Electronic Health Records used to derive Control Arms for Single-Arm oncology trials: Proof of concept using RCT's in lung cancer

Gonzalo Duran-Pacheco on behalf of: Gillis Carrigan, Samuel Whipple, William B. Capra, Michael D. Taylor, Michael Lu, Brandon Arnieri, Jeffrey S. Brown, Amy Abernethy, Ryan Copping and Kenneth J. Rothman Basel Biometrics Society, May 2019

Disclaimer



- I'll present this work on behalf of those who actually conducted this study
- I might not be able to answer some questions, I'll do my very best
- EC instead of SC



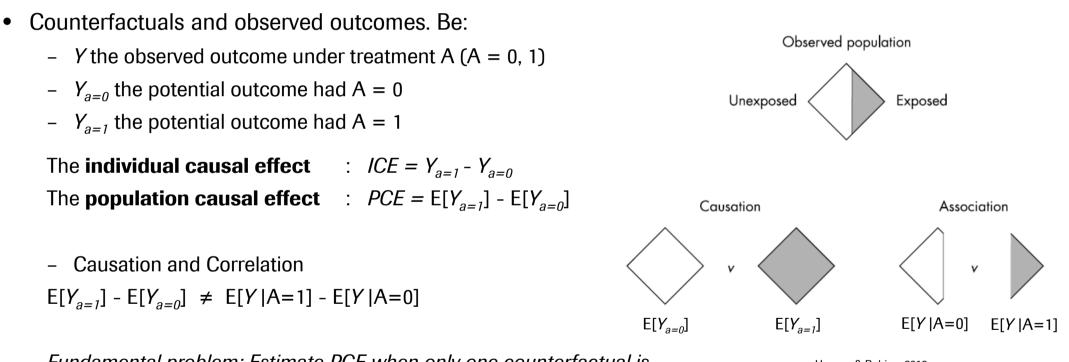
Content

- The RCT as a gold standard for the study of causation
- External controls for lung cancer trials using EHR
 - Background
 - Process
 - Results
 - Conclusions and next steps

The Randomized Controlled Trial as a Gold Standard for the study of causation

Causal Effects of Treatment A





Fundamental problem: Estimate PCE when only one counterfactual is observed

Hernan & Robins, 2018

Causal Effects and RCT



- Under which conditions is $E[Y_{a=1}] E[Y_{a=0}] = E[Y|A=1] E[Y|A=0]?$
 - Exchangeability: $Y_a \perp A$
 - Consistency: $Y_a = Y$ when a subject received treatment A = a
 - Positivity: $f_{A|L}(a|l) > 0$ if $f_L(l) \neq 0$ (with confounding factors L)
- Under ideal RCT conditions (i.e. full compliance, no loss to-follow-up, blind assignment)
 - $E[Y|A=1] = E[Y_a | A=1]$ a for given a

 $= \mathbb{E}[Y_a \mid A=0]$ $= \mathbb{E}[Y_a]$ $\mathbb{E}[Y_a] \perp A \quad \forall a \ (0, 1)$ $\therefore \text{ Equation above holds}$

Roch

Conditional exchangeability

 $E[Y_a] \perp A \mid L$ Same as above but ۲

where L is a vector of covariates

 $E[Y_{a=1}] - E[Y_{a=0}] | L = E[Y|A=1] - E[Y|A=0] | L$

- Example: ullet
 - The Propensity Scores Theorem:
 - Be treatment A wit values 0 and 1
 - The propensity of "choosing" treatment given covariates L: PS(L) = P(A=a | L)

If $Y_a \perp A \mid \mathbf{L}$ (Conditional Independence Assumption)

then $Y_a \perp A \mid PS(L)$ (PST)

0000000 **External Controls for Lung Cancer RCTs using EHR: Background** 0.0 -----....

