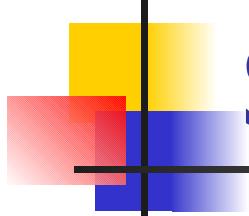


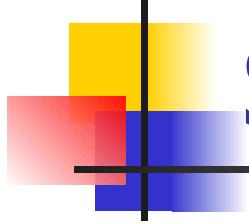
Why there's no cause to randomize

- John Worrall
- Department of Philosophy, Logic and Scientific Method
- London School of Economics



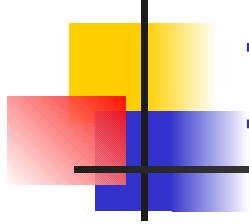
Randomization: the 'Gold Standard'

- *BMJ*, 2001:
- 'Britain has given the world Shakespeare, Newtonian physics, the theory of evolution, parliamentary democracy – and



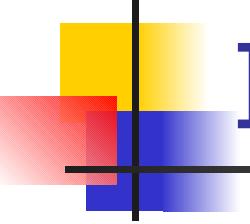
Randomization: the 'Gold Standard'

- *BMJ*, 2001:
- 'Britain has given the world Shakespeare, Newtonian physics, the theory of evolution, parliamentary democracy – and the randomized controlled trial.'



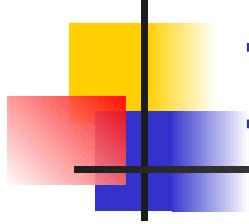
Is it really the gold standard?

- 5 arguments for special epistemic status:



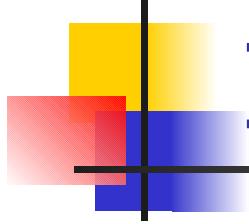
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- 5 arguments for special epistemic status:
- 1. Necessary to underpin the logic of the significance test



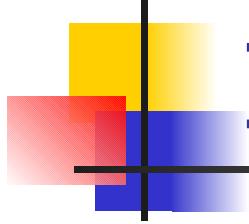
Is it really the gold standard?

- 5 arguments for special epistemic status:
- 1. Necessary to underpin the logic of the significance test
- 2. Controls for all possible confounders – ‘known’ and ‘unknown’.



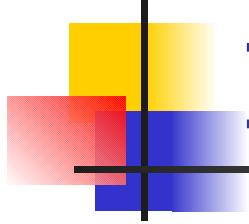
Is it really the gold standard?

- 3. Controls for a known (possible) confounder:
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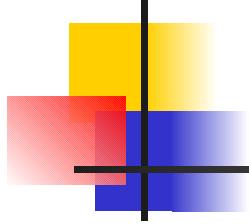
Is it really the gold standard?

- 3. Controls for a known (possible) confounder: 'selection bias'*
- 4. Whatever the epistemic rights and wrongs, it is just *a matter of fact* that non-randomized studies routinely give an exaggeratedly positive result.



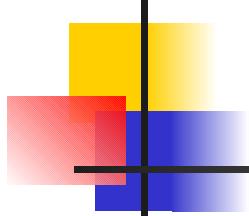
Is it really the gold standard?

- 5. Randomization is necessary if there is to be evidence of genuine *causation* as opposed to 'mere association'.



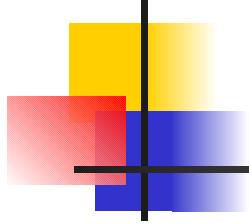
Randomization and causation

- Will concentrate on 5



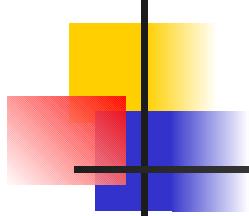
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- Will concentrate on 5 – and in particular Judea Pearl's version of it



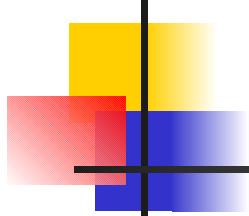
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- Will concentrate on 5 – and in particular Judea Pearl's version of it
- And will contend that Pearl's argument is not convincing



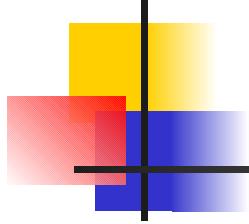
Randomization and causation

- Will concentrate on 5 – and in particular Judea Pearl's version of it
- And will contend that Pearl's argument is not convincing
- Then at the end just make a few brief more positive comments about the way in which I think that progress can best be made



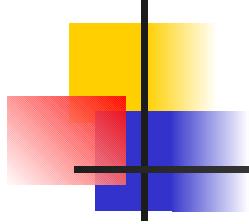
Randomization and causation

- What drives this whole literature is the attempt to ensure that we avoid the *post hoc ergo propter hoc* fallacy.



