

# Considerations for Developing External Control Arm from Real-World Data

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*Roche Auditorium, Viaduktstrasse, Basel*



## Regulatory perspective

Flatiron's real-world evidence generation platform

## Real-world control



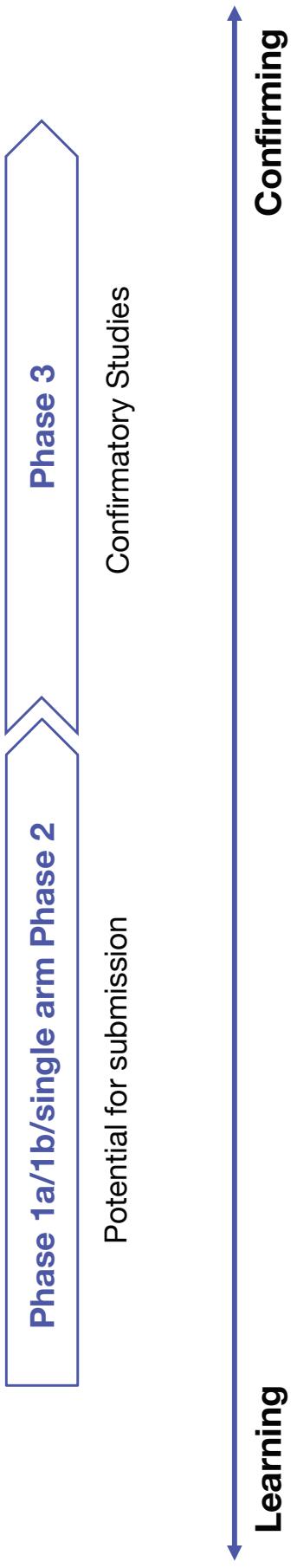
# Regulatory perspective

# Change in oncology drug development paradigm

## Expanded Phase I or Single Arm Phase II often leads straight to Pivotal



- “the desire to provide earlier access to highly effective drugs should encourage further use of **seamless expansion cohort**”
- “greater attention to statistical **rationale and analysis plan**, more careful **selection of drugs** to be studied in this fashion”



## Number of drug approvals in oncology has increased with availability of novel therapies

- In 2018 over 20 drugs were recommended for approval by EMA<sup>1</sup>
- In the US ~50 new cancer drugs or combinations were approved in 2018, compared with 2 in 2005<sup>2</sup>
- FDA granted 25 Breakthrough Designations
  - Over last 25 years: 72% of AAs were from single-arm trials<sup>3</sup>

<sup>1</sup>[https://www.ema.europa.eu/en/documents/report/human-medicines-highlights-2018\\_en.pdf](https://www.ema.europa.eu/en/documents/report/human-medicines-highlights-2018_en.pdf)

<sup>2</sup>Blumenthal GM, Pazdur R. Approvals in 2018: a histologyagnostic new molecular entity, novel end points and real-time review. Nat Rev Clin Oncol. 2019 Mar;16(3):139-141

<sup>3</sup>JCO Editorial, Volume 36, June 20, 2018

<sup>4</sup>Blood Cancer Journal (2016) 6, e473; doi:10.1038/bcj.2016.84

<sup>5</sup>CHOICE OF CONTROL GROUP AND RELATED ISSUES IN CLINICAL TRIALS E10

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- Regulatory Guidance recognized option of using external historical controls<sup>4, 5</sup>
- Disease is rare and has no satisfactory treatment
  - New treatment appears very promising based on preliminary data

# Recent work by EMA exploring the use of RWD in regulatory decisions



EUROPEAN MEDICINES AGENCY  
SCIENCE MEDICINES HEALTH

5 November 2018  
EMA/763513/2018

**Discussion paper:**  
**Use of patient disease registries for regulatory purposes – methodological and operational considerations**

The Cross-Committee Task Force on Patient Registries

EMA released a discussion paper on methodological and operational considerations in the use of patient disease registries for regulatory purposes.

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EUROPEAN MEDICINES AGENCY  
SCIENCE MEDICINES HEALTH

13 February 2019  
EMA/105321/2019

**HMA-EMA Joint Big Data Taskforce**  
Summary report

e.g. “development of a framework to articulate for what questions and contexts RWE may be acceptable across the product life cycle”

“strongly support exploration of novel analytics approaches”

See website for contact details  
Haus of Medicines Agencies www.homa.eu  
European Medicines Agency is an agency of the European Union



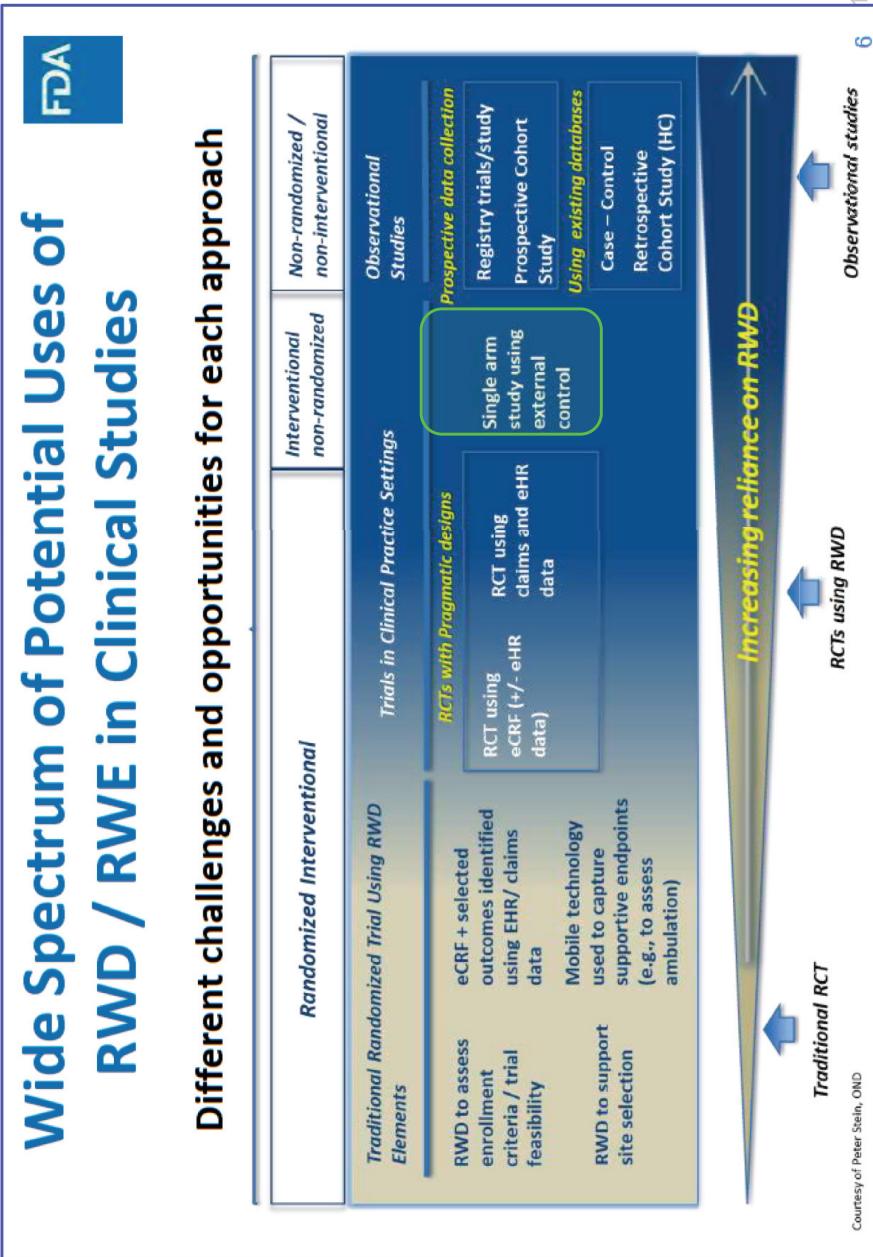
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**EMA Regulatory Science to 2025**  
Strategic reflection

Creating a sustainable, quality assured, flexible framework delivering rapid access to and analysis of representative, longitudinal RWD throughout a product's lifecycle

See website for contact details  
Haus of Medicines Agencies www.homa.eu  
European Medicines Agency is an agency of the European Union

FDA has legislative mandate to explore IF and WHEN RWE may support new indications (approved drugs)/post marketing requirements

