

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

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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”

Phase 1a/1b/single arm Phase 2

Potential for submission

Phase 3

Confirmatory Studies

Learning

Confirming

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

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

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

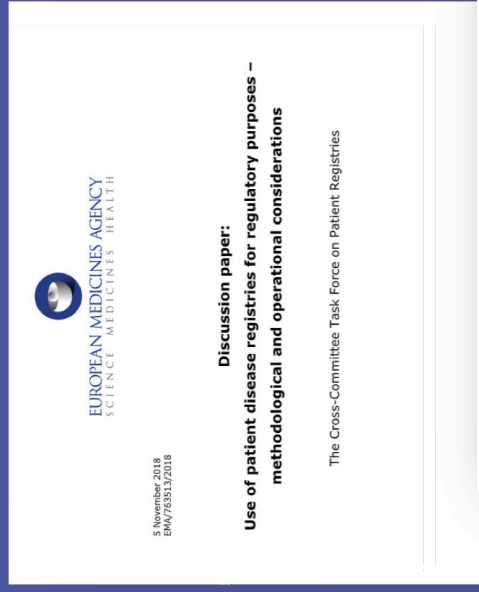
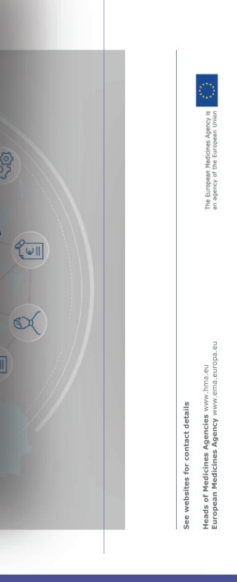


*Creating a sustainable, quality assured, flexible framework delivering rapid access to and analysis of representative, longitudinal RWD throughout a product's lifecycle*

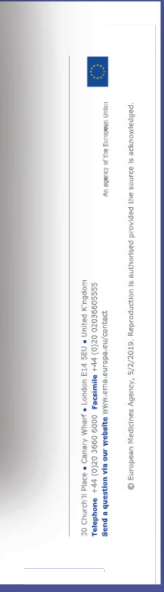


*e.g. "development of a framework to articulate for what questions and contexts RWE may be acceptable across the product life cycle"*

*"strongly support exploration of novel analytics approaches"*



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

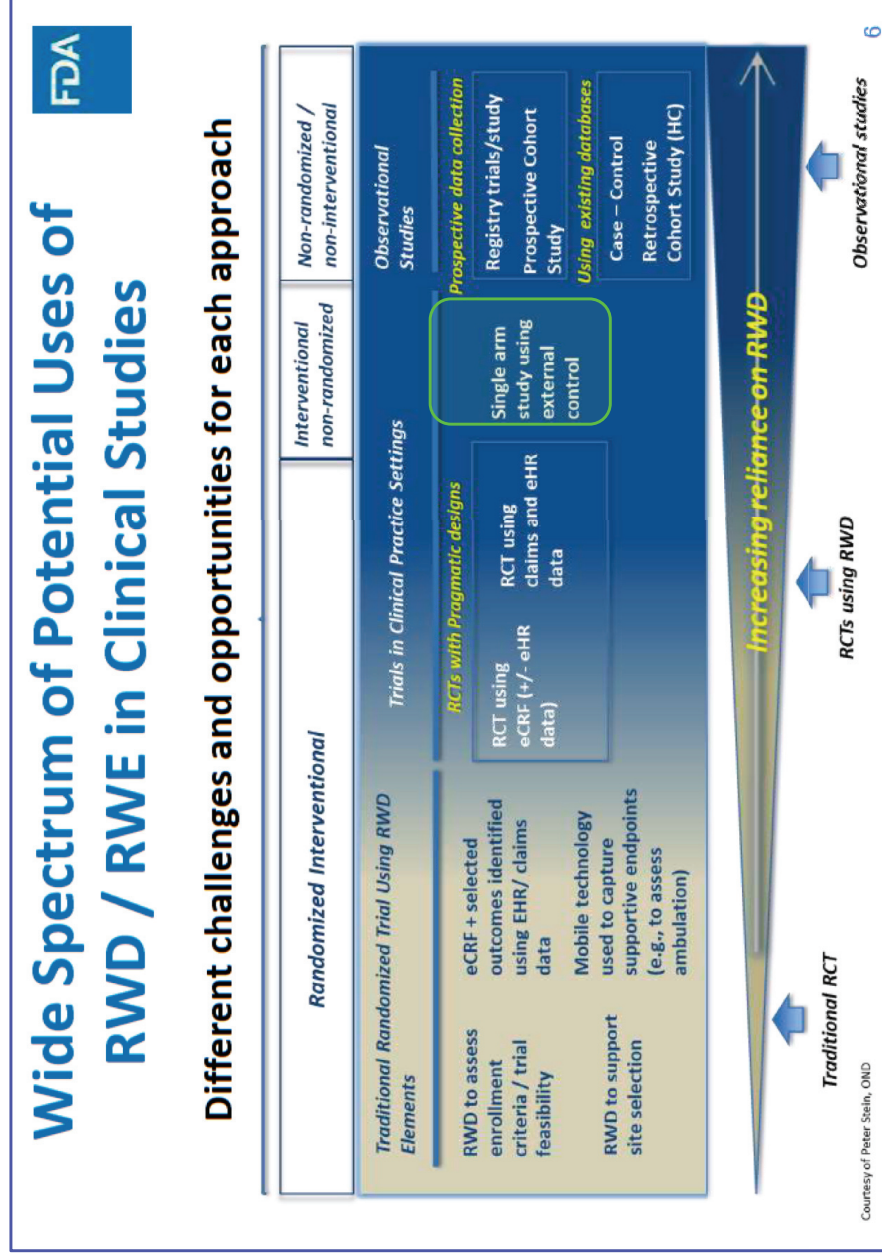


FDA has legislative mandate to explore IF and WHEN RWE may support new indications (approved drugs)/post marketing requirements