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Background

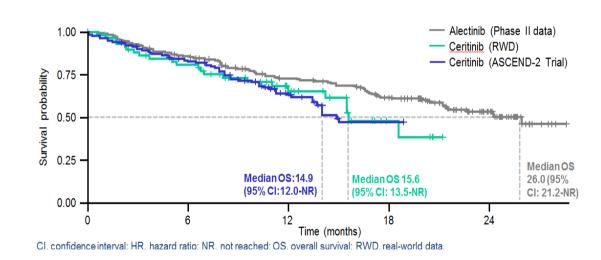
- Accelerated or breakthrough regulatory approval based on single-arm trials often
 - standard-of-care control arm is not included, challenges in interpretation of efficacy
- External controls (EC) derived from electronic health record (EHR) databases may provide an additional context for interpretation



- Curated EHR datasets are now large enough, with sufficient clinical detail, to create contemporaneous EC groups
- The Flatiron Health database is a longitudinal, demographically and geographically diverse database derived from EHR data
 - 260 community-based cancer treatment clinics and 3 academic networks, > 2 million active cancer patients in the US
 - High quality mortality data for lung cancer benchmarked against the US National Death Index

Background

• Efforts towards EC

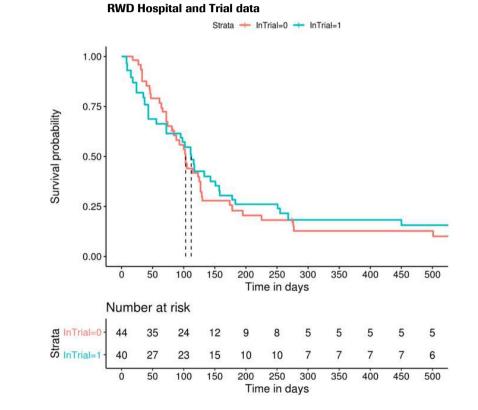


Research Article

Comparative effectiveness from a single-arm trial and real-world data: alectinib versus ceritinib

tesska Davier**, Michael Martinec*, Paul Delmar*, Mathieu Couder*, Wa Gordongar*, Sophie Golding*, Reynaldo Martina* de Grany Crane* Norh-Inducti List, Filteran Was, Birt Fahr, Walero Gealer, Car, AZ 1790, UK Filterhanel Allerito List, Insulger Balanciar Calles, Tenze Filterhanel Allerito List, Insulger Balanciar Calles, Tenze Alleritor for compendance List. 45: 733-7347. 2022. Serial, Balanciarchine norm.







Objective



• To assess how closely results from RCTs on aNSCLC could be replicated by substituting EHR-based EC groups as the comparator

External Controls for Lung Cancer RCTs using EHR: Cohorts creation and Analysis

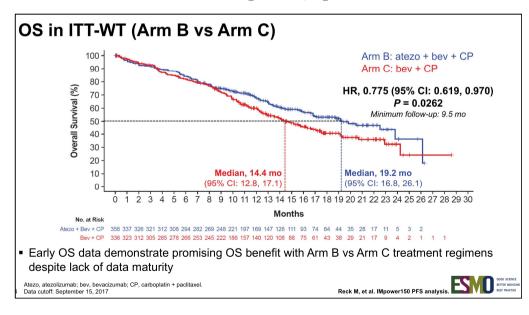
Trials selection



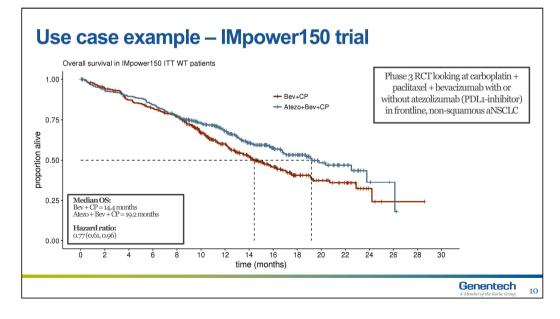
- Study on all Roche-sponsored aNSCLC RCT meeting the following:
 - a) First patient enrolled on or after January 1, 2011
 - b) mOS attained, findings presented in a journal or at a congress, by March 31, 2018
 - c) including at least one US study site
 - d) in the case of a biomarker-defined study population, availability of the biomarker within the curated EHR dataset

