### **ABSTRACT**

1. **Practical Problem:** Test whether it is MORE PROBABLE THAN NOT that the defendant's action was a NECESSARY CAUSE for the plaintiff's injury (or death).

### 2. Theoretical Problems:

- (a) What is the meaning of PN(x,y):

  "Probability that event y would not have occured if it were not for event x, given that x and y did in fact occur."
- (b) Under what conditions can PN(x,y) be estimated from statistical data, i.e., observational, experimental and combined.

## **REVIEW OF COUNTERFACTUALS**

- 1. **Semantics:**  $Y_x(u) \triangleq \text{ solution for } Y \text{ in } M_x$   $Y_x(u) = y : Y = y \text{ if } X \text{ were } x \text{ (in background } u)$
- 2. Abreviations:  $Y_x(u) = y \Leftrightarrow y_x(u)$  or  $y_x$   $P(Y_x = y) \Leftrightarrow P(y_x)$

 $y' \stackrel{\triangle}{=}$  complement of y

e.g.,

$$y'_x \equiv Y_x \neq y$$
 $y_x \lor y'_x \equiv true$ 
 $P(y_x|x) = P(y|x)$ 
 $P(y'_x) = 1 - P(y_x)$ 

## THE SEMANTICS OF NECESSARY AND SUFFICIENT CAUSES

## **Necessary Cause**

Event x was a necessary cause of event y if the probability

$$PN = P(y'_{x'}|x,y)$$
 is HIGH

### **Sufficient Cause**

Event  $\boldsymbol{x}$  is a sufficient cause of event  $\boldsymbol{y}$  if the probability

$$PS = P(y_x | x', y')$$
 is HIGH

(e.g., benefit of treating the untreated sick)

Necessary and Sufficient Cause Event x is a necessary-and-sufficient cause of event y if the probability

$$PNS = P(y_x, y'_{x'})$$
 is HIGH

### **EXOGENEITY**

**Definition 1** (no confounding = exogeneity)

M = model of the data-generating process.

 $P_M(y|do(x)) = probability \ of \ Y = y \ under \ the$ hypothetical intervention X = x.

We say that X and Y are **not confounded** in M (or, X is **exogenous**) if and only if

$$P_M(y|do(x)) = P_M(y|x)$$

Alternatively,

$$P_M(y_x) \stackrel{\Delta}{=} P_M(Y_x = y) = P_M(y|x)$$

 $Y_x$ — the value of Y if X were x[Neyman, 1926; Rubin, 1974]

### **BOUNDS AND BASIC RELATIONSHIPS**

What if we do not have the functional relationships behind the data-generation mechanisms?

What can be done with joint distribution alone?

### Theorem 1:

Under condition of exogeneity, PNS is bounded as follows:

max 
$$[0, P(y|x) + P(y'|x') - 1] \le PNS$$
  
 $PNS \le \min[P(y|x), P(y'|x')]$ 

and

$$PN = \frac{PNS}{P(y|x)}$$
  $PS = \frac{PNS}{1 - P(y|x')}$ 

## **Interpretation:**

The probability of necessity can take on any value in the interval

$$ERR \stackrel{\triangle|}{=} 1 - \frac{1}{RR} \stackrel{\triangle}{=} \frac{P(y|x) - P(y|x')}{P(y|x)} \le PN \le 1$$

# IDENTIFIABILITY UNDER EXOGENEITY AND MONOTONICITY

### Theorem 3

If X is exogenous and Y is monotonic in X, then the probabilities PN, PS, and PNS are all identifiable, and are given by:

$$PNS = P(y|x) - P(y|x')$$
 risk-difference 
$$PN = [P(y|x) - P(y|x')]/P(y|x)$$
 excess-risk-ratio 
$$PS = [P(y|x) - P(y|x')]/P(y'|x')$$
 susceptibility

### **Interpretation:**

There is wisdom to epidemiological myths, BUT:

- 1. Caution: We need to ascertain monotonicity (no prevention, no reversal)
- Relief: No need to assume independence (between susceptibility and background factors)