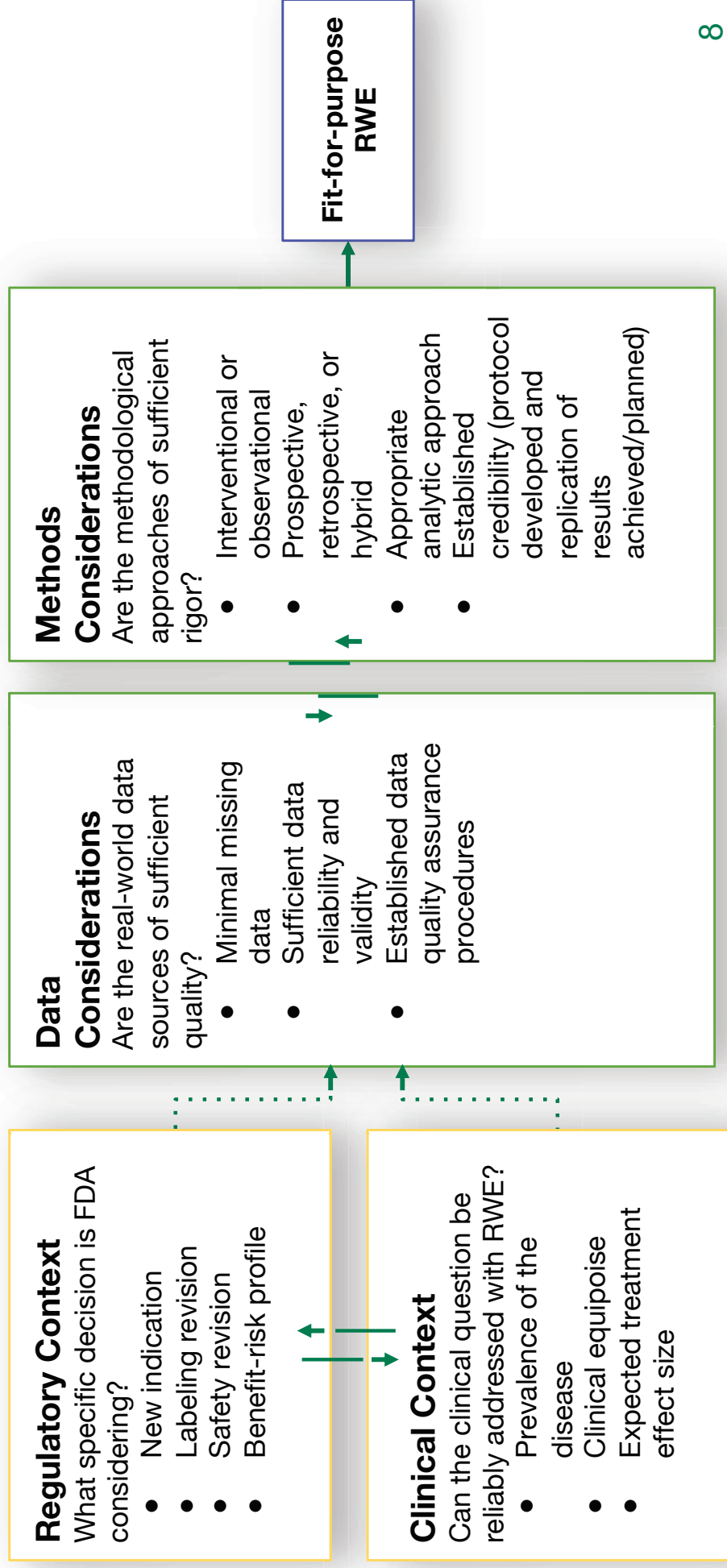
Framework

<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





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

**Regulatory Context**  
What specific decision is FDA considering?

- New indication

**Example of Real-world Control:**

- Rare diseases
- New indication
- Expected large treatment benefit

disease

- Clinical equipoise
- Expected treatment effect size

**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)

**Fit-for-purpose RWE**

# Flatiron's real-world evidence generation platform



# Rapid adoption of EHRs in oncology

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

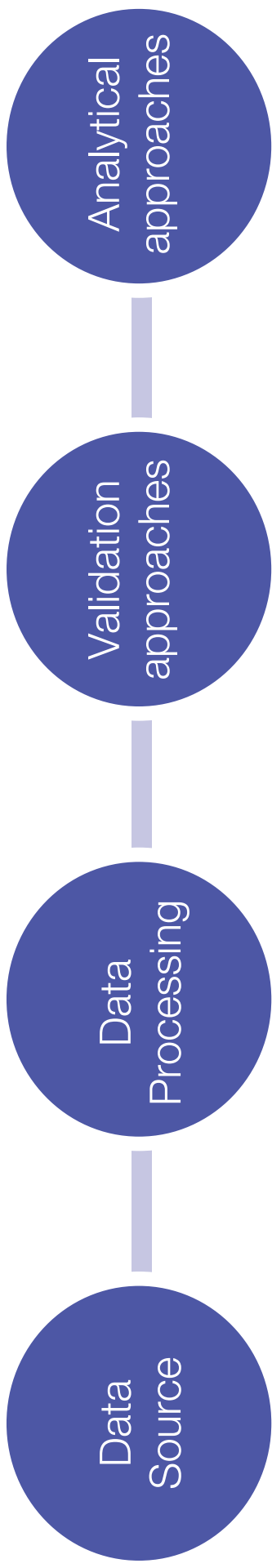
The Senate joined the House on Friday evening, February 13, 2009, in passing the American Recovery and Reinvestment Act, which includes provisions relating to Health Information Technology. Title XIII of Division A and Title IV of Division B together are known as the "Health Information Technology for Economic and Clinical Health Act" or the "HITECH Act." We will be highlighting attributes of the HITECH Act