Retrieve patient level data and verify trial results

- Verify RCT results published in public forums
 - BL and demographics (Table 1)
 - Main results



OS results for IMpower150 presented at ESMO



Reproduced results with received data cut



Roche

Review trial protocol and identify criteria to apply

- Done cross-functionally with the study team
- Go through the I/E criteria one by one, flagging those criteria which can be applied to the EC cohort
 - Not everything will be able to be applied
 - e.g. life expectancy, comorbidities, other medications, pregnancy, etc..
 - For transparency, those criteria that are unable to be applied should be called out
- We found it especially beneficial to sit down with clinical scientists to review certain criteria and decide how best to apply to Flatiron
 - Often some "translation" needs to occur between protocol and Flatiron (e.g. staging)

Build EC cohort



- Select patients from the EHR cohort that received standard-of-care treatment as in the trial
- Apply RCT I/E criteria available in the EHR to select EHR-based controls comparable in terms of demographic and clinical characteristics with RCT patients
 - Attrition rates displayed at each step
 - Alternatively, make each criteria a flag in your dataset so that you can easily turn them on/off in different orders
 - Some criteria will be straightforward
 - Therapy of interest, Histology, Age
 - For others (ECOG and lab values), we've developed some business rules to alleviate issues like high levels of missingness

Data Analysis



- **Primary endpoint:** time from randomization or treatment initiation (EHR) to death (OS)
- Statistical Analysis:
 - Proportional hazards cox model used to estimate treatment effects (HR) comparing the experimental trial arms with EC
 - PS obtained: Probability of being in the trial treatment arm rather than in the EC given L
 - L = age, gender, race, smoking history, histology, disease stage at initial diagnosis, time from initial diagnosis to either the start of treatment (EHR data) or randomization (trial data)
 - L derived from discussions with subject matter experts



Data Analysis

• PS Methods applied:

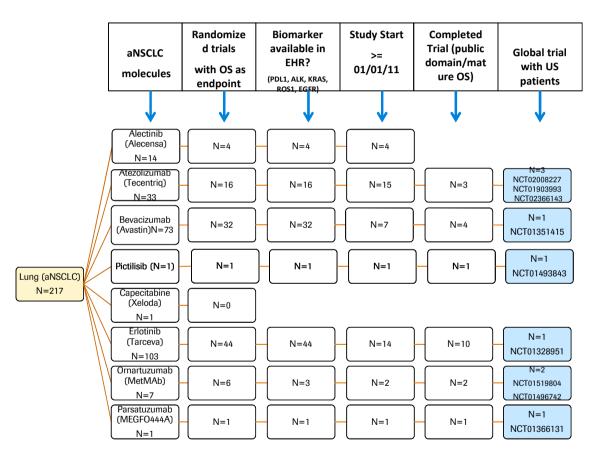
- PS stratification
- IPTW (ATE, ATT)
- Cox PH adjusting directly for L
- Weights stabilization: trimming/truncation
- Sensitivity Analysis