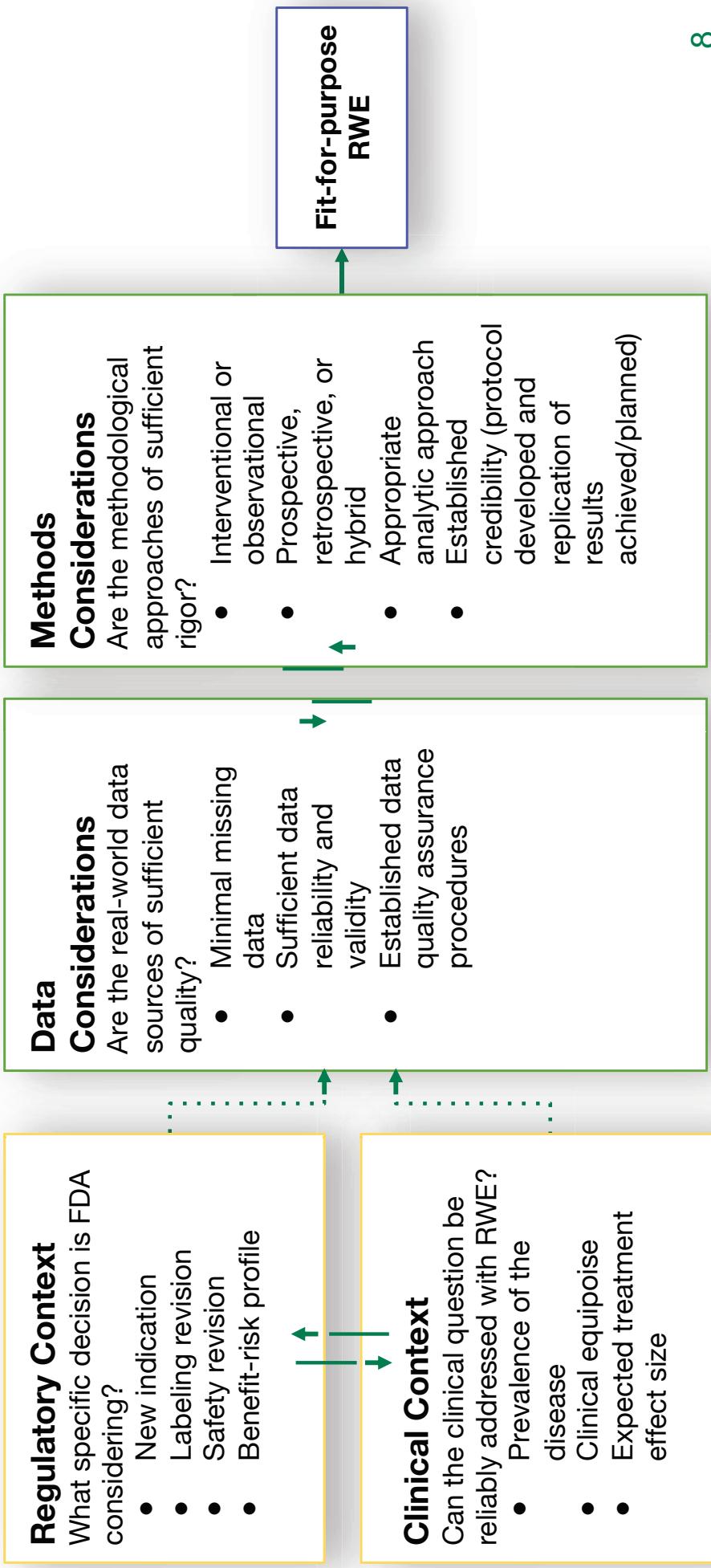


<https://www.fda.gov/downloads/ScienceResearch/SpecialTopics/RealWorldEvidence/UCM627769.pdf>



Source: Dr. Jacqueline Corrigan-Curay (Director Office of Medical Policy), FDA, "Framework for FDA's Real-World Evidence Program", webinar on March 15, 2019

# Considerations for Generating RWE Fit for Regulatory Purposes: Duke Margolis White Paper



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“Characterizing RWD Quality and Relevancy for Regulatory Purposes” Presented at the 2018 Duke Margolis RWE Collaborative, October 1, 2018. [https://healthpolicy.duke.edu/sites/default/files/atoms/files/characterizing\\_rwd.pdf](https://healthpolicy.duke.edu/sites/default/files/atoms/files/characterizing_rwd.pdf)

# Considerations for Generating RWE Fit for Regulatory Purposes: Duke Margolis White Paper

## Regulatory Context

What specific decision is FDA considering?

- New indication

## Data Considerations

Are the real-world data sources of sufficient quality?

- Minimal missing data
- Sufficient data reliability and validity
- Established data quality assurance procedures

## Methods Considerations

Are the methodological approaches of sufficient rigor?

- Interventional or observational
- Prospective, retrospective, or hybrid
- Appropriate analytic approach
- Established credibility (protocol developed and replication of results achieved/planned)

## Example of Real-world Control:

- Rare diseases
- New indication
- Expected large treatment benefit

- Clinical equipoise
- Expected treatment effect size

Fit-for-purpose  
RWE

# Flatiron's real-world evidence generation platform



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# Rapid adoption of EHRs in oncology

President Obama to Sign ARRA's  
HITECH provisions Tuesday,  
February 17, 2009, in Denver, CO

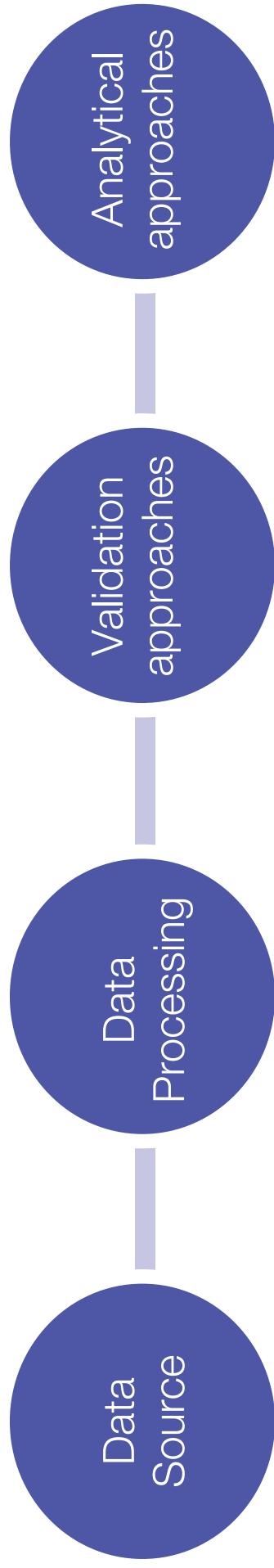
Adoption of EHRs in  
Oncology clinics go from  
~10% → 95%



The Senate joined the House on Friday evening, February 13, 2009, in passing the American Recovery and Reinvestment Act, which includes provisions relating to Health Information Technology. Title XIII of Division A and Title IV of Division B together are known as the "Health Information Technology for Economic and Clinical Health Act" or the "HITECH Act." We will be highlighting attributes of the HITECH Act  
Source: [www.flatiron.com](http://www.flatiron.com) © Flatiron Health 2017

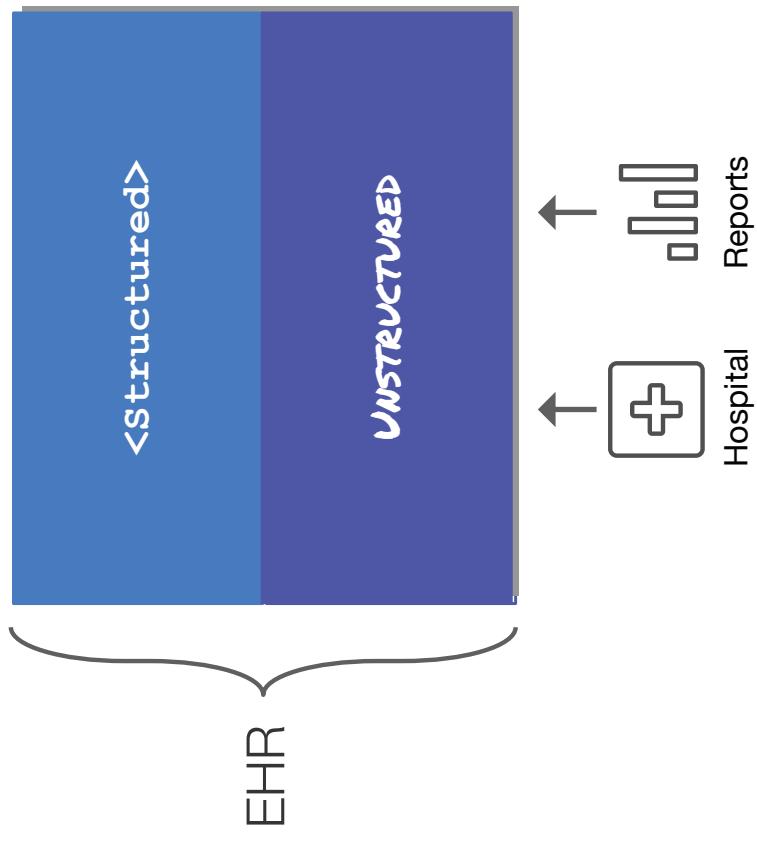
Recent adoption in US parallels UK ambition to shift to a “paperless NHS”