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

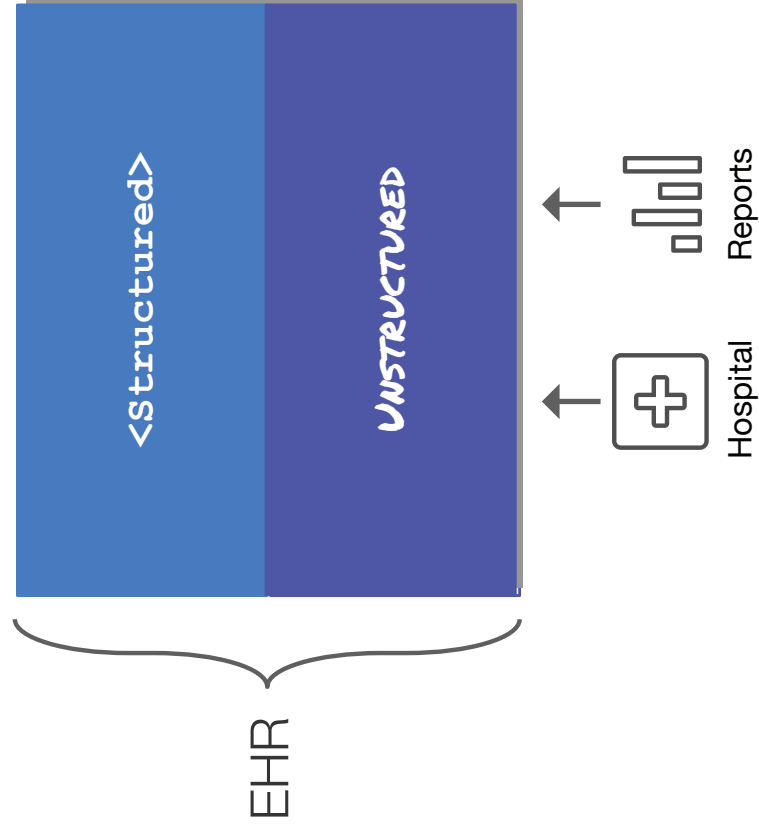
*Recent adoption in US parallels UK ambition to shift to a "paperless NHS"*

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

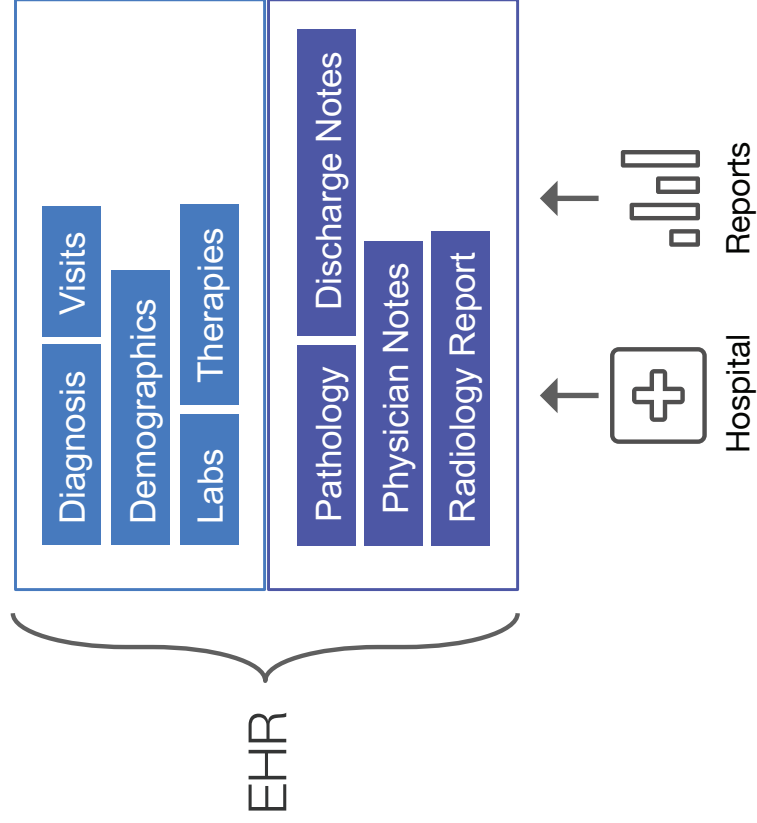


“RWE is derived from RWD through the application of research methods”