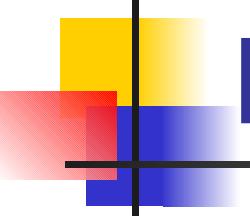
Pearl on causality

- The central Pearlian project:



Pearl on causality

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- Development of algorithms that will produce from probabilistic 'data' a Bayesian network satisfying the Markov condition

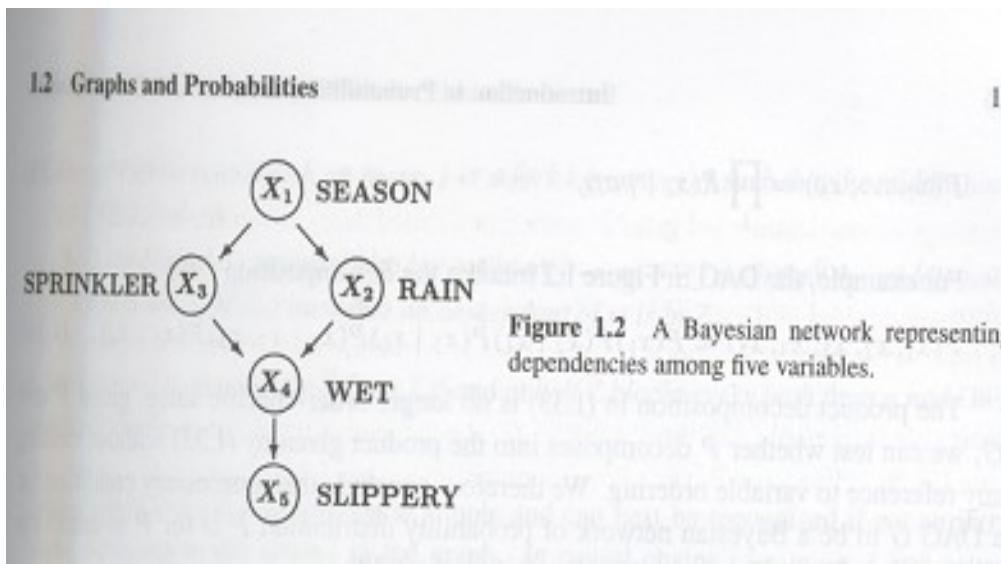


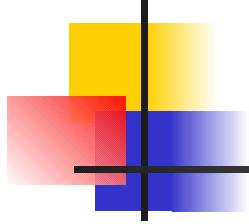
Pearl on causality

- The central Pearlian project:
- Development of algorithms that will produce from probabilistic 'data' a Bayesian network satisfying the Markov condition
- But the networks produced by such an algorithm need not be causal.

Pearl on causality

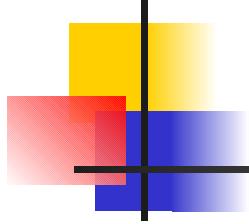
- Pearl's own initial example:





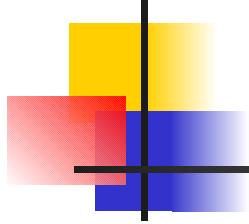
Pearl on causality

- Pearl's own initial example:
- this network will be underwritten not just by 'causal intuition' but also by certain relationships between the *conditional probabilities* that each of the variables possess a particular value (spring, yes, etc), given values of other variables.



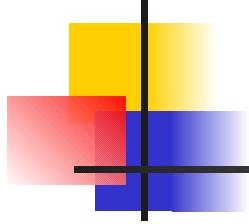
Pearl on causality

- For example, the absence in the graph of a direct connection between X_1 and X_5 , is reflected in the fact that if we already know that the pavement is wet (or dry) we already know the probability that the slipperiness variable takes on the value it does, independently of any of the values of the other variables.