# .....Journey to Research/Regulatory Grade Real-World Evidence

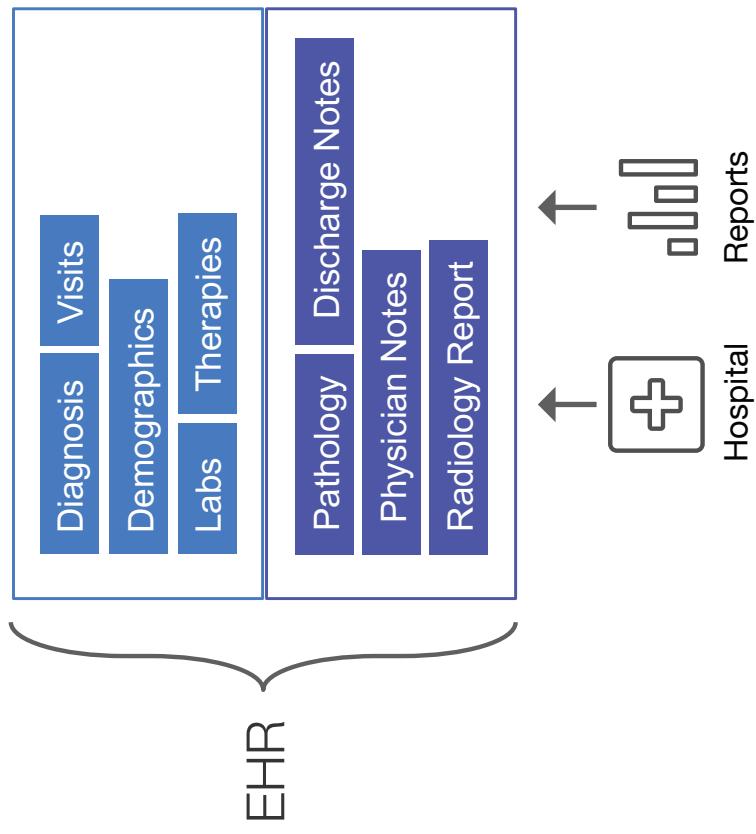


“RWE is derived from RWD through the  
application of research methods”

