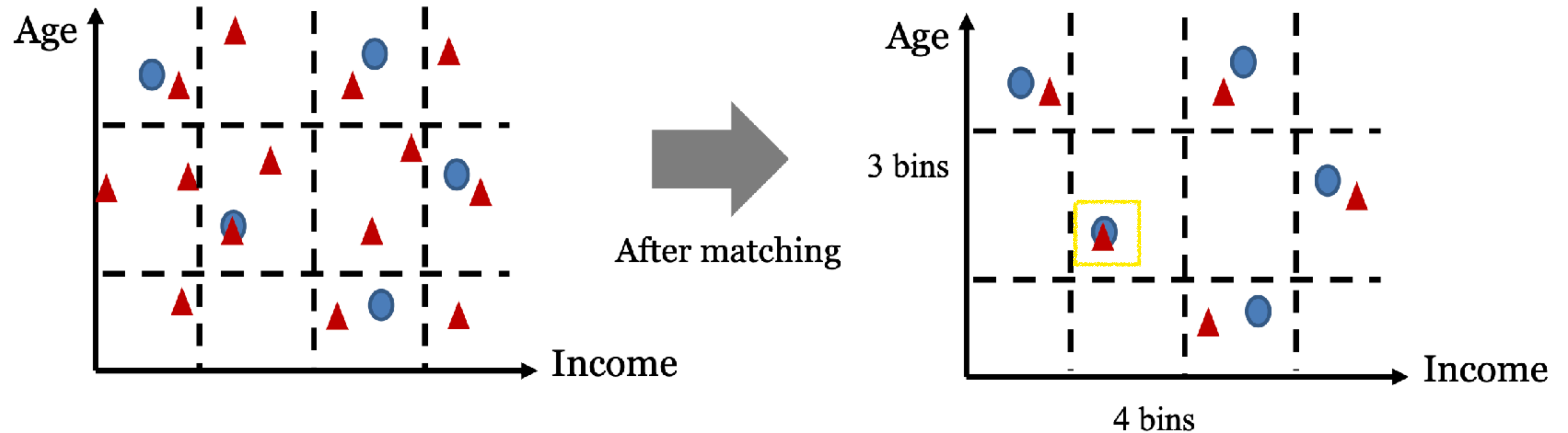
04. Matching Techniques

4-1) Propensity Score Matching (PSM)