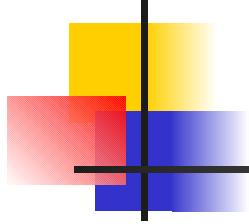
Pearl on causality

- That is, $P(x_5/x_4) = P(x_5/x_1, x_2, x_3, x_4)$



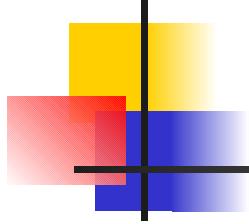
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- That is, $P(x_5/x_4) = P(x_5/x_1, x_2, x_3, x_4)$
- Even though they satisfy various probabilistic constraints nets are not yet guaranteed to be causal



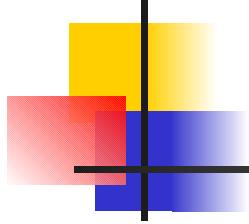
Pearl on causality

- That is, $P(x_5/x_4) = P(x_5/x_1, x_2, x_3, x_4)$
- Even though they satisfy various probabilistic constraints nets are not yet guaranteed to be causal
- The extra factor is centrally to do with 'intervention'



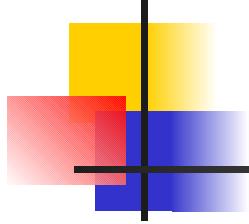
Pearl on causality

- Mutilated network/surgery on network



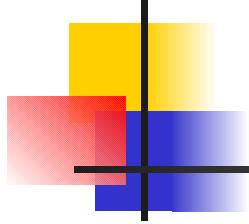
Pearl on causality

- Mutilated network/surgery on network
- Intuitively: you know you have a causal network if you can still predict what will happen after 'surgery'



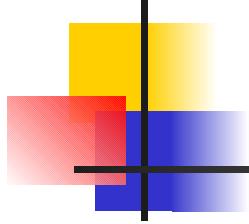
Pearl on causality

- Clearest in deterministic cases:



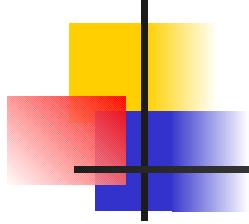
Pearl on causality

- Clearest in deterministic cases:
- Say that a machine takes some rod of any length X as input, doubles that length in its first sub-component and then there's an independent second sub-component that stretches any rod that comes into it by 1m



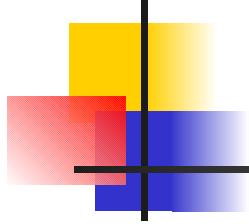
Pearl on causality

- If we 'delete' the normal connection between X and Y (by bypassing the first component) and simply fix the value of Y at y



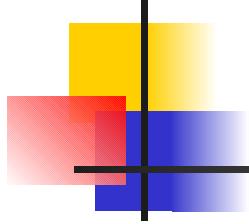
Pearl on causality

- If we 'delete' the normal connection between X and Y (by bypassing the first component) and simply fix the value of Y at y
- Then of course we can still predict what the length of the outgoing rod will be, viz $y + 1$



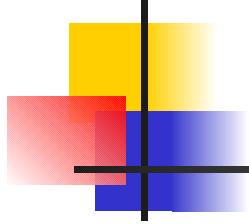
Pearl on causality

- Similar but somewhat more complex in the stochastic case:



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- Suppose we 'mutilate' the sprinkler network by simply deciding to set the sprinkler to 'on'



Pearl on causality

- Similar but somewhat more complex in the stochastic case:
- Suppose we 'mutilate' the sprinkler network by simply deciding to set the sprinkler to 'on'
- Hence

Pearl on causality

1.3 Causal Bayesian Networks

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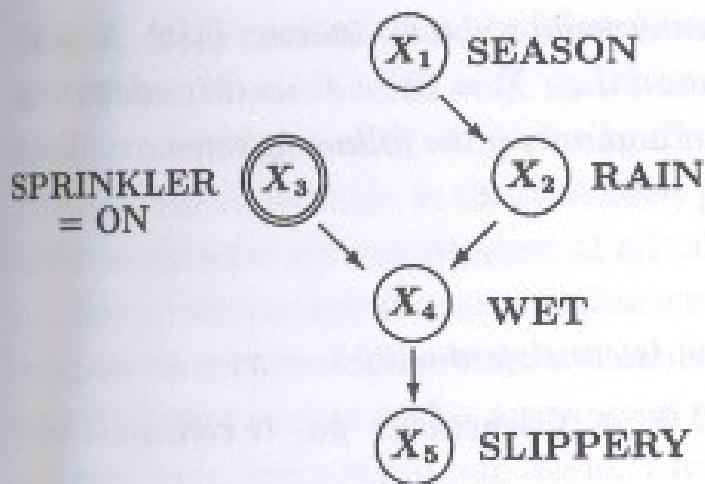
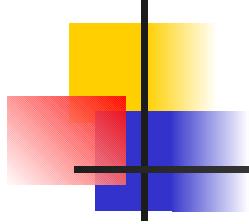
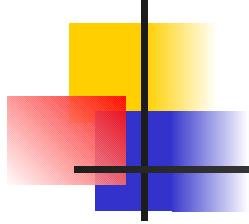