# BOUNDS UNDER EXOGENEITY AND NONMONOTONICITY

**Theorem** 3' (Tian & Pearl 2000) If X is exogenous then the probabilities PN, PS, and PNS are all

### Lower Bounded by:

$$PNS \geq P(y|x) - P(y|x')$$
 risk-difference  $PN \geq [P(y|x) - P(y|x')]/P(y|x)$  excess-risk-ratio  $PS \geq [P(y|x) - P(y|x')]/P(y'|x')$  susceptibility

## **Interpretation:**

There is wisdom to epidemiological myths, since we need **not** ascertain monotonicity (no prevention, no reversal) **for lower bounding PN**.

## WHEN IS THE PROBABILITY OF CAUSATION IDENTIFIABLE?

**Theorem:** If Y is **monotonic** in X, then the probabilities of causation PNS, PN and PS are identifiable whenever the **effect of action**  $P(y_x)$  is identifiable, and are given by:

$$PNS = P(y_x) - P(y_{x'})$$

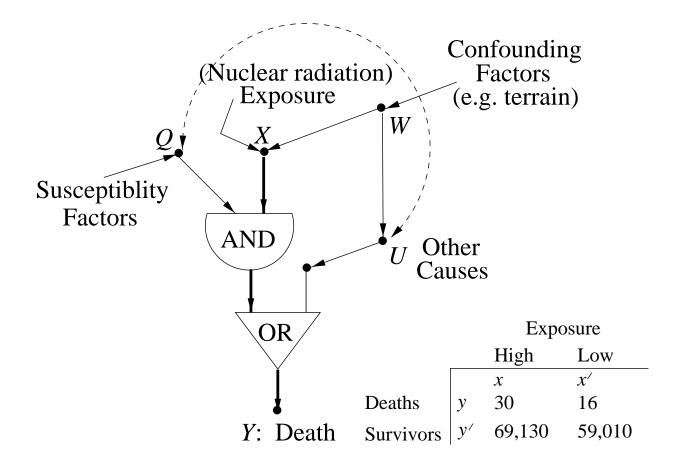
$$PN = \frac{P(y|x) - P(y|x')}{P(y|x)} + \frac{P(y|x') - P(y_{x'})}{P(x,y)}$$

$$PS = \frac{P(y|x) - P(y|x')}{P(y'|x')} + \frac{P(y_x) - P(y|x)}{P(x',y')}$$

Note:  $P(y_x) = P(Y = y | do(X = x))$  is identifiable

- 1. in experimental studies,
- 2. when x and y are not confounded, or
- 3. when x and y are unconfounded through adjustment for covariates (given G(M)).

## EXAMPLE: WHEN IS A DISEASE ATTRIBUTABLE TO EXPOSURE?



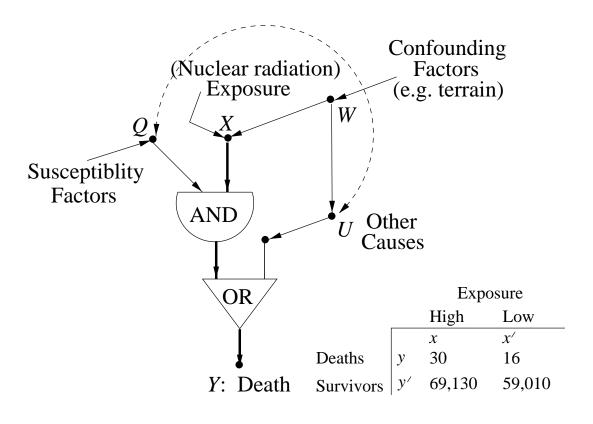
- **Q.** What is the probability PN that a child who died from leukemia after exposure would have survived had he/she not been exposed?
- Α.

