13 November 2018



Does Treatment Make a Difference/Have an Effect?



knowledge. understanding.

The power o The value o

RTI(h)(s)

- Typically, analysis of RCT data will look at mean differences between treatment arms on some outcome (oftentimes controlling for other factors) or whether treatment has an effect on that outcome
- If there is no significant difference (or a small difference) between treatment arms on the outcome of interest, does that mean that treatment has no effect?
- · There may be a "causal cascade" of effects such that treatment significantly affects one variable which affects another variable which affects the outcome of interest
 - Thus, treatment may have a significant *indirect* effect on the outcome of interest

1



knowledge. understanding.

The power o The value of

Path Analysis

- A long history:
 - Sewall Wright (1918), a geneticist
 - Herbert Simon (1954), Hubert ("Tad") Blalock (1961, 1964), O.D.
 Duncan (1966), Herbert Costner (1969) in the social sciences
- Recognizes and explicitly models indirect pathways between variables
- "Decomposes" correlations between variables (i.e., total effects of one variable on another) into direct and indirect causal paths





- Simple t-test of QLQ-C30 QoL mean score by treatment: Tx = 59.6 Plac = 66.9 t = -3.71 p < 0.001
- > Small, statistically significant difference (but, there are >600 patients)*
- > But why is treatment arm showing poorer QoL?
- * From the EORTC website (accessed 11 September 2018): "When comparing scores, one should take into account that statistically significant differences do not necessarily imply clinically relevant differences and vice versa. For the QLQ-C30, a change in any scale of at least 10 points is considered to be clinically relevant (Ref: Osoba D, Rodrigues G, Myles J, Zee B, Pater J. Interpreting the significance of changes in health-related quality of life scores. J Clin Oncol 1998; 16 (1): 139-144)."

5		knowledge. understanding.					
Path Analytic/Causal Cascade Approach		$\operatorname{RTI}(h)(s)$					
 We recognize that treatment may influence inte (mediating and/or moderating) variables 	rvenin	g					
	y on logic, theory, mechanism of action, clinical and texperience to consider how treatment plays out:						
– Treatment $ ightarred$ diarrhea $ ightarrow$ fatigue $ ightarrow$ physical functioni	ng → Q	oL					



 Simple t-test of QLQ-C30 Diarrhea mean score (on scale of 0-100) by treatment:

Tx = 41.1 Plac = 5.2 t = 15.82 p < 0.001

Now we begin to see that there may be intervening variables between treatment and QoL that can modify the relationship



Decomposition of Effects



•	Explanatory Variable	Direct Effects	Indirect Effects	Total Effects
Global QoL	Treatment	0.02 n.s.	0.12***	0.13***
Fatigue	Treatment	0.001 n.s.	-0.15***	-0.15***
Diarrhea	Treatment	-0.54***		-0.54***
Fatigue	Diarrhea	0.28***		0.28***

The power of knowledge. The value of understanding

Conclusions



- Considering that many causal relationships are not direct (or that the effects of one variable on another are *not entirely* direct), anticipating this and explicitly modeling these more complex relationships – these "Causal Cascades" – may help demonstrate that a treatment has an effect on an outcome of interest when more "standard" analyses do not show a significant relationship
- Path analysis, with a century-long history across many disciplines, can help tease out these more complex causal relationships, and requires only a minor change in the way analyses are typically done.



Challenge of (and potential solution for) Analysis of Missing Data



Missing Data



- Whether an RCT or observational study, participants commonly have missing observations
 - This becomes more common the longer the trial or due to the disease
- For PRO data, the likelihood of missing data is greater
 - Often secondary or tertiary endpoints
 - Participants are generally responsible for completing them
- Can't we just ignore them?
 - "We still have plenty of data from other participants"
 - "We know death and adverse events were not a huge problem"
 - We'll use last observation carried forward...we've always done that

Recent Acknowledgement and Advances



- In 2008, the FDA requested that the National Research Council commission a panel of experts to produce a report "on appropriate study designs and follow-up methods to reduce missing data and on appropriate statistical methods to address missing data for the analysis of results."
- The US Institute of Medicine, FDA, and European Medicines Agency recommend pre-specification of a primary method to accommodate missing data, as well as sensitivity analyses that make alternative assumptions about the missing data.



 "Use of the LOCF method is therefore likely to misrepresent the results of a trial seriously, and so is not a good choice for primary analysis. In contrast, the MMRM method is unlikely to result in serious misinterpretation, unless the drop-out mechanism is missing not at random (MNAR) and there is substantially unequal drop-out.... Neither method is capable of dealing on its own with trials involving MNAR drop-out mechanisms, for which sensitivity analysis is needed using more complex methods."

Lane, P. Handling drop-out in longitudinal clinical trials: a comparison of the LOCF and MMRM approaches. Pharmaceut. Statist. 2008; 7: 93–106

Use Pattern Mixture Models?



Advantages

- Can accommodate mixed-effects models with repeated measures (MMRM) and growth curve models
- Accounts for distribution of missingness in data

Disadvantages

- Suitable for small number of missing data patterns
- Makes assumptions about missing data patterns that cannot be tested
- Does not account for heterogeneity across missing data patterns

The power of	knowledge.
	understanding.