Figure 1.4 Network representation of the action “turning the sprinkler On.”



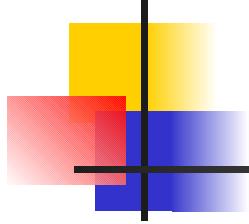
Pearl on causality

- Here what we can predict as a result of 'surgery' is the change in the probability distribution



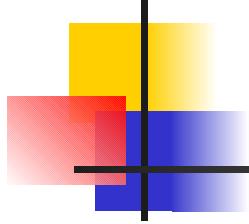
Pearl on causality

- In the 'natural state' the network connections dictate that the joint probability distribution is of the form



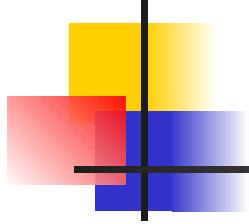
Pearl on causality

- In the 'natural state' the network connections dictate that the joint probability distribution is of the form
- $P(x_1, x_2, x_3, x_4, x_5) = P(x_1)P(x_2/x_1)P(x_3/x_1)P(x_4/x_2, x_3)P(x_5/x_4)$



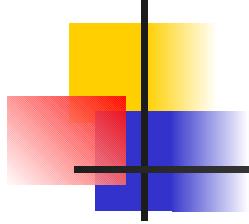
Pearl on causality

- After surgery, it will be



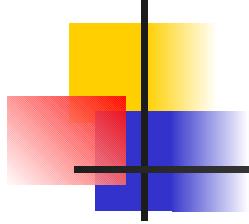
Pearl on causality

- After surgery, it will be
- $P_{x_3=\text{on}}(x_1, x_2, x_4, x_5) = P(x_1)P(x_2/x_1)P(x_4/x_2, X_3=\text{on})P(x_5/x_4)$



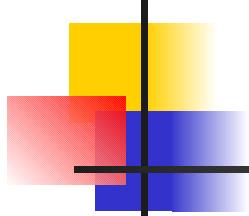
Pearl on causality

- After surgery, it will be
- $P_{x_3=\text{on}}(x_1, x_2, x_4, x_5) = P(x_1)P(x_2/x_1)P(x_4/x_2, X_3=\text{on})P(x_5/x_4)$
- where



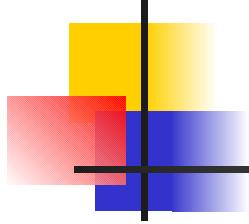
Pearl on causality

“The deletion of the factor $P(x_3/x_1)$ represents the understanding that, whatever relationship existed between seasons and sprinklers prior to the action, that relationship is no longer in effect while we perform the action. Once we physically turn the sprinkler on and keep it on, a new mechanism (in which the season has no say) determines the state of the sprinkler.”



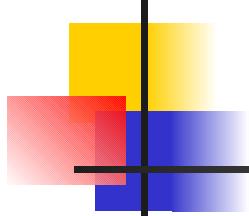
Pearl on causation

- “.. intervention amounts to a surgery on equations (guided by a diagram) and causation means predicting the consequences of such a surgery.”



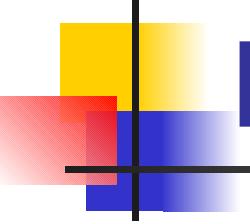
Pearl on causation

- Finally,



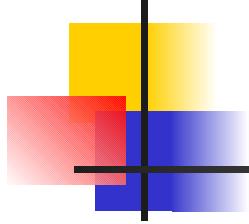
Pearl on causation

- Finally,
- X causes Y if there is a directed line from X to Y in a network that has (a) been inferred from probabilistic ‘data’ in accord with Pearl’s algorithm and (b) is genuinely causal in the sense just explained.