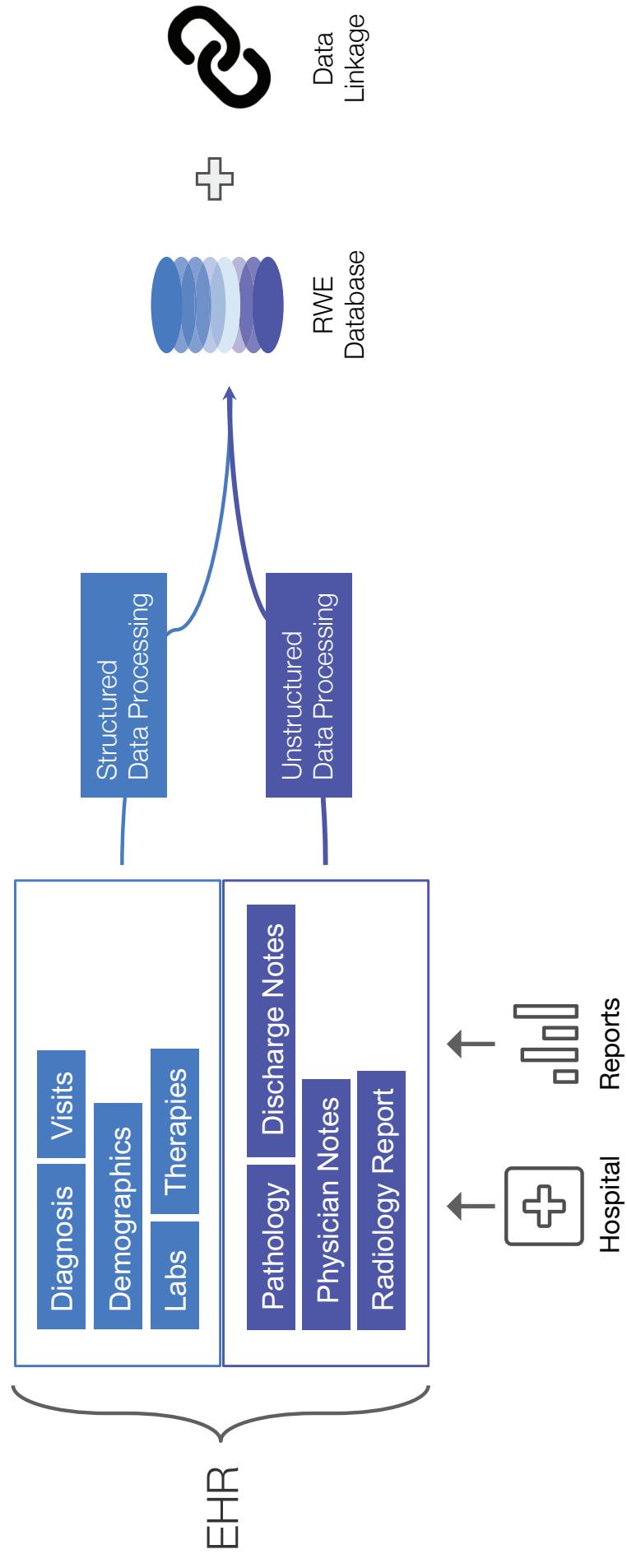
# Data source and curation



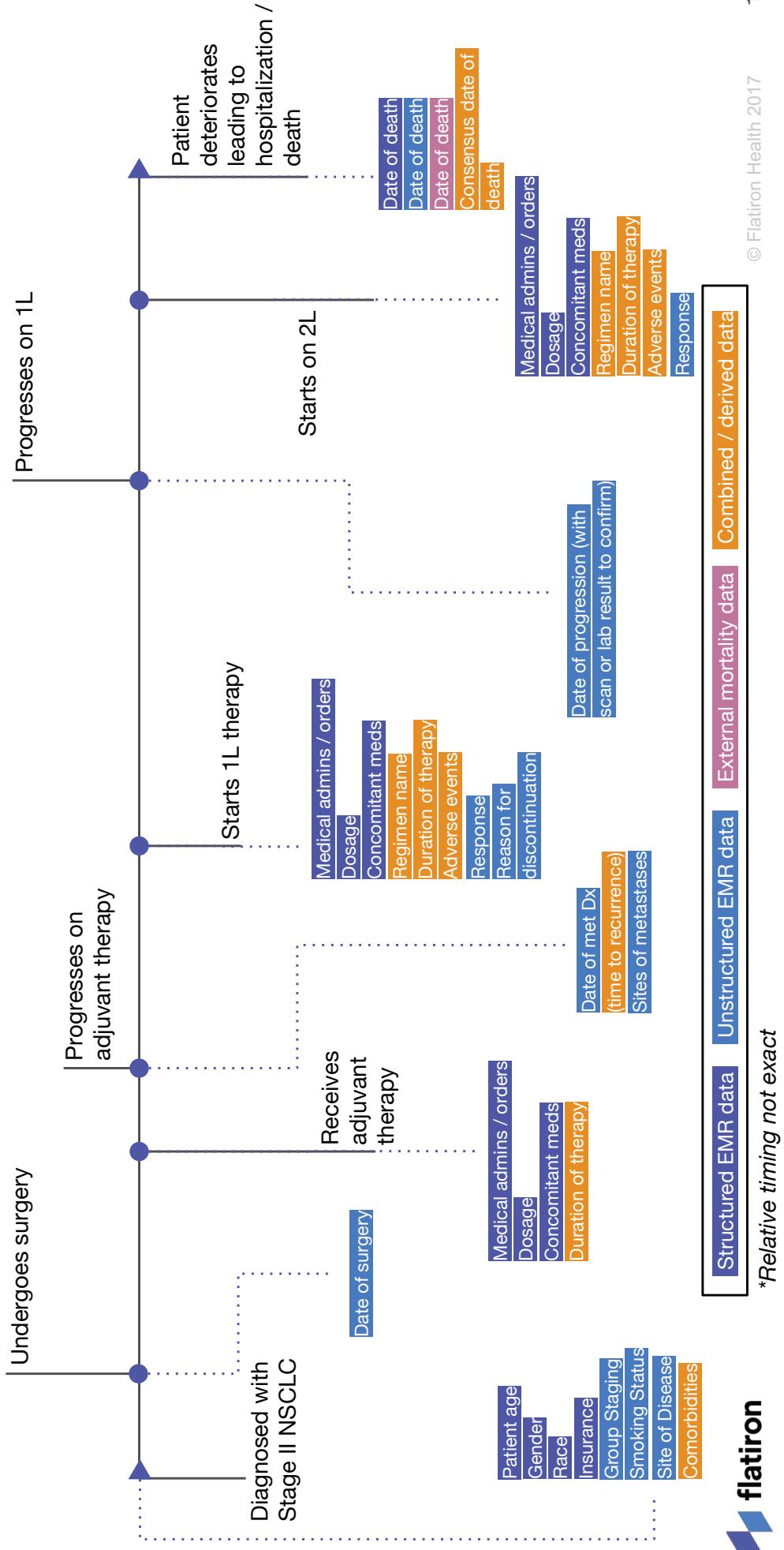
# Data source and curation



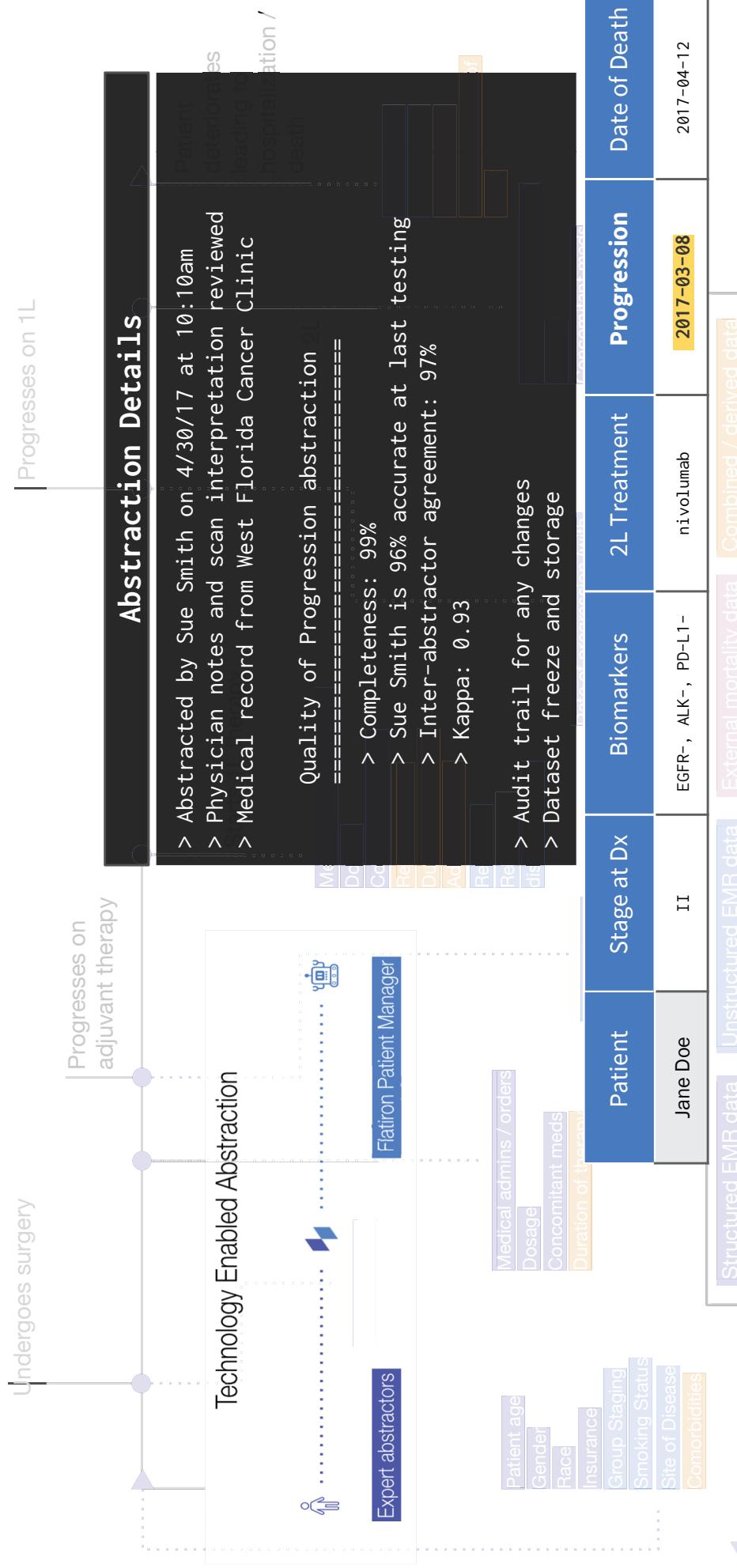
# Data source and curation



# A comprehensive view of the patient journey



# Technology Enabled Abstraction for curating variables



# Data quality across three dimensions



## Accuracy

- Validity of data elements
- Logical plausibility of results
- Data consistency for a given patient



## Completeness

- Extent of missingness in data
- Possible root cause and impact of missing data



## Traceability

- Transparency in data provenance and transformation
- Defined business logic for key variables

## Key metrics:

- Data completeness
- Missing data and impact on analysis?

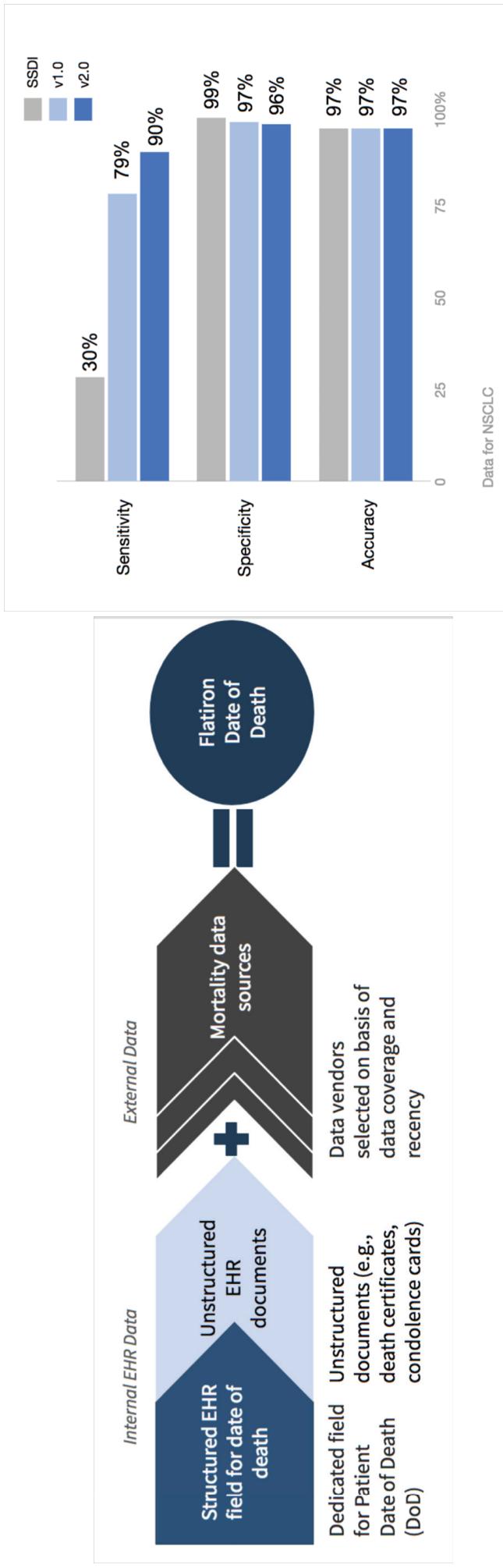
- Inter-abstracter agreement (proxy for accuracy) for derived variables from unstructured data

- Provenance
- Variable versioning

- *Harnessing the power of RWE: A checklist to ensure regulatory grade data quality, Clinical Pharmacology and Therapeutics, 2017; Rebecca Miksad, and Amy Abernathy*
- *Evaluation of the impact of Missing Deaths on Overall Survival Analyses Using A Real-World Mortality Endpoint* Gillis Carrigan, Samuel Whipple, Michael D. Taylor, Aracelis Z. Torres, Anala Gossai, Brandon Amieri, Melisa Tucker, Philip P. Hofmeister, Peter Lambert, Sandra D. Griffith, William B. Capra, PDS 2019

# Dataset linkage → Composite endpoint

## Evaluate underlying data quality (gold standard = NDI)



Melissa D. Curtis, Sandra D. Griffith, Melisa Tucker, Michael D. Taylor, William B. Capra, Gillis Carrigan, Ben Holzman, Aracelis Z. Torres, Paul You, Brandon Amieri, and Amy P. Abernethy, *Health Services Research 2018*

# rwP as a clinician-based endpoint

**Definition of rwP:** All distinct episodes in which the *treating clinician concludes that there has been overall growth or worsening of the disease of interest*

## NSCLC Patient Example

Advanced NSCLC diagnosis  
Started 1L carboplatin /  
pemetrexed

**Imaging showed  
progression; started  
docetaxel**

Assessment	Met Clinical Evidence of Progressive Non Small cell lung cancer Status post Alimta/carboplatin induction therapy followed by 12 cycles of maintenance of Alimta
Histological subtype	Adenocarcinoma
History of Tobacco abuse.	
Good performance status.	
Normocytic Anemia	
Disease Status: Progression of disease.	
Recommendation/Plan	
1.	Discussed pat/ct results and the fact that she <u>has evidence of disease progression</u> . Pros and cons of further treatment options were discussed.
2.	Incurable nature of disease was emphasized
3.	Given her good performance status and the fact that she wants to pursue with further therapy the game plan is to proceed with salvage therapy utilising single agent taxotere at a dose of 50mg/m <sup>2</sup> along with neutasta support.
4.	Also would continue her on Exjeva which we would give her every 6 weeks.