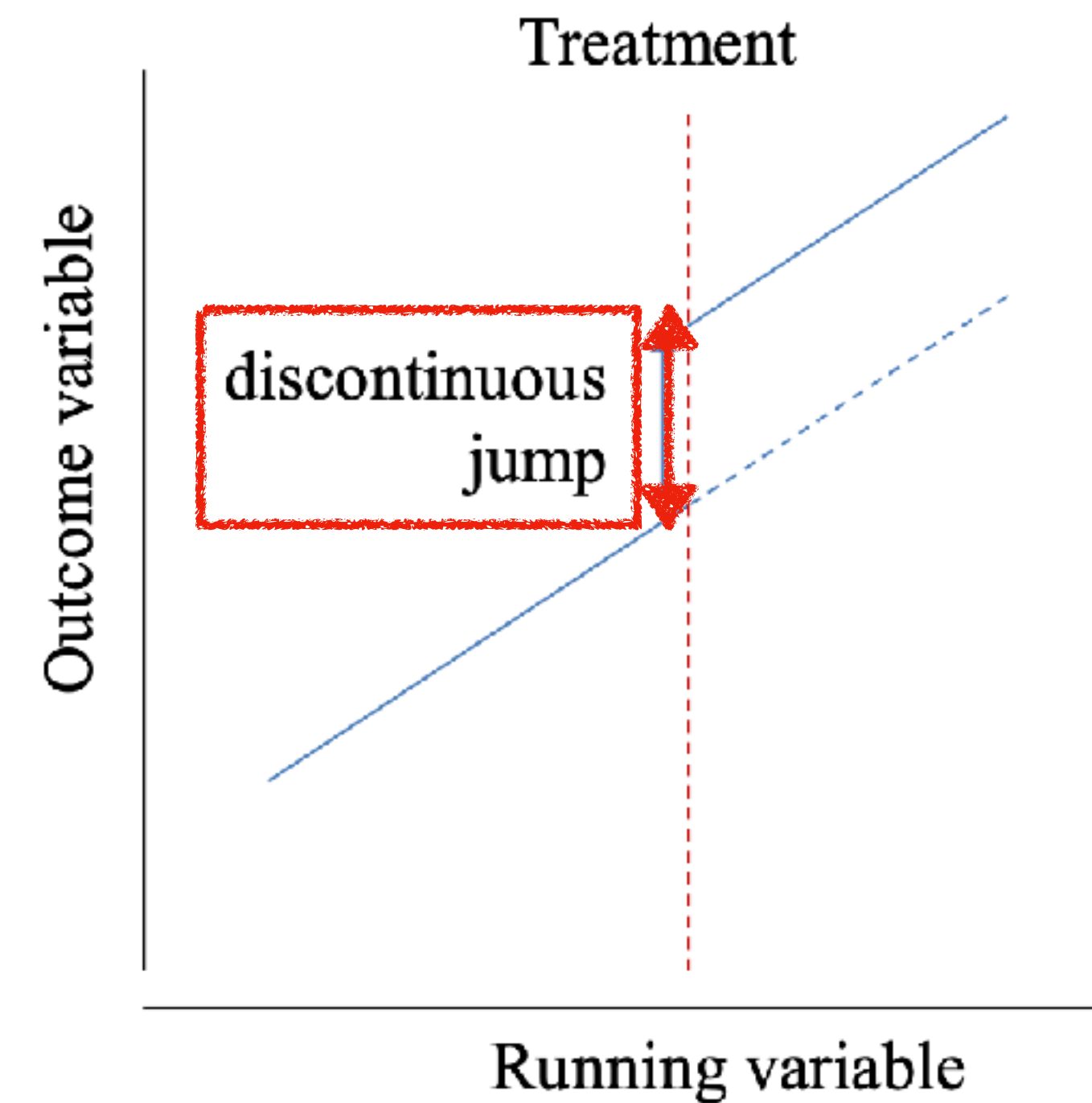