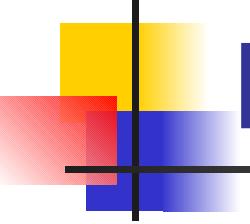
Pearl's argument for RCTs

- “Why do we prefer controlled experiment over uncontrolled studies? Assume we wish to study the effect of some drug treatment on recovery of patients suffering from a given disorder. The mechanism governing the behavior of each patient is similar in structure to the circuit diagram [that Pearl considers earlier – think about the two-component deterministic machine I discussed above].”



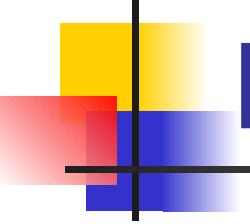
Pearl's argument for RCTs

- Recovery is a function of both the treatment and other factors, such as socio-economic conditions, life style, diet, age, et cetera. ...[In the *diagram* below he collapses all these extra factors into one, just called socioeconomic conditions.] Under uncontrolled conditions, the choice of treatment is up to the patients and may depend on the patients' socioeconomic backgrounds.



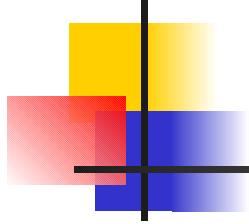
Pearl's argument for RCTs

- This creates a problem, because we can't tell if changes in recovery rates are due to treatment or to those background factors. What we wish to do is compare patients of like backgrounds, and that is precisely what Fisher's *randomized experiment* accomplishes. How? It actually consists of two parts, randomization and intervention.



Pearl's argument for RCTs

- Intervention means that we change the natural behavior of the individual: we separate subjects into two groups, called treatment and control, and we convince the subjects to obey the experimental policy. We assign treatment to some patients who, under normal circumstances, will not seek treatment, and we give placebo to patients who otherwise would receive treatment.



Pearl's argument for RCTs

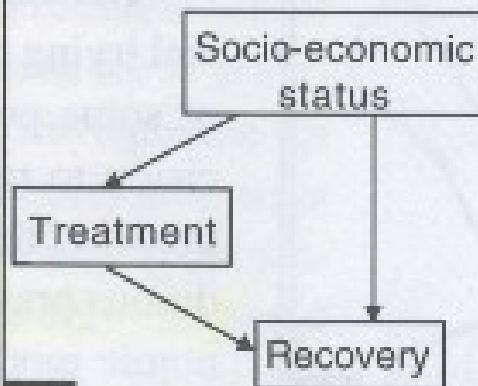
- That, in our new vocabulary, means *surgery* – we are severing one functional link and replacing it with another. Fisher's great insight was that connecting the new link to a random coin flip *guarantees* that the link we wish to break is actually broken. The reason is that a random coin is assumed to be unaffected by anything we can measure on a macroscopic level – including of course, a patient's socioeconomic background.”

Analysis of Pearl on RCTs

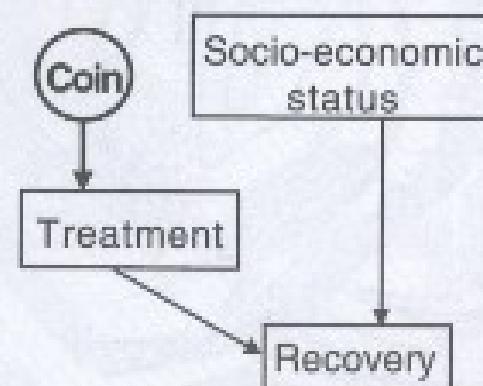
INTERVENTION AS SURGERY (Cont.)

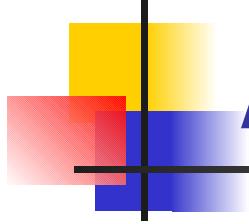
Example 1. Controlled experimentation

Uncontrolled conditions



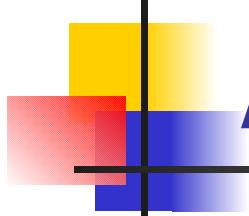
Experimental conditions





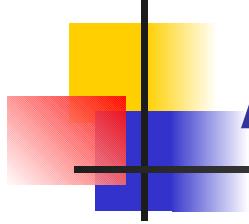
Analysis of Pearl on RCTs

- NB: I am NOT against controls!



