

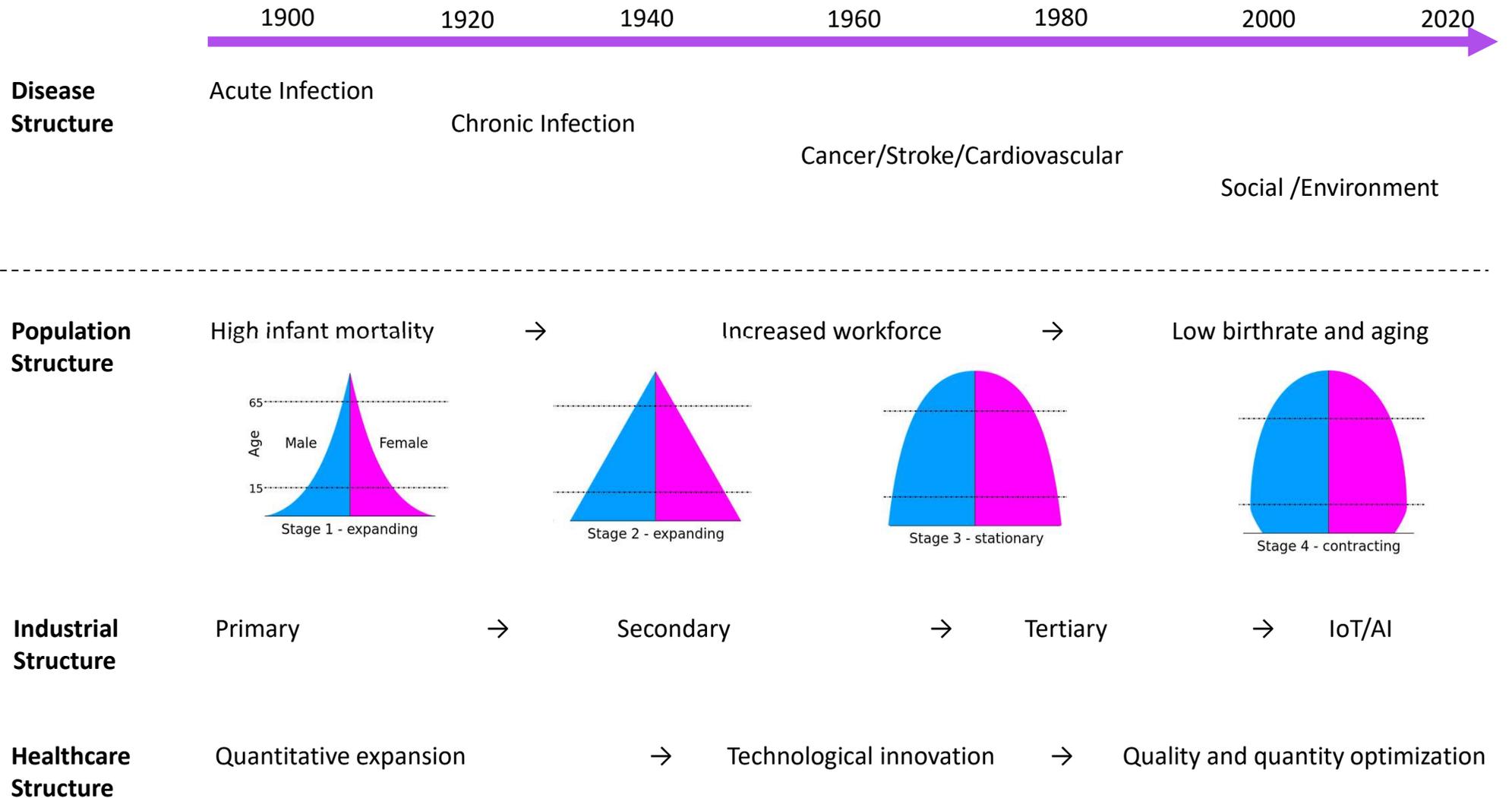
Study designs and statistical methods for current observational studies