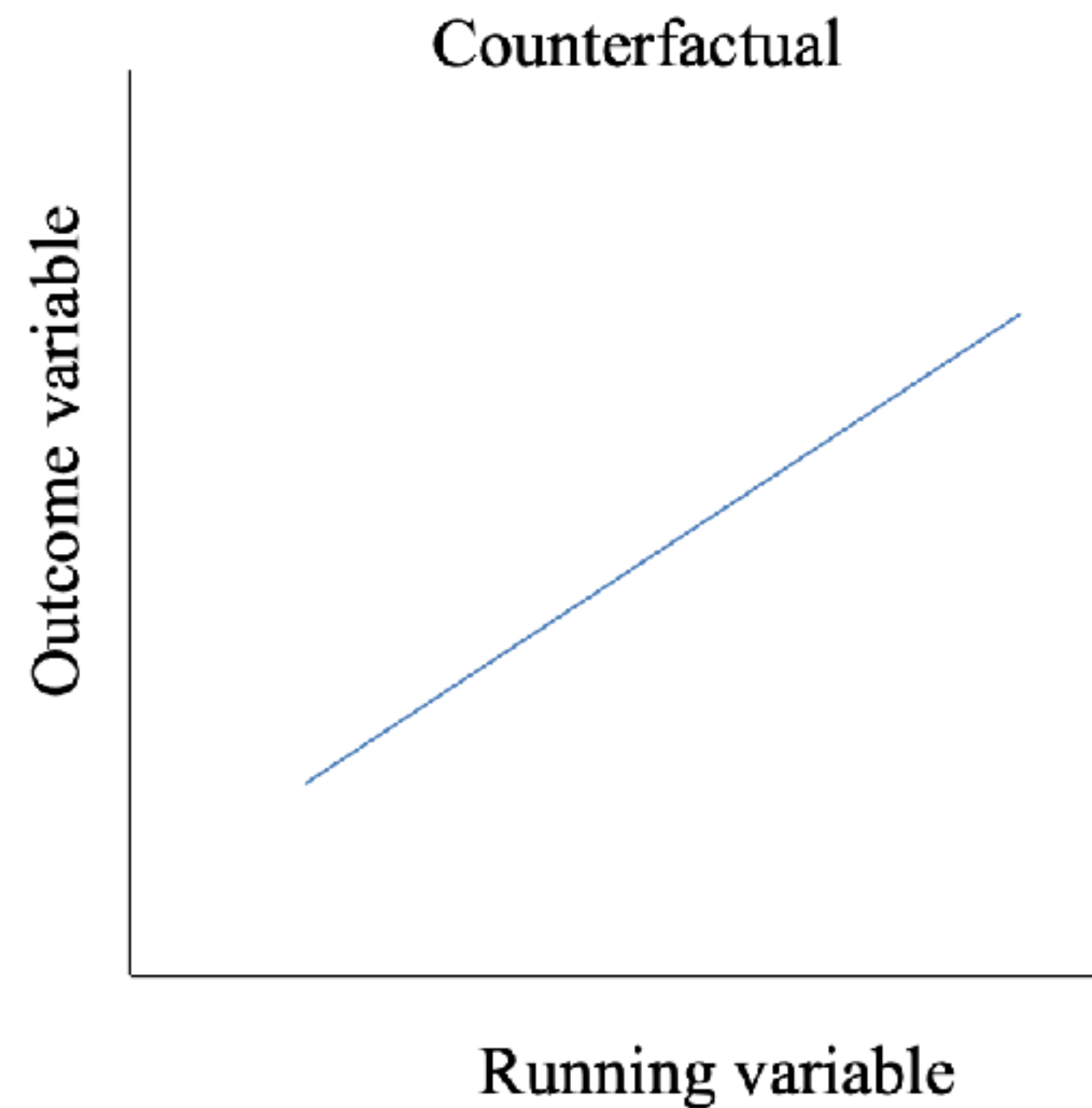
04. Matching Techniques