$$PN = ERR \stackrel{\triangle}{=} [P(y|x) - P(y|x')]/P(y|x)$$

$$PN = ERR + [P(y|x') - P(y_{x'})]/P(x,y)$$

$$P(y_{x'}) = \sum_{w} P(y|w,x')P(w)$$

# WHEN IS A DISEASE ATTRIBUTABLE TO EXPOSURE? (Cont.)



## Numerical computation (assuming no confounding):

$$PNS = P(y|x) - P(y|x') = \frac{30}{30 + 69,130} - \frac{16}{16 + 59,010}$$
  
= .0001625 (9.45)

$$PN = \frac{PNS}{P(y|x)} = \frac{PNS}{30/(30+69,130)} = .37535$$
 (9.46)

$$PS = \frac{PNS}{1 - P(y|x')} = \frac{PNS}{1 - 16/(16 + 59,010)} = .0001625$$
(9.47)

# LEGAL RESPONSIBILITY FROM EXPERIMENTAL AND NON-EXPERIMENTAL DATA

- A lawsuit is filed against the manufacturer of drug x, charging that the drug is likely to have caused the death of Mr. A, who took the drug to relieve symptom S associated with disease D.
- An experimental study shows only minor increase in death rates among drug users.
- The plaintiff argues, however, that the experimental study is of little relevance to this case, because it represents the effect of the drug on all patients, not on patients like Mr. A who actually died while using drug x.
- Moreover, argues the plaintiff, Mr. A is unique in that he used the drug on his own volition, unlike subjects in the experimental study who took the drug to comply with experimental protocols.

# LEGAL RESPONSIBILITY FROM EXPERIMENTAL AND NON-EXPERIMENTAL DATA

Find PN=P(drug x is the cause of Mr. A's death)

Defendant: 
$$\frac{P(y_x) - P(y_{x'})}{P(y_x)} = \frac{0.016 - 0.014}{0.016} = 0.125$$

**Plaintiff:** Mr. A is not a typical subject, he **chose** the drug (x), and **died** (y).

Defendant: non-experimental data is biased.

Plaintiff: 
$$PN \ge \frac{P(y) - P(y_{x'})}{P(y,x)} = \frac{0.015 - 0.014}{0.001} = 1$$

Jury: Guilty! Combined data tell more than each study alone (monotonicity not assumed).

# TESTABLE IMPLICATIONS OF NO-PREVENTION

• If x cannot prevent y, then every combination of experimental and nonexperimental data, taken from the same population, must satisfy the inequalities:

$$P(x',y) \le P(y_{x'}) \le P(y) \le P(y_x) \le 1 - P(x,y')$$

• If the inequalities are violated, then the data are not drawn from the same population.

### SUMMARY OF RESULTS

- 1. Formal semantics for PN(x,y)
- 2. Exogeneity and monotonicity are needed for

$$ERR = [P(y|x) - P(y|x')]/P(y|x)$$

to be an unbiased estimator of PN(x, y)

- 3. Bounds under exogeneity and NON-monotonicity
- 4. Under NON-exogeneity and monotonicity, experimental data alone are useless. Combined experimental and observational data permit unbiased estimation of PN(x, y)
- 5. Correction for confounding yield unbiased estimation of PN under monotonicity, and bounds on PN without monotonicity.

Reference: Tech Report R-271 (Tian & Pearl 2000)

# PN AS A FUNCTION OF ASSUMPTIONS AND AVAILABLE DATA

Assur	mptions	Data Available				
Exo.	Mono.	Exp.	Non-exp.	Combined		
+	+	ERR	ERR	ERR		
+	_	bounds	bounds	bounds		
_				CERR		
_	_			bounds		

Note: CERR stands for corrected ERR.

# BOUNDING BY LP (After Tian & Pearl 2000, R-271)

### **Parameters:**

$$p_{111} = P(y_x, y_{x'}, x) = P(x, y, y_{x'})$$

$$p_{110} = P(y_x, y_{x'}, x') = P(x', y, y_x)$$

$$p_{101} = P(y_x, y'_{x'}, x) = P(x, y, y'_{x'})$$

$$p_{100} = P(y_x, y'_{x'}, x') = P(x', y', y_x)$$

$$p_{011} = P(y'_x, y_{x'}, x) = P(x, y', y_{x'})$$

$$p_{010} = P(y'_x, y_{x'}, x') = P(x', y, y'_x)$$

$$p_{001} = P(y'_x, y'_{x'}, x) = P(x, y', y'_{x'})$$

$$p_{000} = P(y'_x, y'_{x'}, x') = P(x', y', y'_x)$$

## Maximize (Minimize):

$$PNS = p_{101} + p_{100}$$
 (9.18)  
 $PN = p_{101}/P(x,y)$  (9.19)  
 $PS = p_{100}/P(x',y')$  (9.20)