- **Clinically anchored with radiology and pathology reports serving**
- **as corroborative evidence**
- **Most practical and scalable**

# Validation: Patient-level correlation between rwPFS and OS

## Methods:

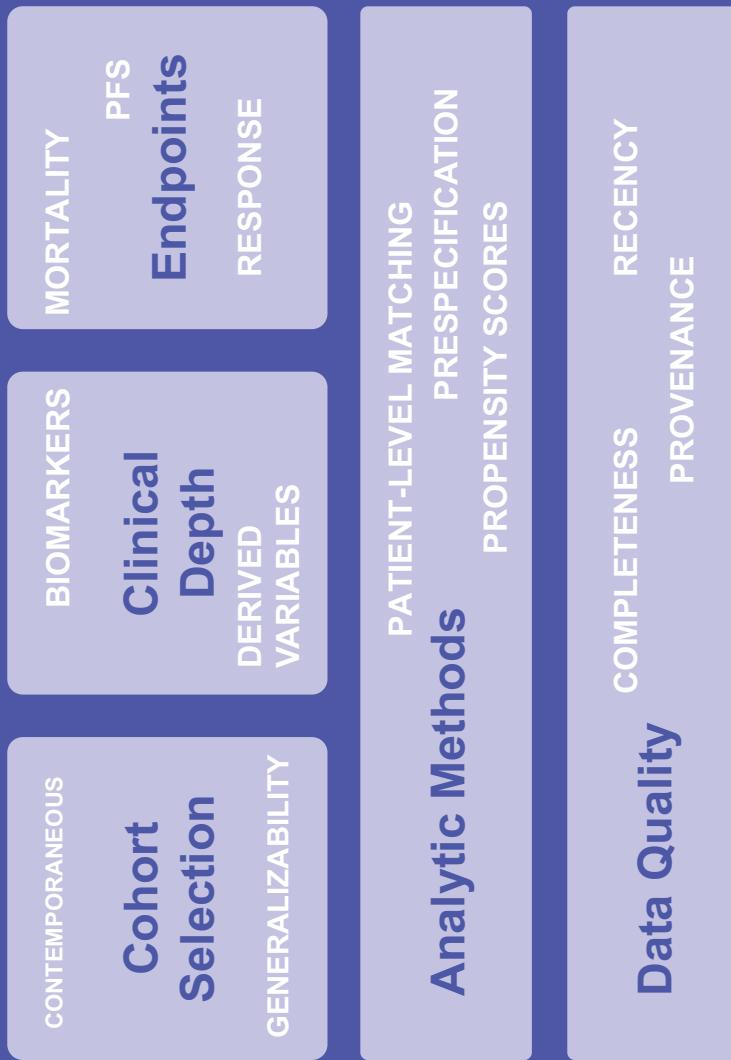
- Real-world time to progression or death was calculated and plotted against time to death for each patient
- Patients without a date of death were excluded from this analysis
- Correlation was calculated using Spearman's rank correlation coefficient

Correlation	N	p (95% CI)
rwPFS vs OS	20,020	0.76 (0.75, 0.77)
rwTTP vs OS	11,902	0.69 (0.68, 0.70)
rwTTNT vs OS	9,269	0.61 (0.60, 0.62)

# Real-World Control



# Foundation for rwCA Development



## Validation through Replication:

Can we replicate the outcomes observed in the control arms of recent clinical trials using Flatiron's real-world data?

*Bennette C et al. Use of a curated electronic health records database to create external control arms for cancer clinical trials. In submission.*



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# Constructing real-world control arms

**Randomized** trials supporting FDA approvals for anticancer drugs/biologics from 1/1/2016 - 4/30/2018  
**N=49**



Control arm was not placebo alone  
(or “observation”)  
**N=36**



Approval in tumor type  
represented by an existing data  
model from Flatiron Health  
EHR database with sufficient data  
**N=15**

*Approvals included initial, supplemental, accelerated  
& regular approval following accelerated approval*

# Constructing real-world control arms

**Randomized** trials supporting FDA approvals for anticancer drugs/biologics from 1/1/2016 - 4/30/2018  
**N=49**

- Identify real-world patients with treatment and molecular features consistent with trial's control arm
- Align real-world patient population with trial's inclusion and exclusion criteria



Control arm was not placebo alone (or “observation”)  
**N=36**



Approval in tumor type represented by an existing data model from Flatiron Health EHR database with sufficient data  
**N=15**

Approvals included initial, supplemental, accelerated & regular approval following accelerated approval

Tumor type	Name of trial	Front line setting?	Primary endpoint(s)
melanoma	CheckMate-067	yes	PFS, OS
kidney	METEOR	no	PFS
kidney	NCT01136733	no	PFS
NSCLC	OAK	no	OS
NSCLC	POPLAR	no	OS
breast	MONARCH-3	yes	PFS
breast	MONALEESA-2	yes	PFS
myeloma	POLLUX	no	PFS
NSCLC	KEYNOTE-024	yes	PFS
head and neck	CheckMate-141	no	OS
myeloma	CASTOR	no	PFS
NSCLC	AURA3	no	PFS
breast	PALOMA-2	yes	PFS
NSCLC	KEYNOTE-021	yes	ORR
urothelial	KEYNOTE-045	no	PFS, OS
NSCLC	ALEX	yes	PFS
kidney	CABOSUN	yes	PFS
breast	OlympiAD	no	PFS
kidney	CheckMate-214	yes	ORR, PFS, OS
breast	MONARCH-2	no	PFS
breast	PALOMA-3	no	PFS

## Summary of randomized clinical trials included in analyses



# Weighting real-world patients to published trials

- Derive “inverse odds” weights ( $w_i = \Pr(C_i=0 | x_i) / \Pr(C_i=1 | x_i)$ ) that represent odds patient was in the trial ( $C_i=1$ ) vs the real-world cohort ( $C_i=0$ ) given baseline characteristics ( $x_i$ )
- Approach is analogous to common method of calculating propensity score weights, except we
  - Use inverse odds rather than inverse probability so that we standardize to patients in the trial (and resulting treatment effect can be interpreted in much the same way it would from a randomized trial)
  - Use generalized method of moments rather than maximum likelihood to estimate logistic regression model because we have only summary data for trial

## Cohort Selection

Table 1. Flatiron cohort attrition

Number	Description
41144	Step 1a: Locally advanced or metastatic NSCLC who received platinum-based therapy in 1st or 2nd line
22036	Step 1b: Received platinum-based therapy in 1st or 2nd line
1136	Step 2a: Docetaxel after platinum-based therapy
377	Step 2b: Docetaxel received before trial enrollment ended
364	Step 3: Disease progression during or following prior platinum-based therapy
357	Step 4: No prior docetaxel, anti-CTLA-4, or PD-L1/PD-1 inhibitor
322	Step 5: Exclude patients with ECOG PS 2+
191	Step 6: Exclude patients with inadequate organ function (per protocol)

## Cohort Selection and data completeness report

### Completeness summary

Table 2. Completeness of key data elements in replicating I/E criteria

Data element	% Complete
Neutrophils or granulocytes	91.3%
Creatinine	99.5%
Platelets	98.9%
Hemoglobin	98.9%
Albumin	90.7%
Lymphocytes	96.2%
White blood cells	98.9%
ALT	96.7%
AST	96.7%
Calcium	95.1%
Bilirubin	91.8%
ECOG	36.6%

# OAK comparison

## Baseline patient characteristics

	RWD I/E aligned	Weighted	Trial Control	Trial Treatment
% <Med Age	35	50	51	50
% Male	56	61	61	61
% White	71	71	70	71
% Squamous	12	11	26	26
% EGFR+	4	6	10	10
% KRAS+	12	12	8	6
% 1 Prior Line	52	54	75	75
% Smoking History	92	80	83	80