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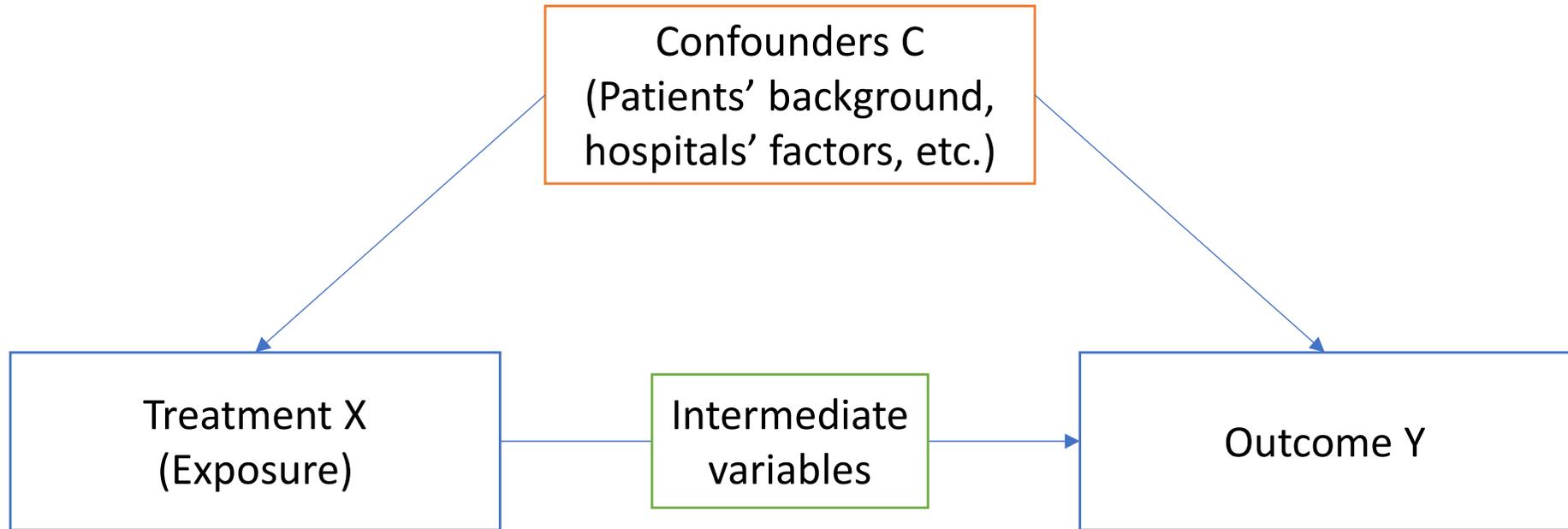
Health Transition



Current features of observational research

- Observational data
(i.e., administrative data, claims data, and GPS data)
- Improved PC computing power
- Development of research questions that we need to consider many confounders and unmeasured confounders

Confounding



- Confounders associate with the outcome and treatment. They are not intermediate variables between treatment and outcome.
- We often use the adjustment, stratification, matching, and weighting to deal with confounders.

Issues on the regression analysis

- If there are many confounders and unobserved confounders, the following issues would occur.
 - Misconfiguration of the model
 - If too many variables need to be included in a model relative to the number of events, the estimates from these models can be incorrect.

Am J Epidemiol 2003; 158. 280-7

Study designs and statistical analysis in current observational studies

1. Difference-in-Differences
 2. Instrumental Variable Analysis
 3. Regression Discontinuity
 4. Panel Data Analysis / Interrupted Time-Series
 5. Propensity Score Analysis (matching, weighting, stratification, and adjustment using propensity scores)
 6. Adjustment(regression), weighting, stratification, and matching
- Natural experimental methods (Quasi-experimental methods)

Difference-in-Differences Analysis

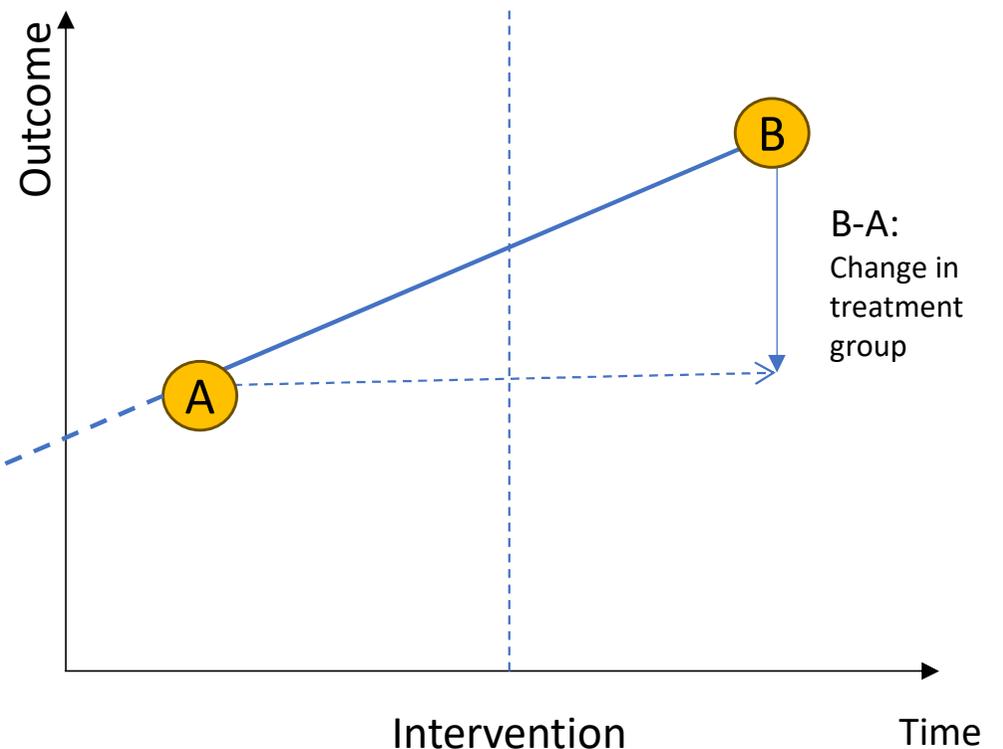
Reference:

Dimick, J. B. and A. M. Ryan (2014). Methods for evaluating changes in health care policy: the difference-in-differences approach. *JAMA* 312(22): 2401-2402.

Pre-post comparison

- Only one group: everyone gets treated
- Counterfactual: the outcome would not have changed without the intervention
- Assumption: the outcome would be constant before and after the intervention
- “Simple differences” over time could represent:
 - Actual interventional effect
 - Other factors that may have affected the group over time
 - Mix of both

→ Absolute changes are weak evidence for causal effects



	Treatment
Before	A
After	B
Change	B-A

Difference-in-Differences design

Key assumptions

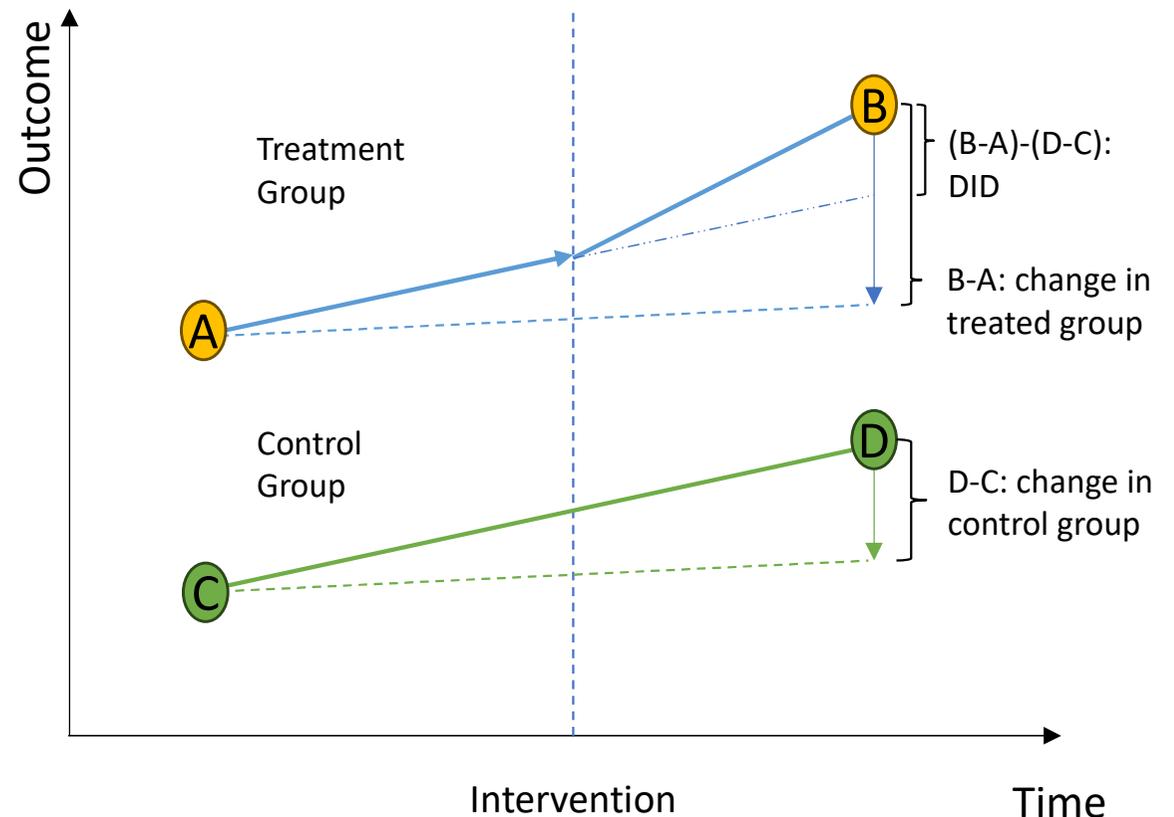
1. Parallel trends assumption

If there is no intervention, the outcomes of the treatment and control group would be a parallel trend.

Demonstrate that the outcomes of the treatment and control groups are parallel at several time points before the intervention.