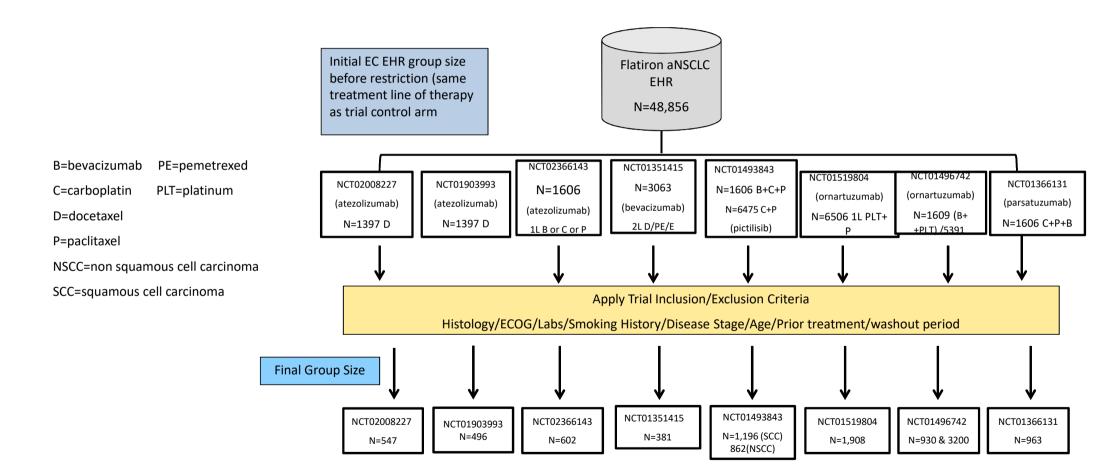
Trial selection

- From 217 RCT (8 drugs) to 9 eligible RCT
- 11 experimental arms











Trial Results

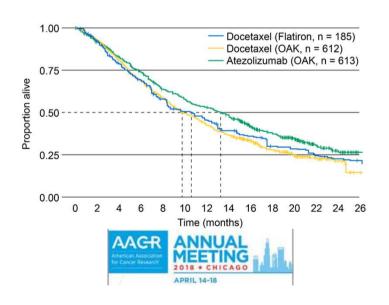
• Treatment effect estimates

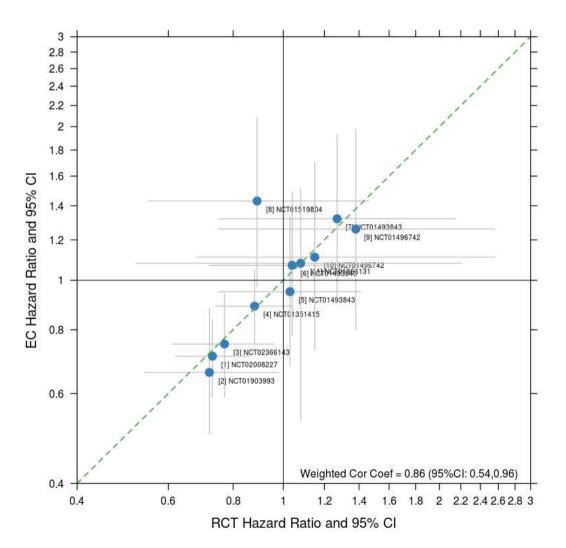
	_	RCT					_	
	_	Experimental		Control			EC adjusted HR	Difference
		Ν	Events	Ν	Events	RCT HR (95% CI)	(95% CI)	Difference
[1] NCT02008227	_	425	271	425	298	0.73 (0.62, 0.86)	0.71 (0.59, 0.84)	0.028 (-0.132, 0.188)
[2] NCT01903993		144	78	143	95	0.72 (0.54, 0.98)	0.66 (0.50, 0.88)	0.087 (-0.176, 0.350)
[3] NCT02366143		356	144	336	166	0.77 (0.61, 0.96)	0.75 (0.59, 0.94)	0.026 (-0.179, 0.231)
[4] NCT01351415		245	194	240	193	0.88 (0.74, 1.04)	0.89 (0.75, 1.05)	-0.011 (-0.202, 0.179)
[5] NCT01493843:	Arm A vs. B	126	79	125	60	1.03 (0.75, 1.41)	0.95 (0.68, 1.33)	0.081 (-0.175, 0.337)
[6] NCT01493843:	Arm C vs. D	79	59	79	43	1.04 (0.72, 1.50)	1.07 (0.78, 1.49)	-0.028 (-0.319, 0.262)
[7] NCT01493843:	Arm E vs. F	62	42	30	13	1.27 (0.75, 2.15)	1.32 (0.90, 1.93)	-0.039 (-0.389, 0.312)
[8] NCT01519804		55	36	54	33	0.89 (0.55, 1.46)	1.43 (0.97, 2.09)	-0.474 (-0.835, -0.114)
[9] NCT01496742:	Cohort 1	69	32	70	29	1.38 (0.75, 2.56)	1.26 (0.80, 1.97)	0.091 (-0.310, 0.492)
[10] NCT01496742:	Cohort 2	59	37	61	36	1.15 (0.68, 2.56)	1.11 (0.73, 1.70)	0.035 (-0.332, 0.403)
[11] NCT01366131		52	24	52	18	1.08 (0.52,2.21)	0.90 (0.53, 1.51)	0.182 (-0.276, 0.640)