Atezolizumab versus docetaxel in patients with previously treated non-small-cell lung cancer (OAK): a phase 3, open-label, multicentre randomised controlled trial  
  
 Achim Rittmeyer, Fabrice Barlesi, Daniel Pao, Fernando Cardillo, Joachim von Pawel, Shiloh M Gadjev, Toyoki Hida,  
 Daniel M Kowalewski, Manuel Ceballos, Diego J Contreras, Joseph Leigh, Jonathan Roskoff, Carlos Bauries, Farrooz Kohanlouvan,  
 Donald Aien Frontiere, Filippo De Marinis, Hsueh Tzeng, Jong-Soo Lee, Marcus Ballinger, Marcus Kowanetz, Pei He, Daniel S Chen, Alan Sandler,  
 David R Gordon, for the OAK Study Group\*

## Baseline Comparison



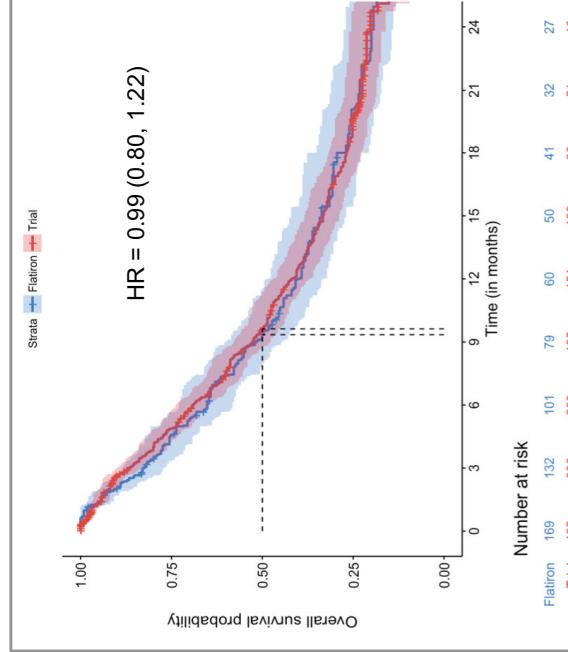
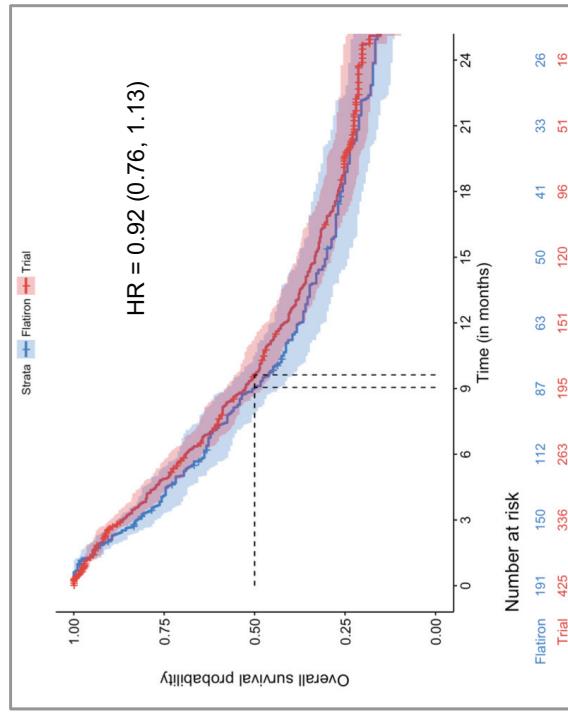
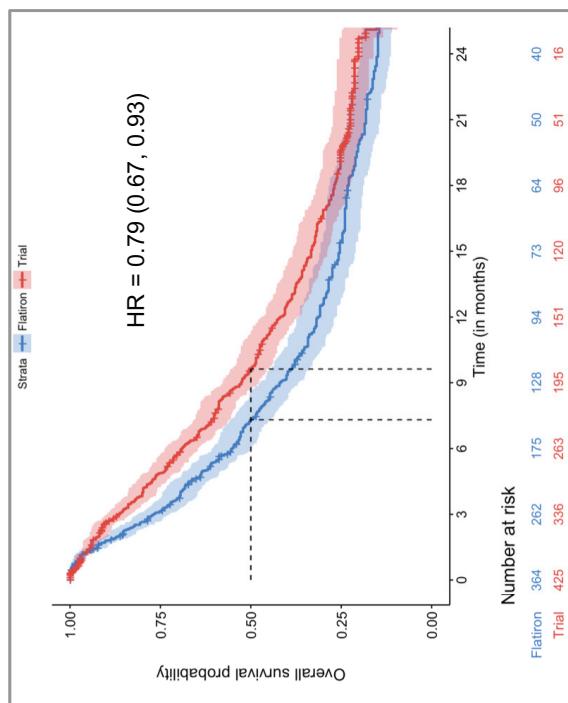
# OAK Comparison