4-2) Coarsened Exact Matching (CEM)



05. Regression Discontinuity

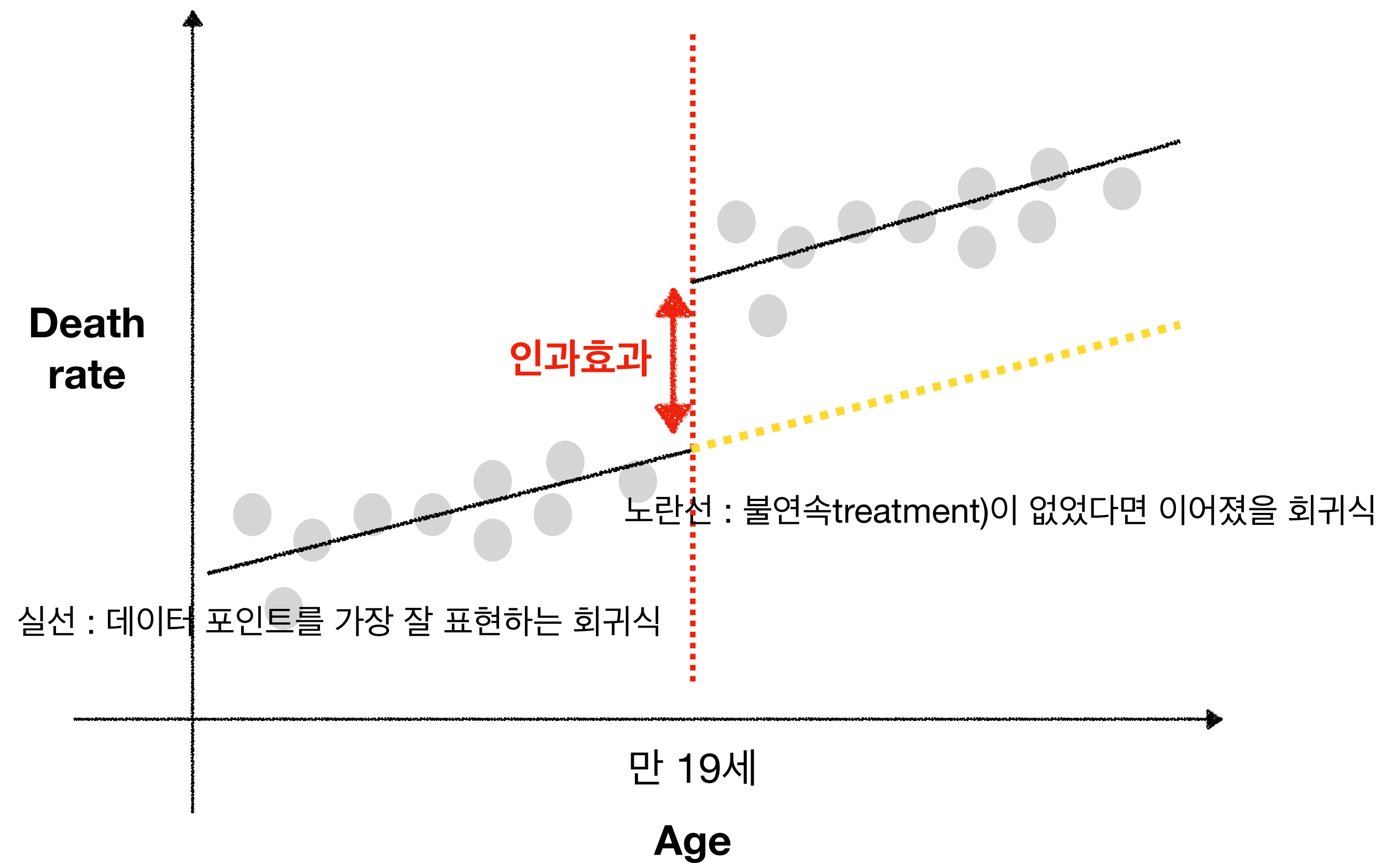
5-1) Discontinuous jump



05. Regression Discontinuity