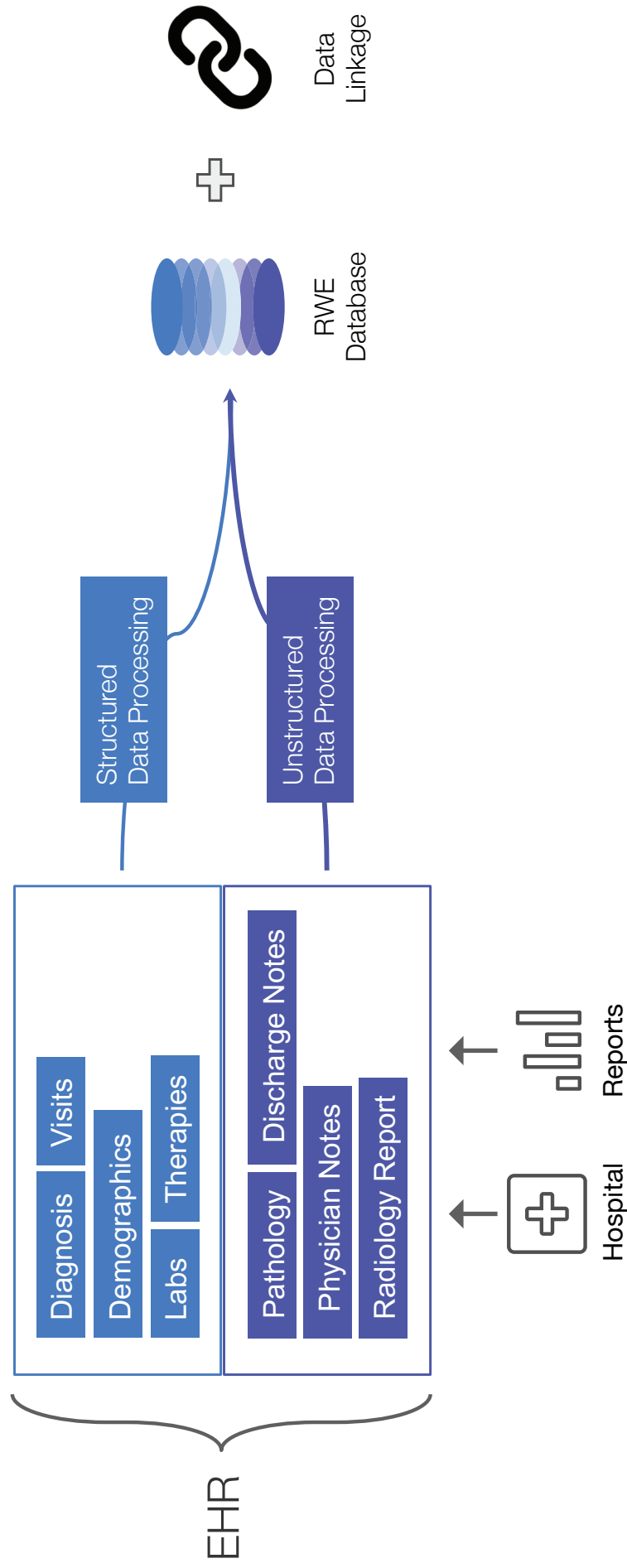
# Data source and curation



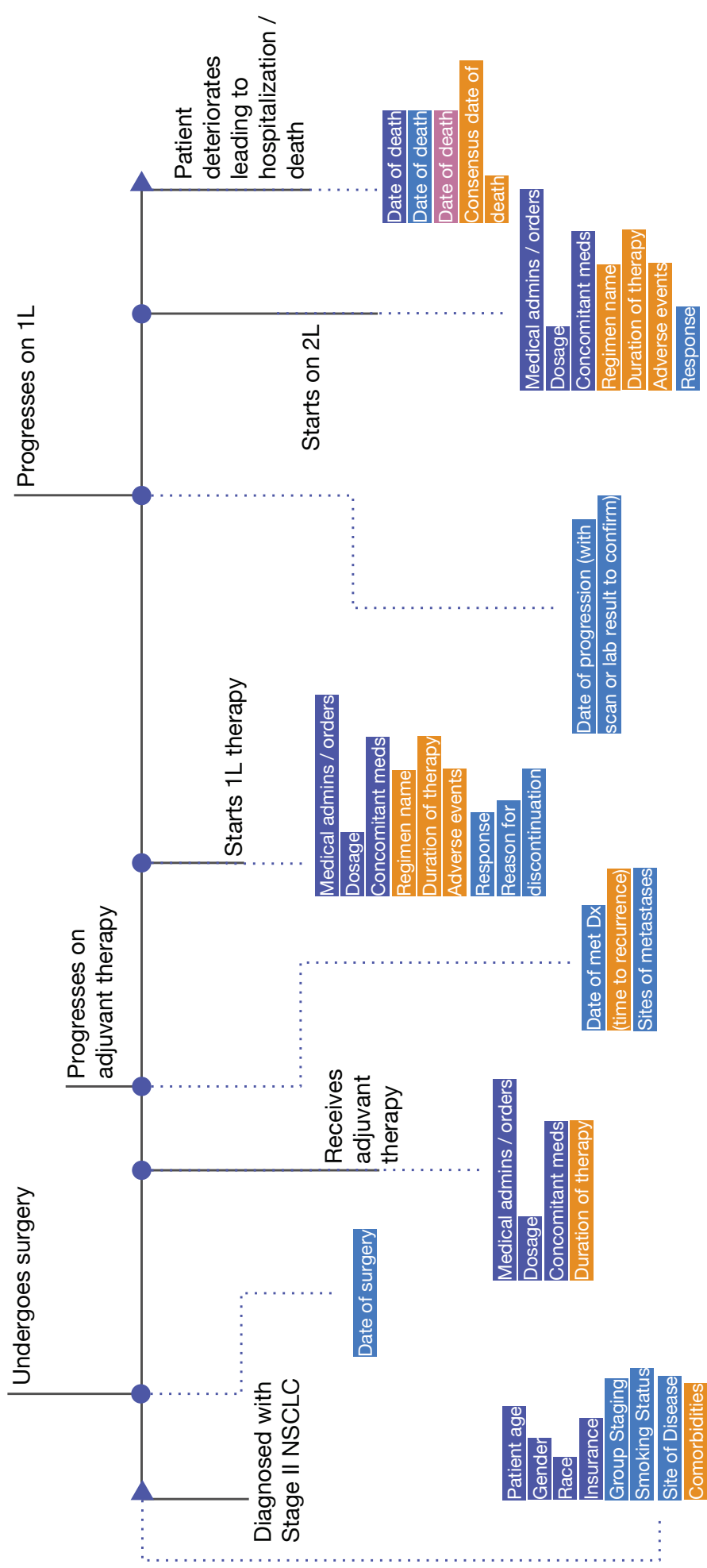
# Data source and curation



# Data source and curation



# A comprehensive view of the patient journey



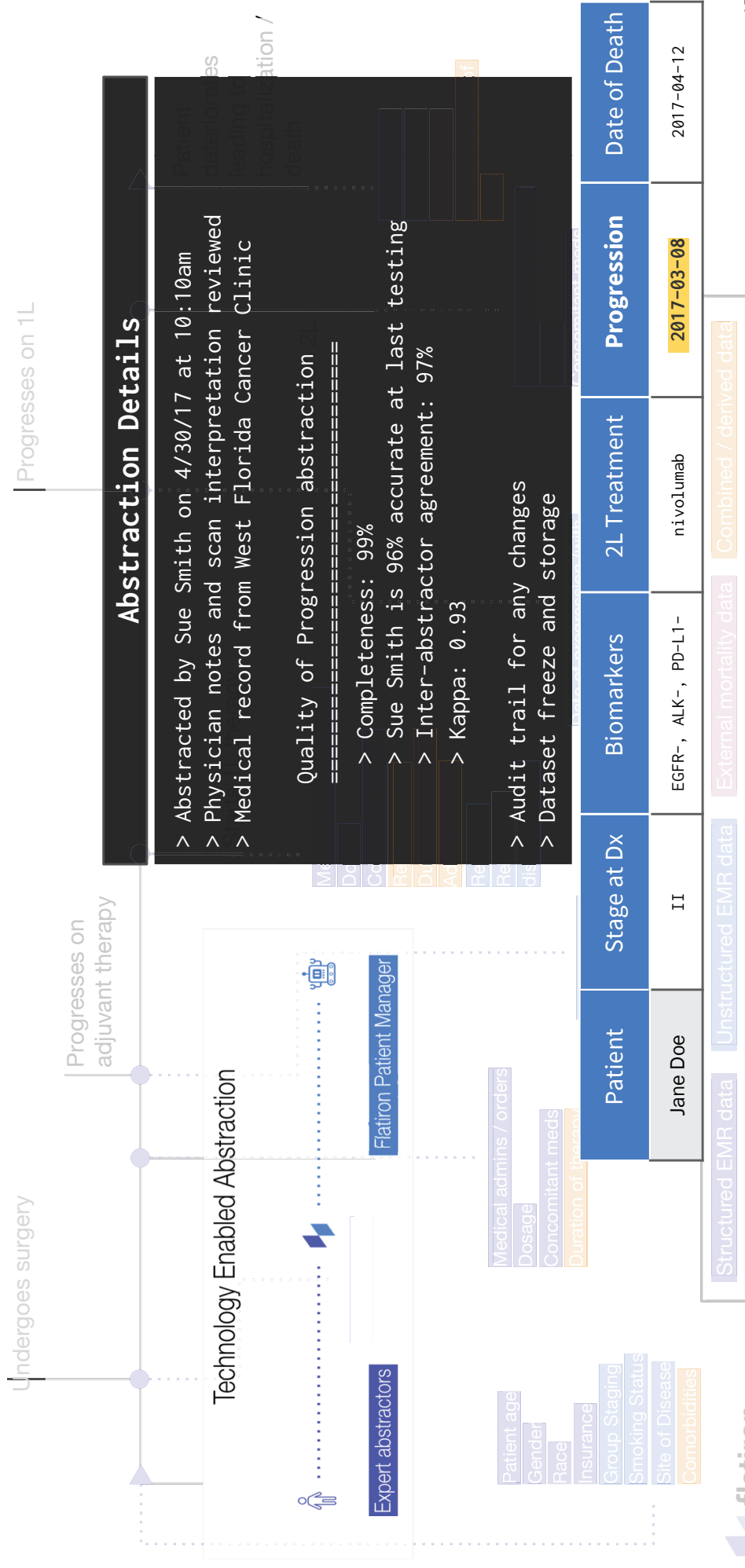
Structured EMR data | Unstructured EMR data | External mortality data | Combined / derived data

\*Relative timing not exact





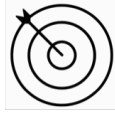
# Technology Enabled Abstraction for curating variables



\*Relative timing not exact



# 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:

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

- Data completeness
- Missing data and impact on analysis?