2. Common shocks assumption

There are no other exogenous events occurring that affect pre- and post-intervention outcomes, or if they occur, they affect the two groups in the same way.



DID

= (change in treatment) – (change in control)

= ($Treat^{After} - Treat^{Before}$) – ($Control^{After} - Control^{Before}$)

	Treatment	Control
Before	A	C
After	B	D
Change	B-A	D-C
DID	(B-A) – (D-C)	

DID example

Did the first COVID-19 policy announced by the Japanese government affect asthma patients?

Intervention: COVID-19 policy (Week 9)

Treatment: 2020

Control: 2017-2019

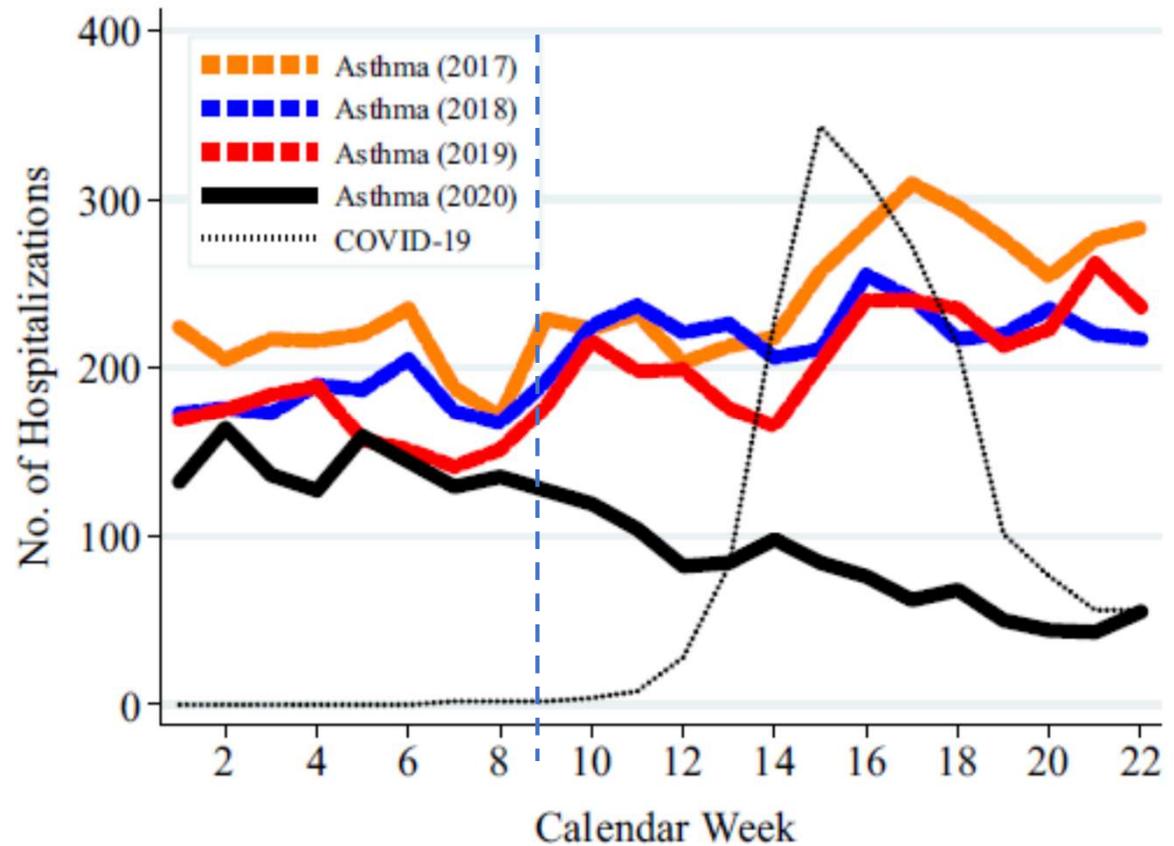
DID

$$=(78.3-140.8)-(229.9-184.9)$$

$$= -107.5 \text{ per week}$$

K. Abe, 2.7.2024

FIGURE 1. Trends in hospitalizations for asthma and COVID-19 for Japanese acute-care hospitals, 2017 through 2020.



Average number of hospitalizations per week					
2017-2019			2020		
Weeks 1-8	Weeks 9-22	Ratio (difference)	Weeks 1-8	Weeks 9-22	Ratio (difference)
184.9	229.9	1.24 (45.0)	140.8	78.3	0.56 (-62.5)

Abe K, Miyawaki A, Nakamura M, Ninomiya H, Kobayahi Y. Trends in Hospitalizations for Asthma During the COVID-19 Outbreak in Japan. The Journal of Allergy and Clinical Immunology: In Practice. 2020;9(1):494-6.e1.

To be continued