## Overall survival (Primary endpoint)

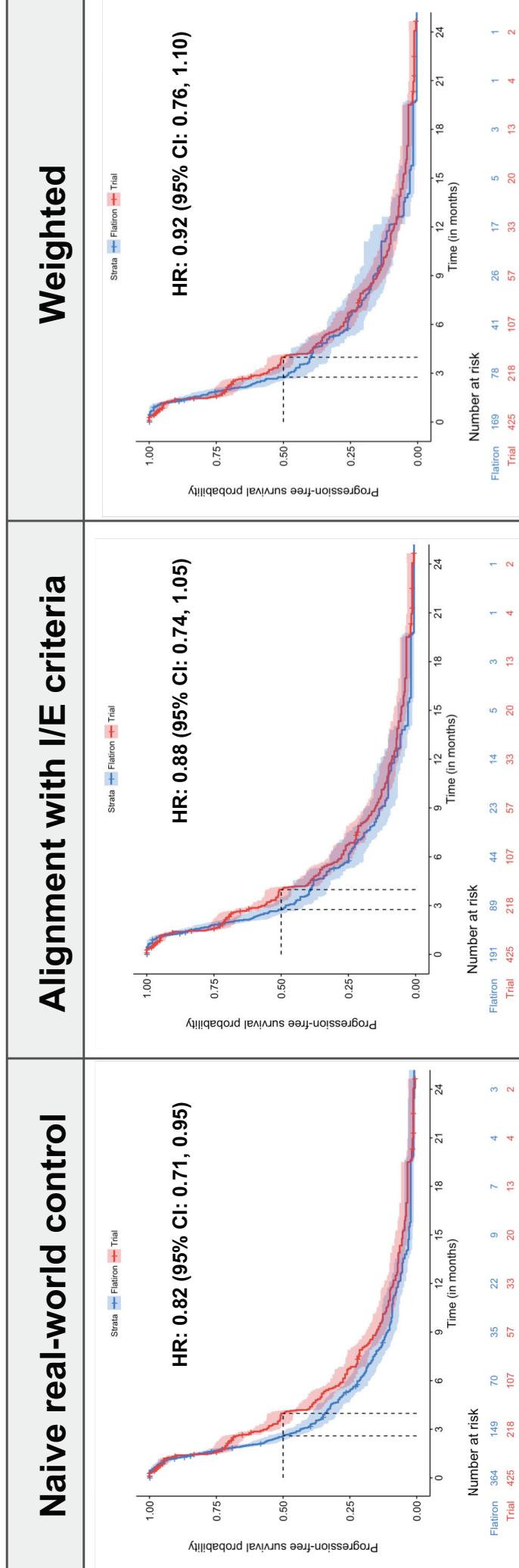
### Naive real-world control

### Alignment with I/E criteria

### Weighted



# Progression-free survival (secondary endpoint)

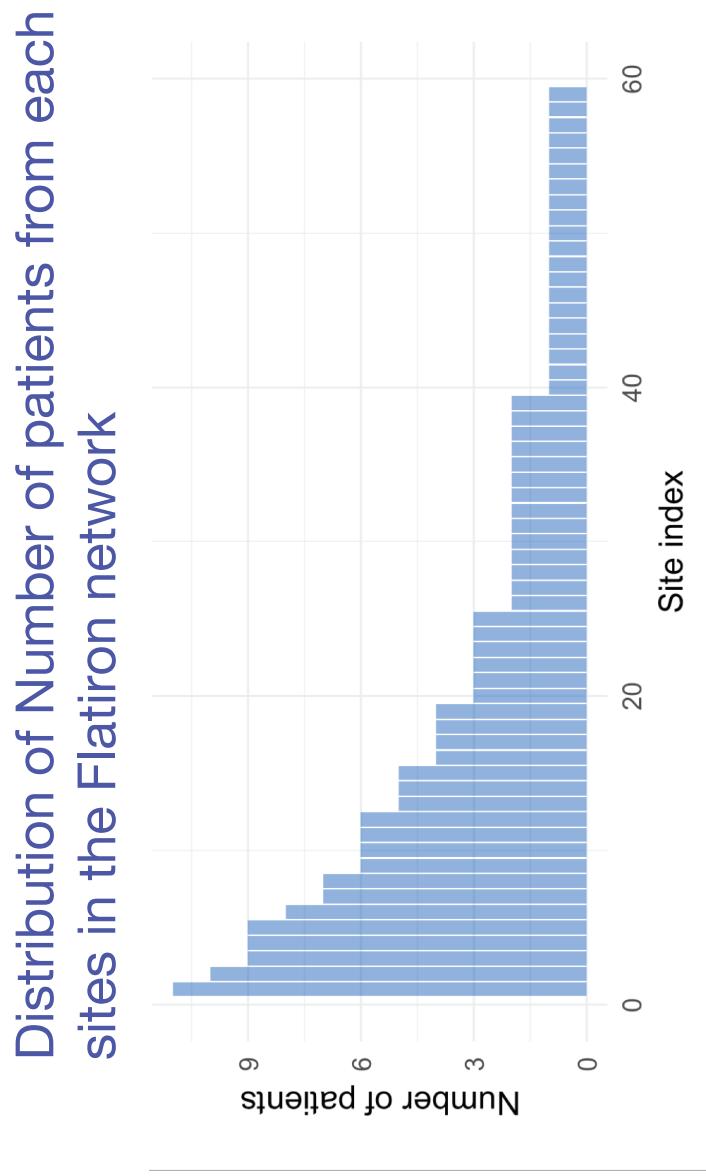


# Planned sensitivity analyses

Comparing patients in original trial's control arm versus :

Trial	Main analyses (weighted) HR (95% CI)	Excluding patients with missing ECOG performance status HR (95% CI)	Excluding patients with missing laboratory results used to define organ function HR (95% CI)	Excluding patients treated before trial enrollment period started HR (95% CI)
OAK (OS)	0.99 (0.80, 1.22)	1.21 (0.86, 1.70)	1.04 (0.82, 1.33)	1.05 (0.77, 1.44)
OAK (PFS)	0.92 (0.76, 1.10)	0.99 (0.75, 1.31)	0.94 (0.77, 1.16)	0.98 (0.75, 1.28)

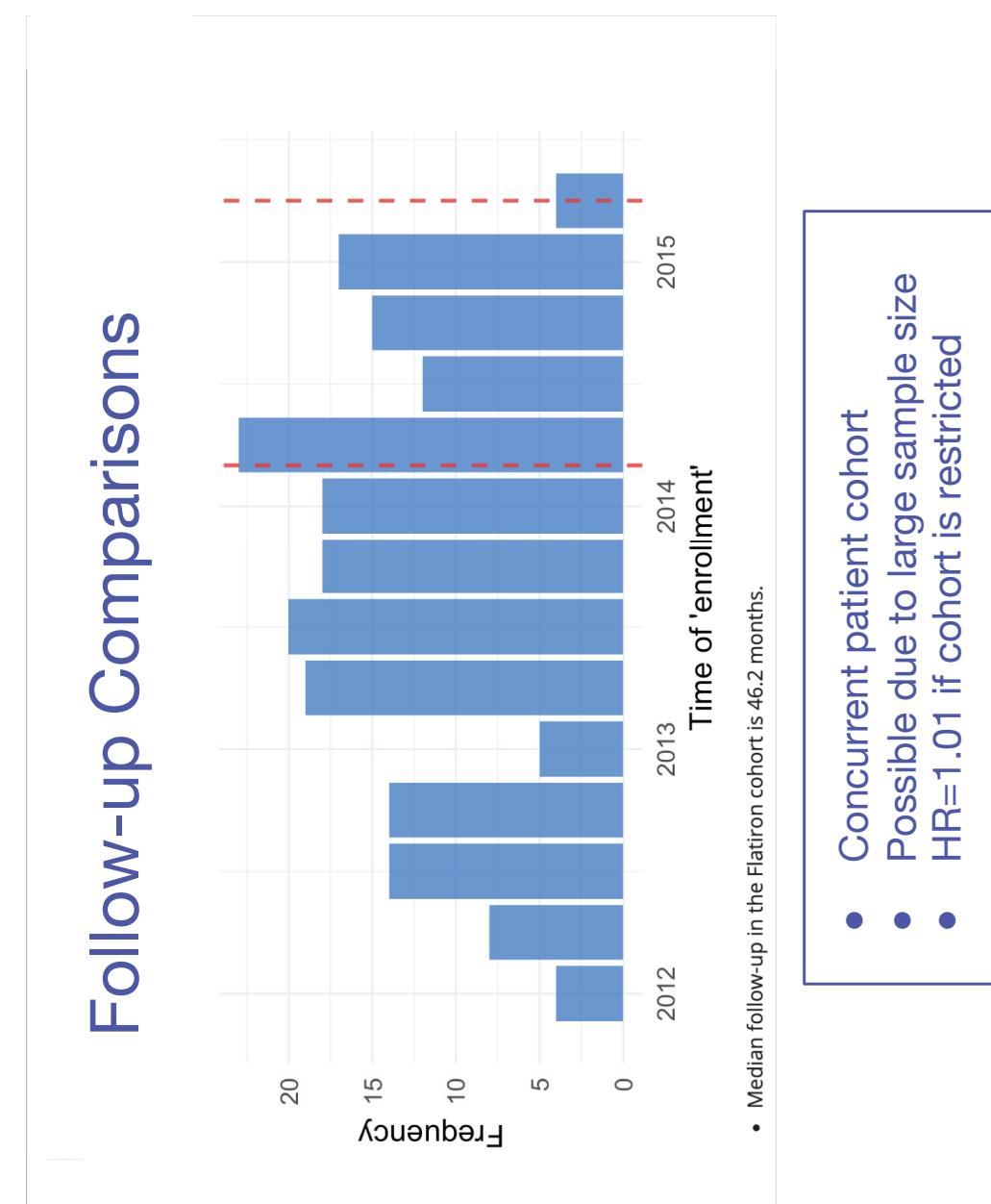
## Diagnostics **Generalizability** (Distribution of sites)