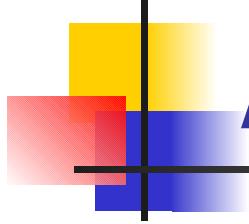
Analysis of Pearl on RCTs

- NB: I am NOT against controls!
- The issue is Pearl's – too ready? – identification of 'controlled' and 'randomized controlled'



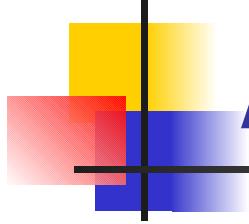
Analysis of Pearl on RCTs

- Once admitted that a real randomization may produce bias in known confounders then...



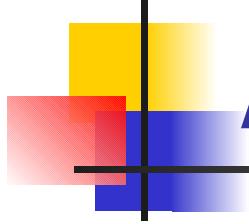
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- Once admitted that a real randomization may produce bias in known confounders then...
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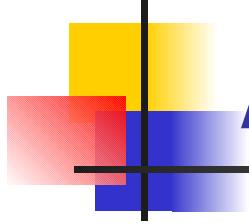
Analysis of Pearl on RCTs

- Once admitted that a real randomization may produce bias in known confounders then...
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- Does it fare any better within Pearl's framework?



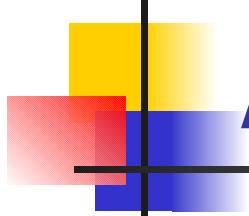
Analysis of Pearl on RCTs

- Once admitted that a real randomization may produce bias in known confounders then...
- Brings us back to an old argument
- Does it fare any better within Pearl's framework?
- Stick to practice and don't switch, implicitly, to the indefinite long run



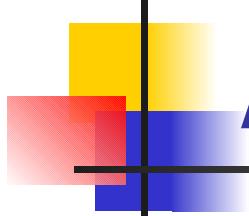
Analysis of Pearl on RCTs

- Clearly a real random allocation *may* be biased



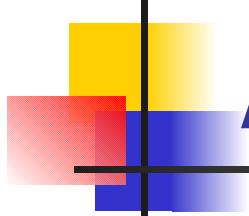
Analysis of Pearl on RCTs

- Clearly a real random allocation *may* be biased
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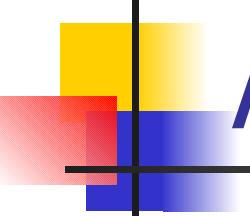
Analysis of Pearl on RCTs

- Clearly a real random allocation *may* be biased
- So talk of a Fisherian “guarantee” is just bluster
- But surely no one could expect a literal guarantee – Pearl should be interpreted as promising only a “probabilistic guarantee”



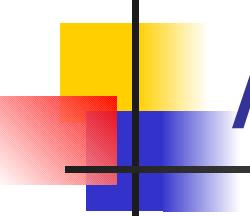
Analysis of Pearl on RCTs

- Two ways to make sense of this – (a) Bayesian and (b) frequentist



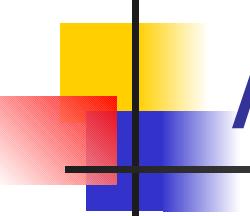
Analysis of Pearl on RCTs

- Two ways to make sense of this – (a) Bayesian and (b) frequentist
- Doesn't work on at least the straightforward Bayesian account



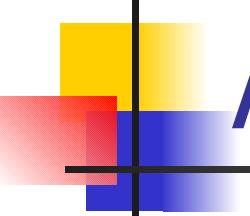
Analysis of Pearl on RCTs

- Two ways to make sense of this – (a) Bayesian and (b) frequentist
- Doesn't work on at least the straightforward Bayesian account
- As for frequentist account ..



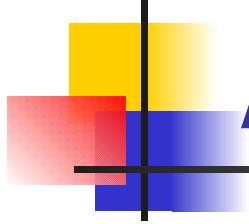
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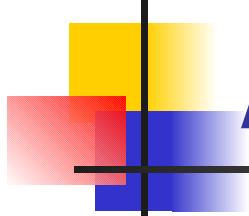
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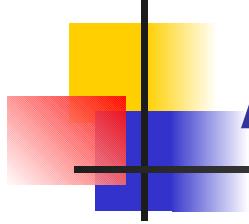
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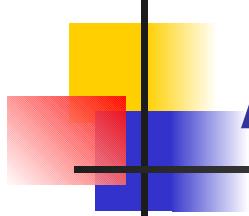
Analysis of Pearl on RCTs

- We are not in the indefinite long run
- And why should what might happen there be of any concern or consolation?