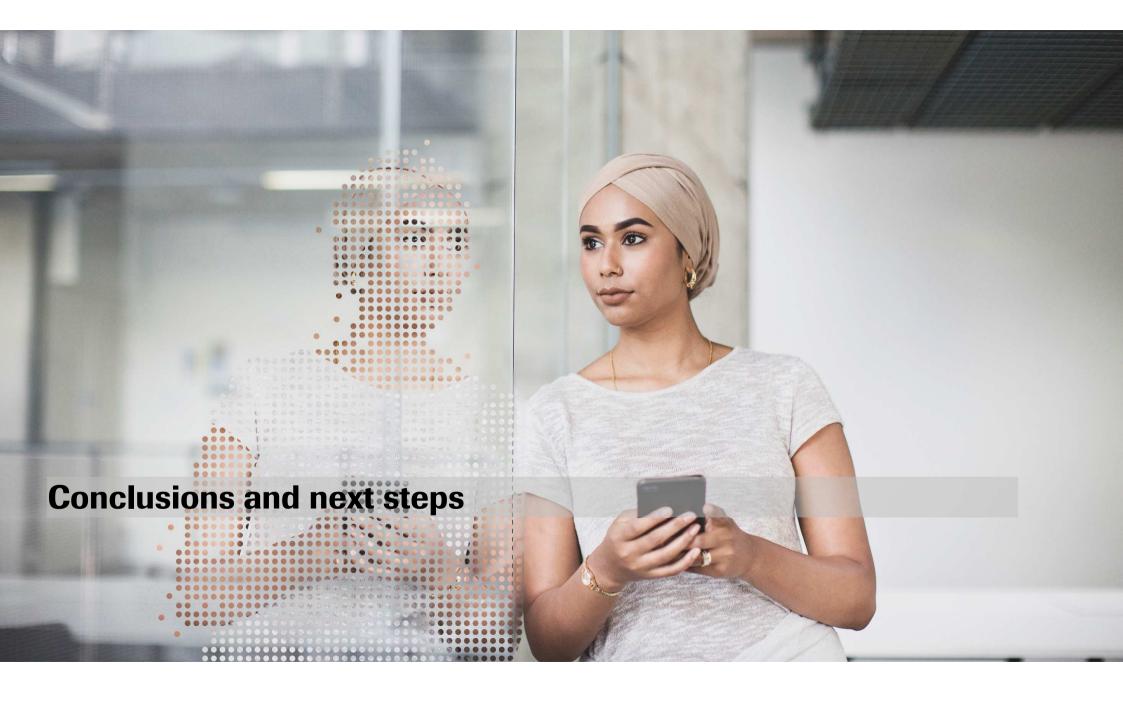
Treatment effects with EC



- Trials results replicated:
 - Treatment effect estimates, except for one trial
 - Conclusions from statistical tests (H0: logHR = 0)







Conclusions



• Properly selected and adjusted control arms from high quality contemporaneous EHR data could be used to replicate results from RCT in aNSCLC



Next steps

- Fully understand why and when EC don't work
- Methods to optimize and validate EC for single arm trials
 - Estimands & PS methods
 - Unmeasured confounding
 - rwPFS and rwOS
 - Bayesian methods
- Understand data
- Apply learnings and do the same in other tumor types (Breast, mCRC)
- Hybrid Controls (HC)



Doing now what patients need next