- Total 59 sites
- Both academic and community are well represented

## Diagnostics

### Trial recruitment window (51.5% of Flatiron cohort)

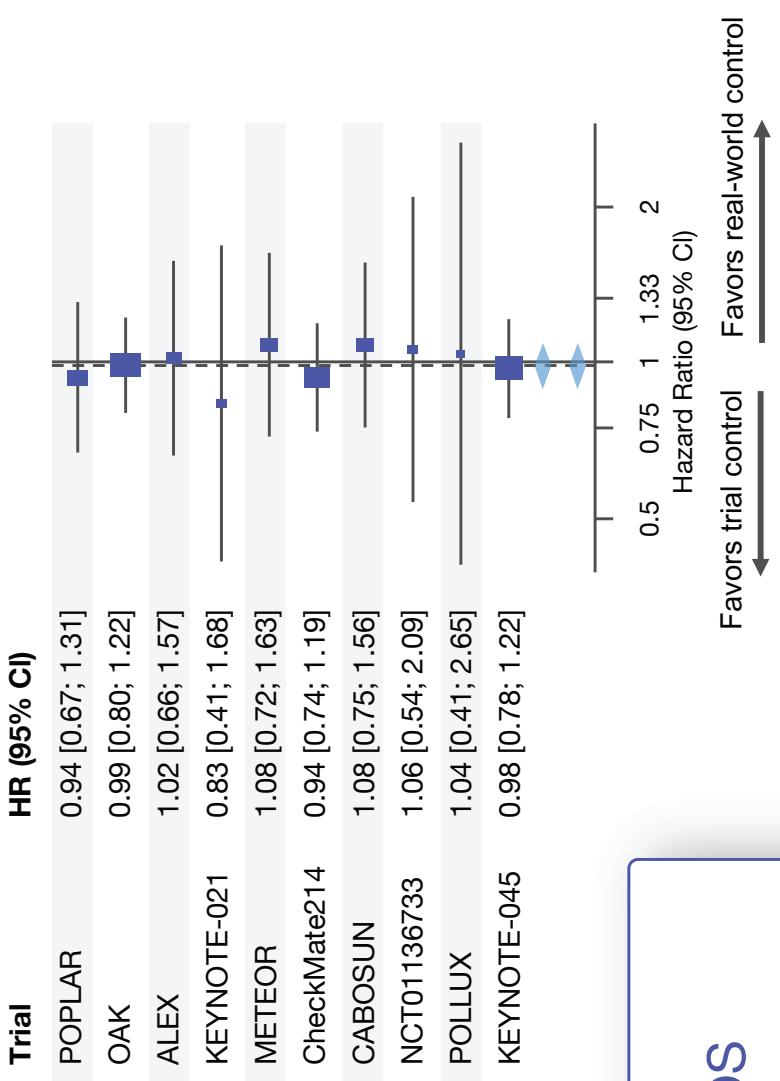


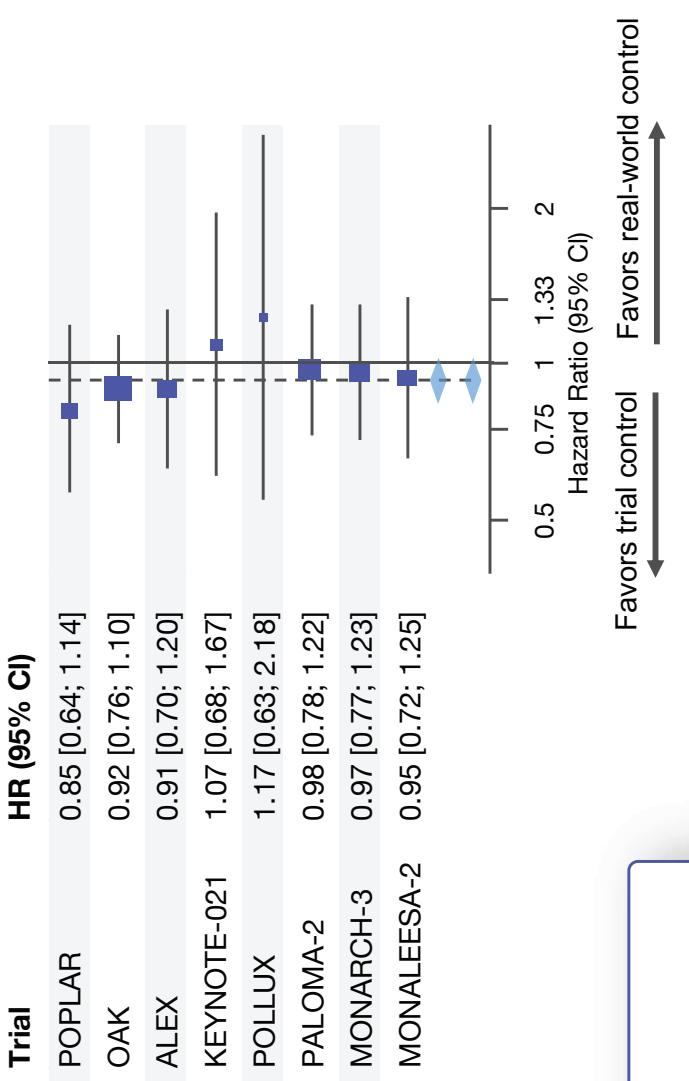
- Median follow-up in the Flatiron cohort is 46.2 months.

- Concurrent patient cohort
- Possible due to large sample size
- $HR=1.01$  if cohort is restricted

**OS Combined  
HR (95% CI):**  
**0.98 (0.89 - 1.09)**

- Overall consistent finding for OS

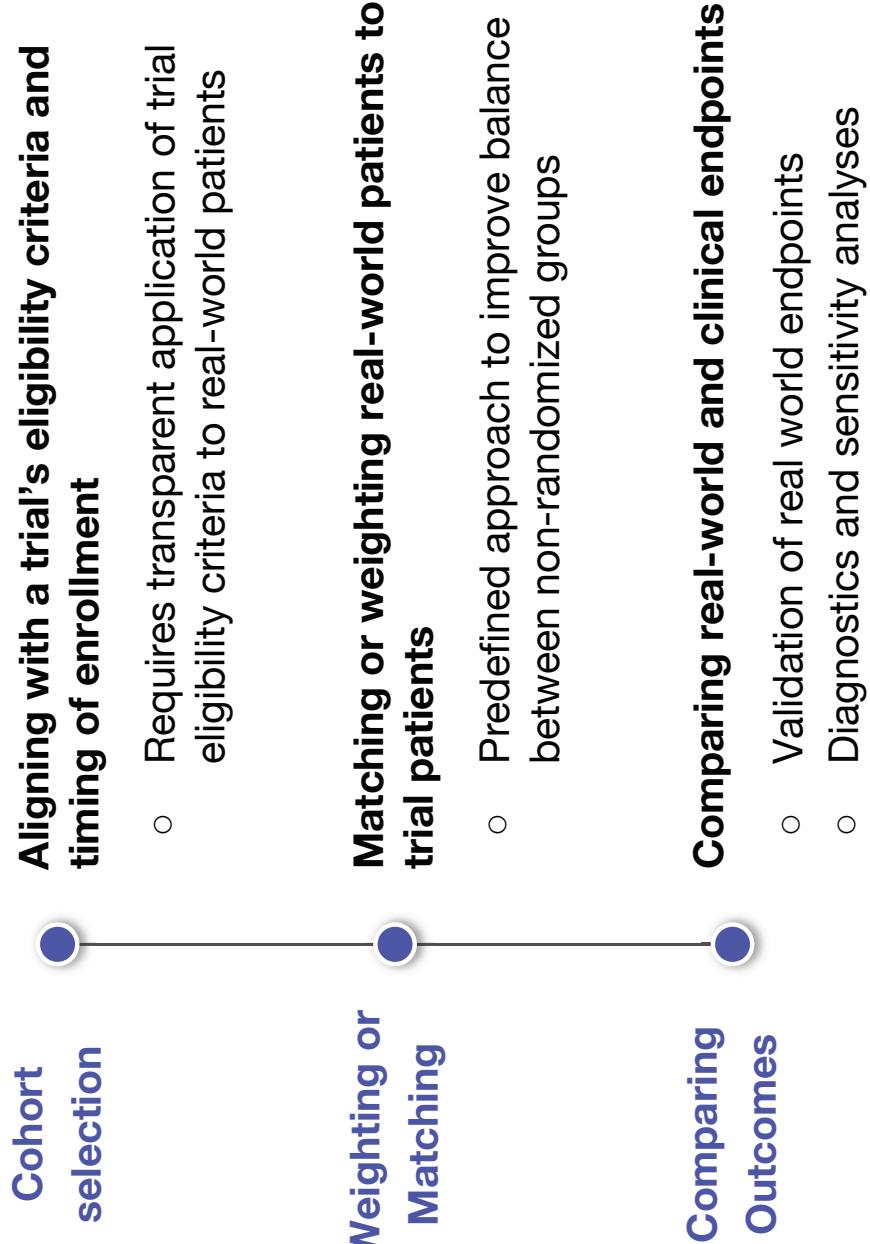




**PFS Combined HR (95% CI):**  
0.94 (0.86 - 1.04)

- Similarly consistent finding for PFS

# Three Key Steps to construct rwCA from curated EHR databases



## Cohort selection

- Ideally RW patients treated during the same timeframe of a trial enrollment
- Broader windows may be chosen in absence of no substantial change in SoC and rarer patient population

## Weighting or Matching Comparing Outcomes

- Documenting which eligibility criteria were implemented
  - Identify patients who are similar in prognosis, noting features driving prognosis are different even within a disease setting
  - Many of I/E criteria are more difficult to implement (some are feasible with addition abstraction and/or proxies)
  - Clinical importance of infeasible eligibility criteria depends on the context

## Cohort selection

## Selection of covariates

- Systematic literature review to identify key prognostic factors that are measured in both datasets

## Weighting or Matching

## Comparing Outcomes

## Handling potential missingness in RWD

- e.g. for non-routinely performed lab tests it may be reasonable to assume: absence of a test as absence of the underlying condition (**e.g. viral hepatitis tests**)

- Attrition diagram and sensitivity analyses

## Are the comparison groups balanced on known baseline characteristics?

- Planned sensitivity analysis showing consistency

## Cohort selection

- Systematic differences in how the index date was defined may result in biased results
- Ensure that outcome assessments are occurring at reasonable intervals and can be captured reliably
- Evaluate the timing of follow up assessments & censoring patterns to compare to the clinical trial endpoint

## Weighting or Matching

## Comparing Outcomes

## “Threshold crossing” framework,

- Anticipated benefit is robust and efficacy threshold is specified *a priori*

- Oncology drug development/regulatory paradigm continues to change
  - Rapidly changing standard of care
  - Blurring of retrospective & prospective research (recency of RWD)
- EHR data have great potential to provide research/regulatory-grade evidence
  - Important to demonstrate quality, validity and analytical considerations of RWE

## Conclusion

# Thank you