- Provenance
- Variable versioning

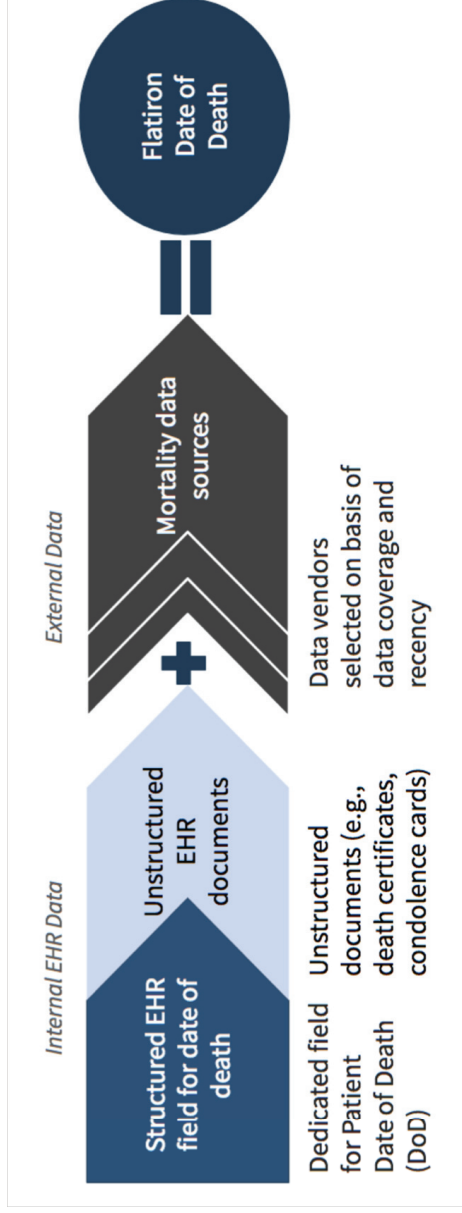
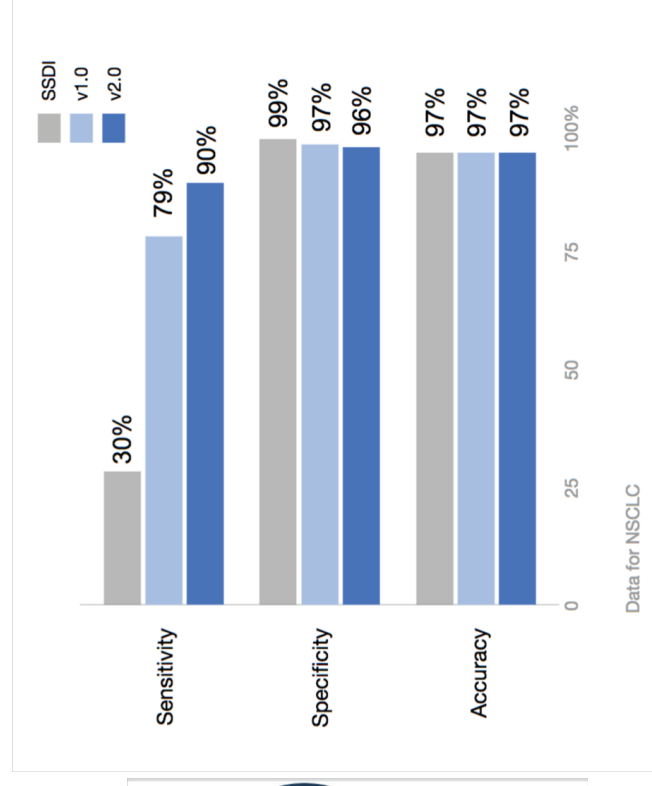
- **Harnessing the power of RWE: A checklist to ensure regulatory grade data quality**, *Clinical Pharmacology and Therapeutics*, 2017; Rebecca Miksad, and Amy Abernathy
- **Evaluation of the impact of Missing Deaths on Overall Survival Analyses Using A Real-World Mortality Endpoint** Gillis Carrigan,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



**flatiron**

# 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, Benjamin 2017 Holzman, Aracelis Z. Torres, Paul You, Brandon Armieri, 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



2013

**Imaging showed  
progression; started  
docetaxel**



2014

1

**Assessment**  
Pet Ct 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 pet/ct results and the fact that she has evidence of disease progression. 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 taxolere at a dose of 60mg/m2 along with neulasta support.  
4. Also would continue her on Exjiva 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:

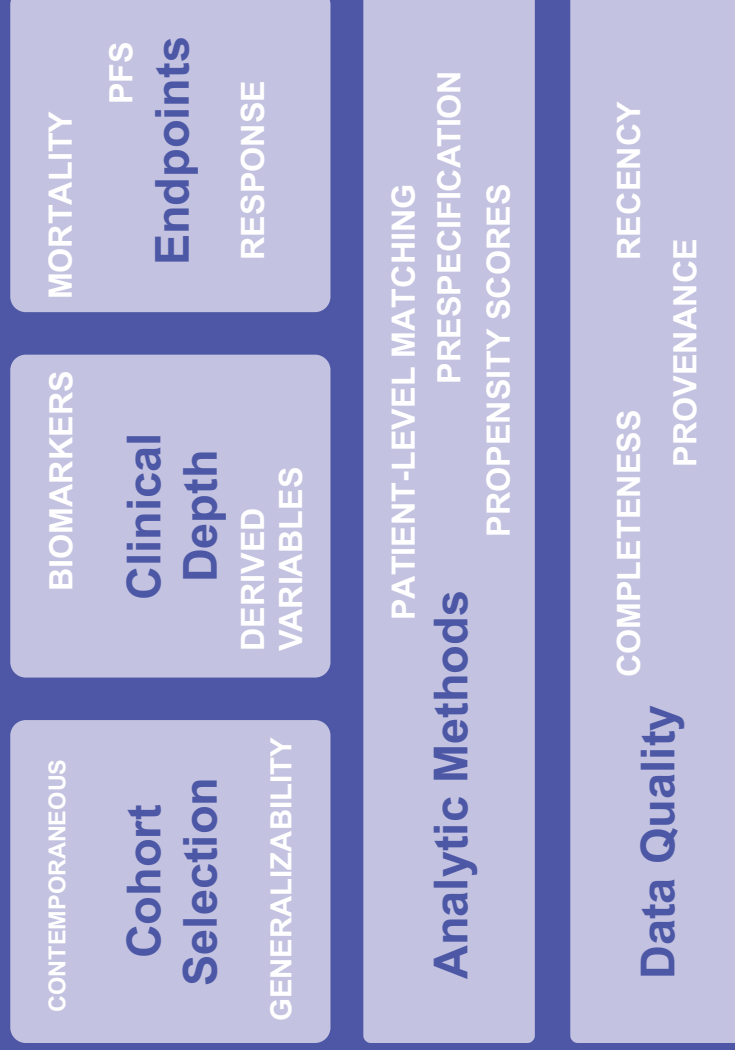
- Real-world time to progression or death was calculated and plotted against time to death for each patient
- Patients without a date of death were excluded from this analysis
- Correlation was calculated using Spearman's rank correlation coefficient

| Correlation  | N      | $\rho$ (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