However, data are seldom missing at random in clinical trials, especially in some therapeutic areas such as oncology AND there can be many missing data patterns



1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 EQSD_BL x x x x x x x EQSD_BL x x x x x x x x x x x x x x x x x x x	x x x	X X 2 X X 2 X X 2 X 2 X 2 X 2 X 2 X 2 X	x x x x x x x x	35 36 x x x x x x x x x x		x x x	940 x x x x x x x x
EQ50 BL * * * * * * * * * * * * * * * * * *	x x x x x	x x x z	x x x x			x	x x x x
русланы ж.	x x x x x	x x x z	x x x x			x	x x x x
EQ5D_1 x x x x x x x x x x x x x x x x x x x	x x x x x	x x x z	x x x x			x	x x x x
	x x x x x	x x x z	x x x x			x	x x
	x x x x x	x 3	хх	x x x x	x		
EQ5D_4 x x x x x x x x x x x x x x x x x x x	x x	X J	хх	xx	ŝ		
	x x	1					x
EQ5D_501 x x x x x x x x x x x x x x x x x x x	x			x x		~	×
26505502 x x x x x x x x x x x x x x x 2025503 x x x x x x x x 2025503 x x x x x x x x x x x x x x x x x x x			xx	x	×		â
ВОБЪ 502 жихихих хих хихихих вобъ 503 кихи хихихи вобъ 503 кихи хихихихихи вобъ 504 кихихих вобъ 504 кихихих вобъ 504 кихихих вобъ 504 кихихих вобъ 505 кихихихихихихихихихихихихихихихихихихих				x			
EQ5D 504 x x x x x x x x x x x EQ5D 504 x x x x x			хх		х		хх
EQ5D_505 x x x x x x x x x x x x x x x x x x	x			х	x	х	x
EQSD_505 x x x x x x x x x EQSD_505 x x x EQSD_506 x x x x x x x x EQSD_506 x x x	x	3	x	x	х		x
	69 70 7:	71 72 7	3 74	75 76	6 77	78 '	79 80
EQ5D_BL × × × × × × × × × × × × × × × × × × ×	X X X	x x :	хх	хх	к х	х	хх
EQSD_1 x x x x x x x x x x x x x x x x x x EQSD_1 EQSD_2 x x x x x x x x x x x x x x x x x x x							
EQ5D_2 x x x x x x x x x x x x x x x x x x x	x x x	x x	x x	x >	x x	x	x x
EQ5D 3 X X X X X X X X X X X X FOSD 3 Y Y Y Y Y Y Y Y Y Y Y Y Y Y Y Y Y Y	x x x	xx	x x	xx	x x	x	x x
EQ5D_4 x x x x x x x x x x x x x EQ5D_4 x x x x x x x x	x x x	xx	xx	xx	k x	x	xx
EQ5D_4 x x x x x x x x x x x x x x EQ5D_4 x x x x x x x x x x x x x x x x x x x	xxx	xx	x				
EQ5D 502 X X X X X X X X X X X X X X X X X X X			- x	x x	x x	x	x x
EQ5D_503 X X X X X X X X X X X X X EQ5D_503 X X X X		x x	÷			÷	x
EQ5D_504 x x x x x x x x x x x EQ5D_504 x x x x x		xx	ŝ	xx	k x	^	x
EQ5D 505 X X X X X X POED EQ5		• •	x	x	· ·		x
EQ5D_506 x x x x x EQ5D_506 x x x x x	x	x	x	× ,	ĸ	x	x



Solution? Identify Latent Classes based on *Patterns of Change for every Individual Patient – Extended* Pattern Mixture Models

Can help us understand if those who have worsening PRO scores are also more likely to drop out. That is, the trajectory of change in PRO score helps 'inform" us about "why" the patient dropped out.



The power of knowledge. The value of understanding







How does accounting for missing data patterns inform overall survival? (Treatment arm only)







Conclusions

- Examining patterns of missing data and identifying differential change in patient-reported outcomes scores can provide clues about differential survival
- We have now taken this differential missing data on the PROs and used it to inform us about the censoring of survival

Voting Questions



Instructions: Go into ISPOR app and select 'Take a Poll' or to myispor.cnf.io and select session W10

Can you think of instances when the path analytic way of thinking about or analyzing data could be beneficial?

- a) Yes
- b) No

The power of <mark>knowledge.</mark> The value of <mark>understanding</mark>

Live Content Slide When playing as a slideshow, this slide will display live content

Poll: Can you think of instances when the path analytic way of thinking about or analyzing data could be beneficial

Voting Questions



Instructions: Go into ISPOR app and select 'Take a Poll' or to myispor.cnf.io and select session W10

Given our discussion of addressing missing data in the analysis of clinical trial data, how likely is it that you would encourage alternative approaches to LOCF being built into analysis plans?

- a) Very unlikely
- b) Somewhat unlikely
- c) Somewhat likely
- d) Very likely

23 The power of knowledge. The value of understanding

> *Live Content Slide* When playing as a slideshow, this slide will display live content

Poll: Given our discussion of addressing missing data in the analysis of clinical trial data, how likely is it that you would encourage alternative approaches to LOCF being built into analysis plans?





25

he power of knowledge. he value of understanding.