# BOUNDING BY LP (Cont.)

### **Probabilistic constraints:**

$$\sum_{i=0}^{1} \sum_{j=0}^{1} \sum_{k=0}^{1} p_{ijk} = 1$$

$$p_{ijk} \ge 0 \text{ for } i, j, k \in \{0, 1\}$$
 (9.15)

## Nonexperimental constraints:

$$p_{111} + p_{101} = P(x, y)$$
  
 $p_{011} + p_{001} = P(x, y')$  (9.16)  
 $p_{110} + p_{010} = P(x', y)$ 

## **Experimental constraints:**

$$P(y_x) = p_{111} + p_{110} + p_{101} + p_{100}$$

$$P(y_{x'}) = p_{111} + p_{110} + p_{011} + p_{010}$$
(9.17)

## FROM COUNTERFACTUALS TO PERSONAL DECISION MAKING

	Experimental		Nonex	Nonexperimental	
	$\overline{x}$	x'	$\overline{x}$	x'	
Deaths $(y)$	16	14	2	28	
Survivals $(y')$	984	986	998	972	
	1,000	1,000	1,000	1,000	

Mr. B, survived without drug.
 Would he risk death by starting now?

Nonexperimental data: P(y|x) = 0.002

Experimental data:  $P(y_x) = 0.016$ 

**Answer:** Risk =  $PS = P(y_x|x',y')$ 

Bounded by:

$$\frac{P(y_x) - P(y)}{P(x', y')} \le PS \le \frac{P(y_x) - P(x, y)}{P(x', y')}$$
$$0.002 \le PS \le 0.031$$

Assuming monotonicity (no curing): PS = 0.002

# FROM COUNTERFACTUALS TO TEMPORAL REASONING

### When is

P(future outcome|current action, past conditions and actions)

#### reducible to

 $P(\text{present outcome} \mid \text{hypothetical past action},$  actual past conditions and actions)

Symbolically, when can we assume the equalities

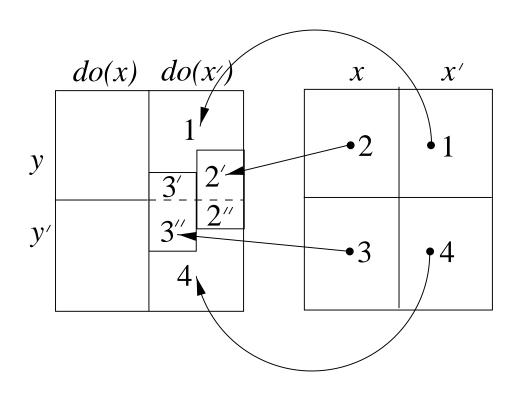
$$P(y(t+1)|do(x(t)), x'(t-1), y(t))$$

$$= P(y_{x(t)}(t+e)|x'(t-e), y'(t))$$

$$= P(y_x|x', y')$$

## HOW DATA CAN UNCOVER THE TRUE CAUSE OF DEATH

	Experimental		Nonexpe	Nonexperimental	
	do(x)	do(x')	$\overline{x}$	x'	
Deaths $(y)$	16	14	2	28	
Survivals $(y')$	984	986	998	972	
	1,000	1,000	1,000	1,000	



$$P(y_{x'}) = \frac{n_1 + n_{2'} + n_{3'}}{n_1 + n_2 + n_3 + n_4} = \frac{14}{1000}$$

$$P(x', y) = \frac{n_1}{n_1 + n_2 + n_3 + n_4} = \frac{28}{2000}$$

$$n'_2 + n'_3 = 0 \Rightarrow P(y_{x'}, x) = 0 \Rightarrow P(y'_{x'}|x) = 1$$