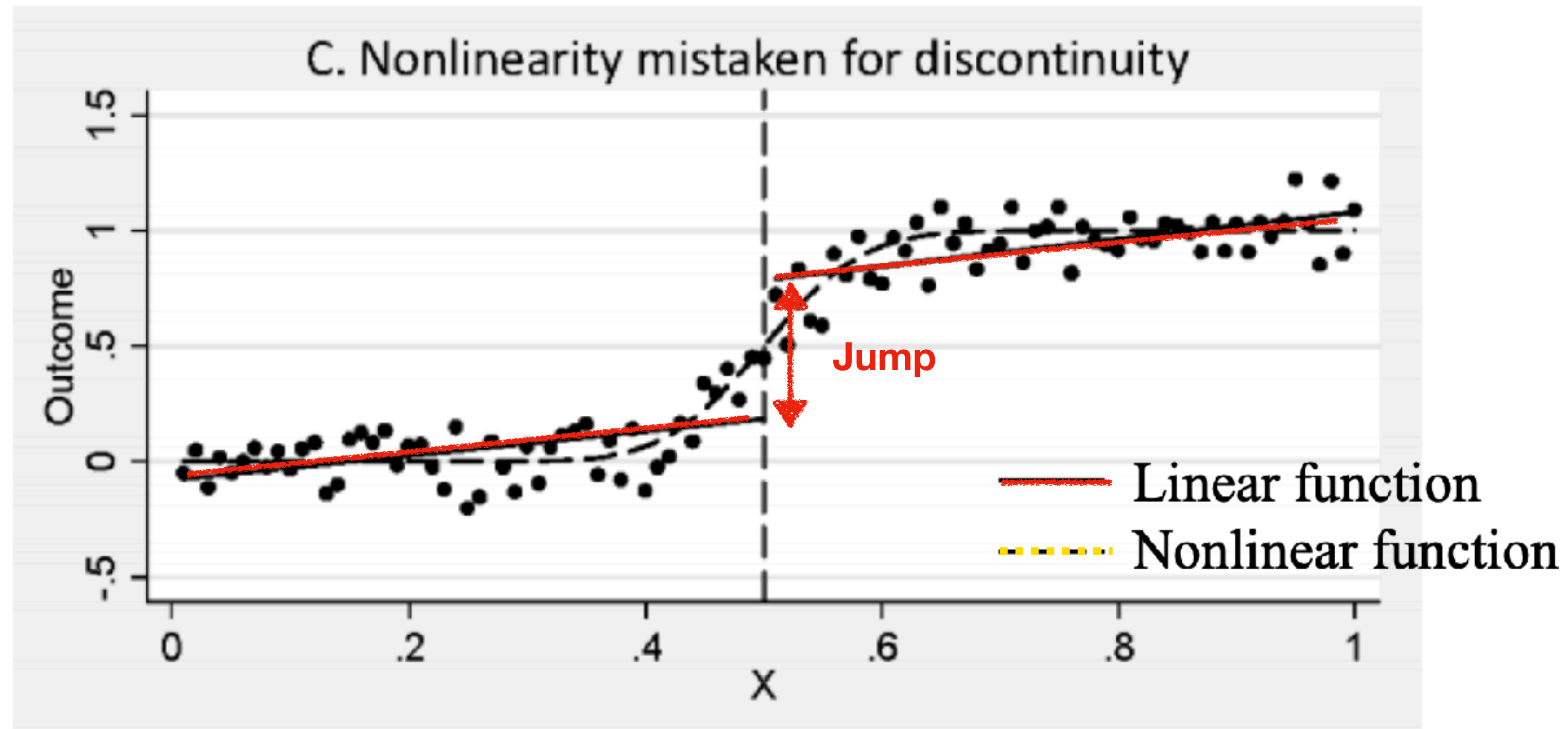
5-1) Discontinuous jump

< 법적 최소 ^{처치} 음주 연령제도가 ^{결과변수} 사망률에 주는 영향 >



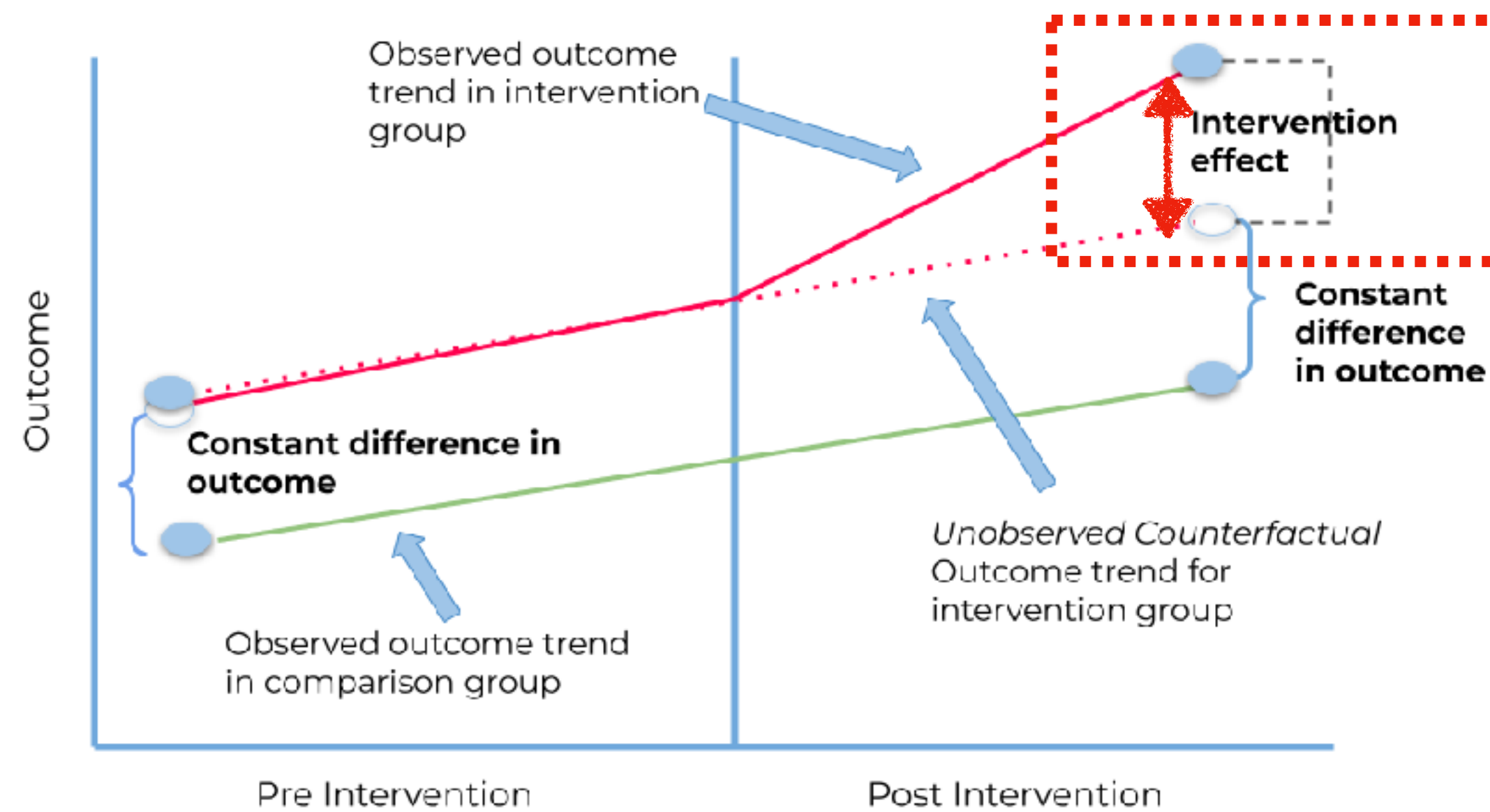
05. Regression Discontinuity