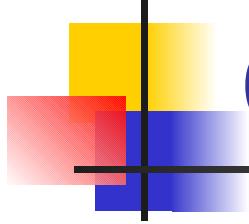
Analysis of Pearl on RCTs

- CONCLUSION:
- Pearl has given no *practical* reason based on causal considerations for randomizing or for automatically giving special weight to results of RCTs

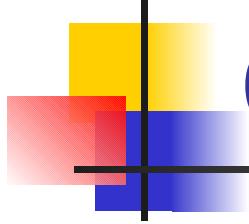


Analysis of Pearl on RCTs

- CONCLUSION:
- Pearl has given no *practical* reason based on causal considerations for randomizing or for automatically giving special weight to results of RCTs
- Take home message

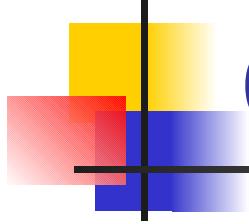


Causality reconsidered



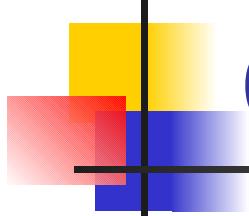
Causality reconsidered

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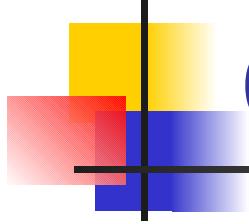
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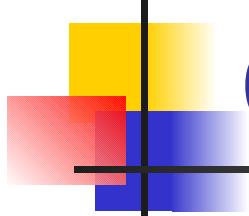
Causality reconsidered

- No such thing as probabilistic causality
- Rather: probabilities (robust under sensible conditionalisations) that *indicate* causal connections
- Randomization is sometimes a help, but never a sine qua non for discovering these



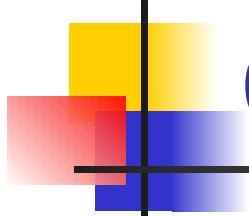
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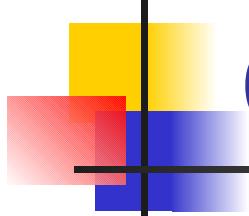
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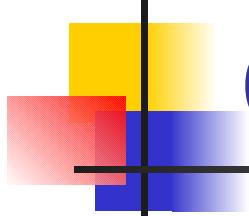
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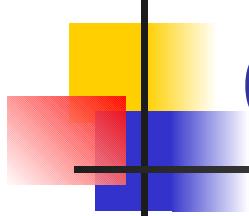
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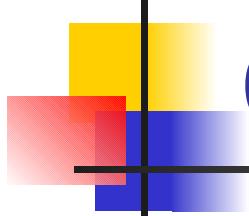
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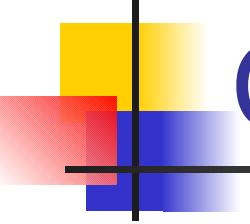
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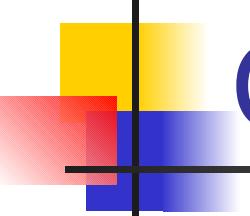
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- Etc



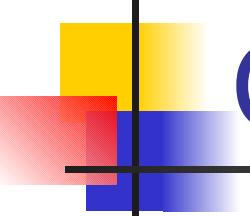
Causality Reconsidered

- “What I do *not* believe [is] that we can usefully lay down some hard-and-fast rules of evidence that must be obeyed before we can accept cause and effect. None of my nine viewpoints can bring indisputable evidence for or against the cause-and-effect hypothesis and none can be required as a *sine qua non*. What they can do, with greater or less strength, is to help us make up our minds on the fundamental question – is there some other way of explaining the set of facts before us, is there any other answer equally, or more, likely than cause and effect.”



Causality Reconsidered

- “No formal tests of significance can answer those questions. Such tests can, and should, remind us of the effects that the play of chance can create, and they will instruct us in the likely magnitude of those effects. Beyond that they contribute nothing to the ‘proof’ of our hypotheses.”



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- Austin Bradford Hill ‘The Environment and Disease:Association or Causation?’ *Proceedings of the Royal Society of Medicine*, **58**, 1965, pp 295-300