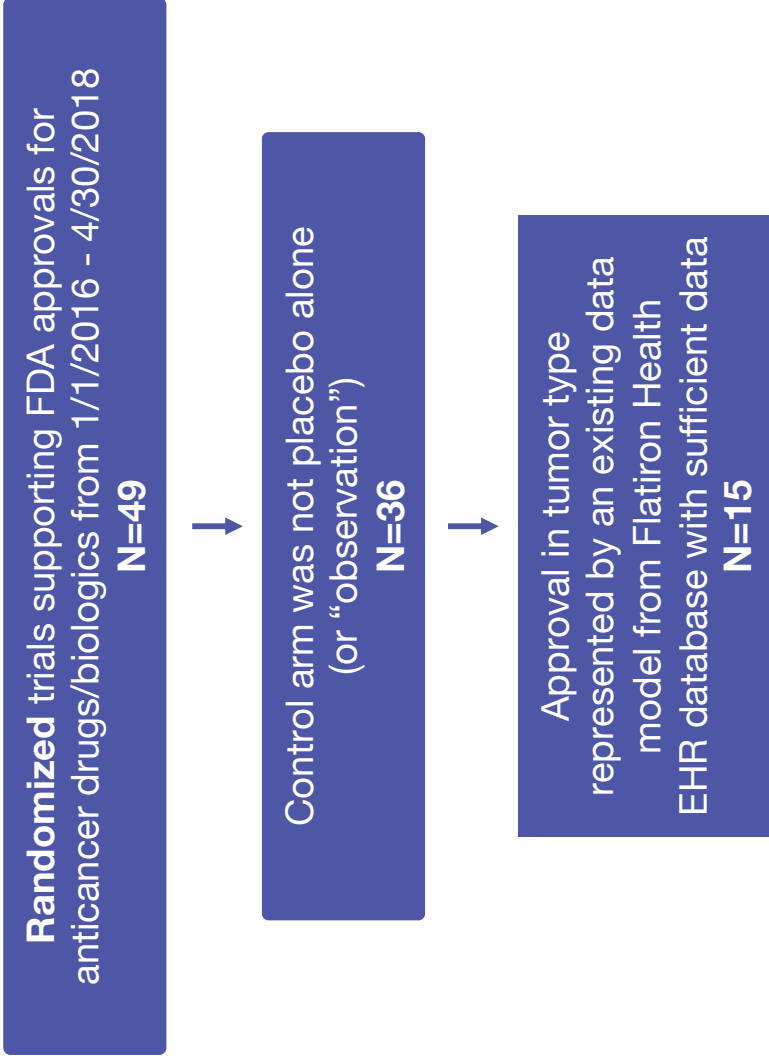
# Validation through Replication:

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.*

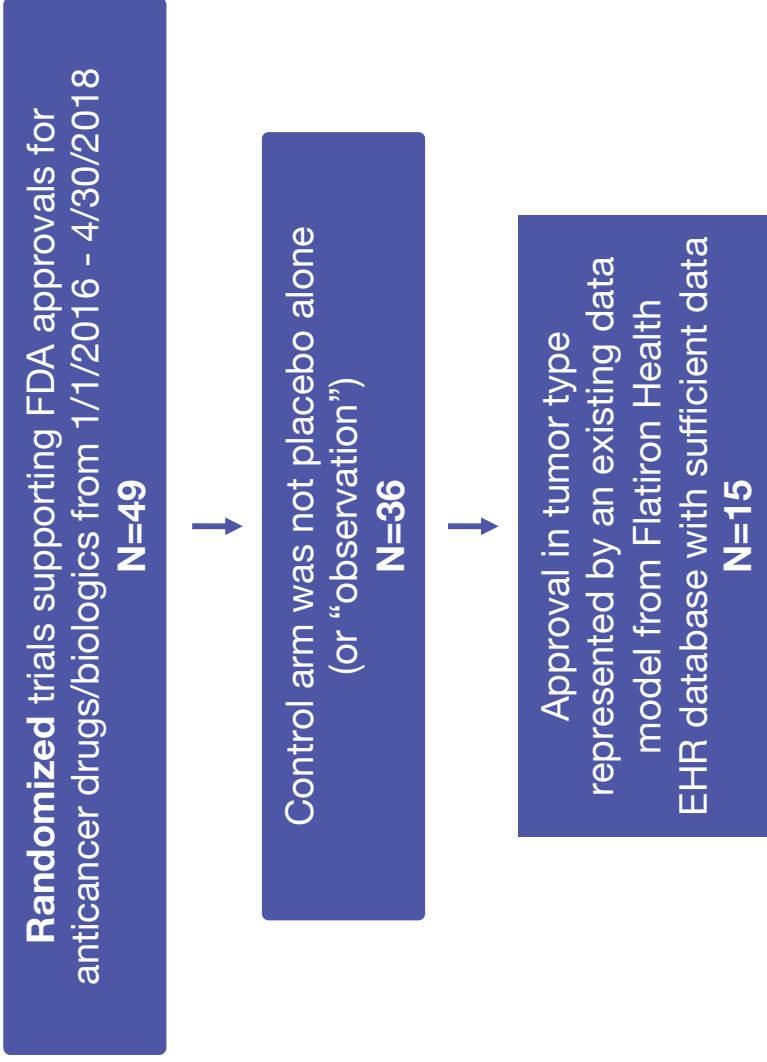


# Constructing real-world control arms



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

# Constructing real-world control arms



**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

If imbalances wrt key baseline characteristics, **weight** eligible real-world patients to match published baseline characteristics

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

# Summary of randomized clinical trials included in analyses

| 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                 |

# Weighting real-world patients to published trials

- Derive “inverse odds” weights ( $w_i = \Pr(C_i=0 \mid x_i) / \Pr(C_i=1 \mid x_i)$ ) 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