5-1) Discontinuous jump

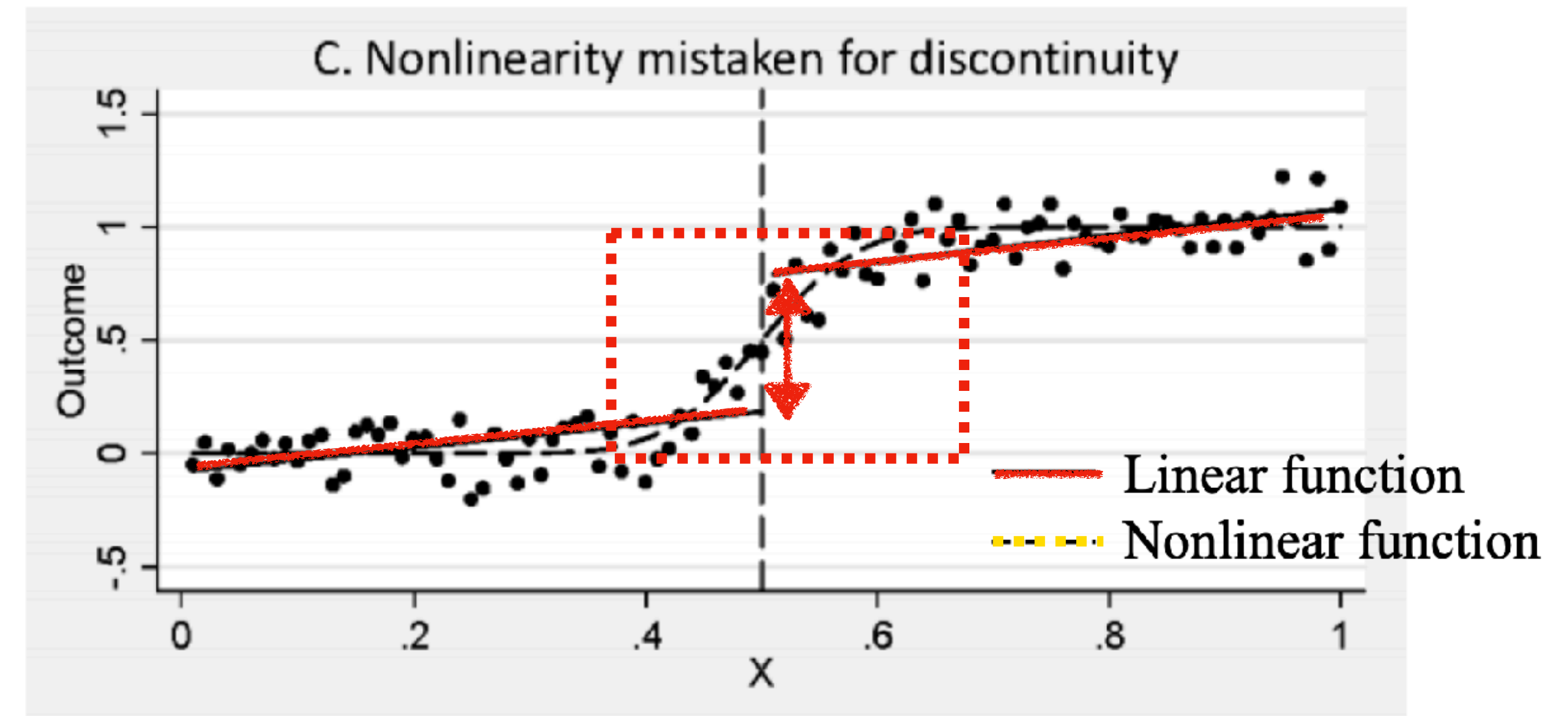


05. Regression Discontinuity

5-2) DID vs RD



< *Difference in Difference* >



< *Regression Discontinuity* >