**Atezolizumab versus docetaxel in patients with previously treated non-small-cell lung cancer (OAK): a phase 3, open-label, multicentre randomised controlled trial**

Achim Bittner, Fabrice Barlesi, Daniel Waterkamp, Konrad Peik, Fortunato Giridillo, Joachim von Pawel, Sheikh M Gadgil, Toyooka Hida, Daniela M Kowalski, Manuel Cabo Rob, Diego Cortinovis, Joseph Leach, Jonathan Prollhoff, Carlos Barrios, Faouzi Kabbaj, Oweida Aïen Fontana, Filipo De Marinis, Hansel Triana, Jong-Seok Lee, Marcus Kowenz, Per He, Daniel S Chen, Alan Sandler, David R Comand, for the OAK Study Group\*



# Cohort Selection and data completeness report

## 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)                               |

## 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%      |

**Atezolizumab versus docetaxel in patients with previously treated non-small-cell lung cancer (OAK): a phase 3, open-label, multicentre randomised controlled trial**

Archie Rothenberg, Fabrice Barthelemy, Daniel Waterkamp, Kenneth Park, Fortunato Ciardiello, Joachim von Pawel, Shihsh M Godge, Toyooka Hiaki, Daniela M Kowalski, Manuel Cabo Robi, Diego Cortinovis, Joseph Leach, Jonathan Prollhoff, Carlos Barrios, Faizoo Kabbani, Oweida Amin, Francisco Filipejo De Alarim, Hansel Trana, Jong-Seok Lee, Marcus Kowenetz, Pei He, Daniel S Chen, Alan Sandler, David R Comandau, for the OAK Study Group\*



# Baseline Comparison

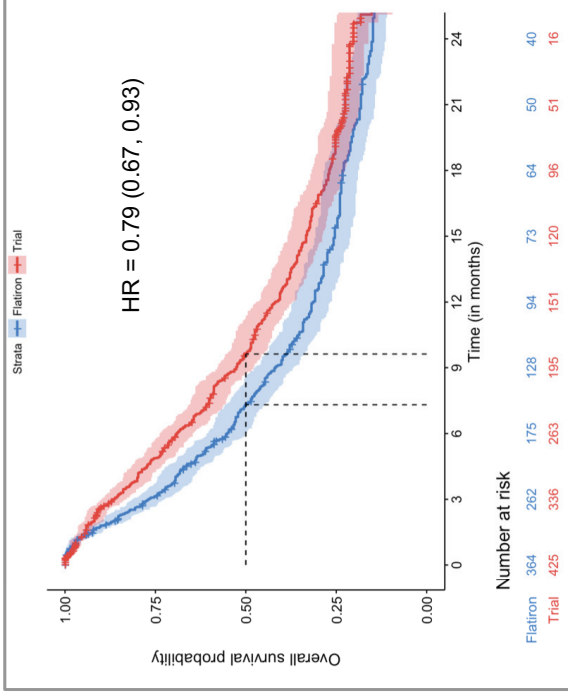
## Baseline patient characteristics

|                   | 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              |

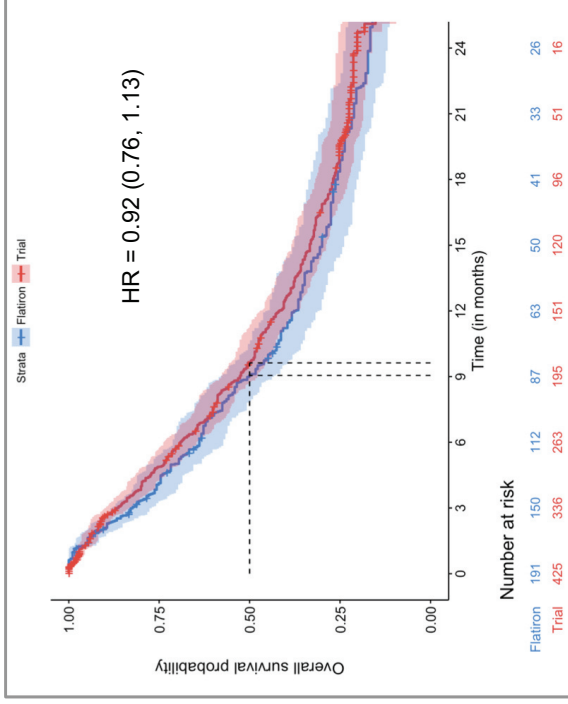
**OAK**  
comparison

# Overall survival (Primary endpoint)

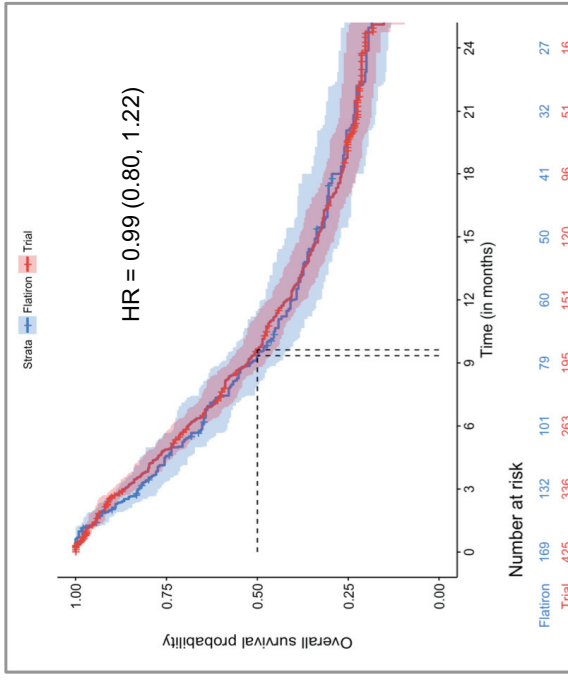
## Naive real-world control



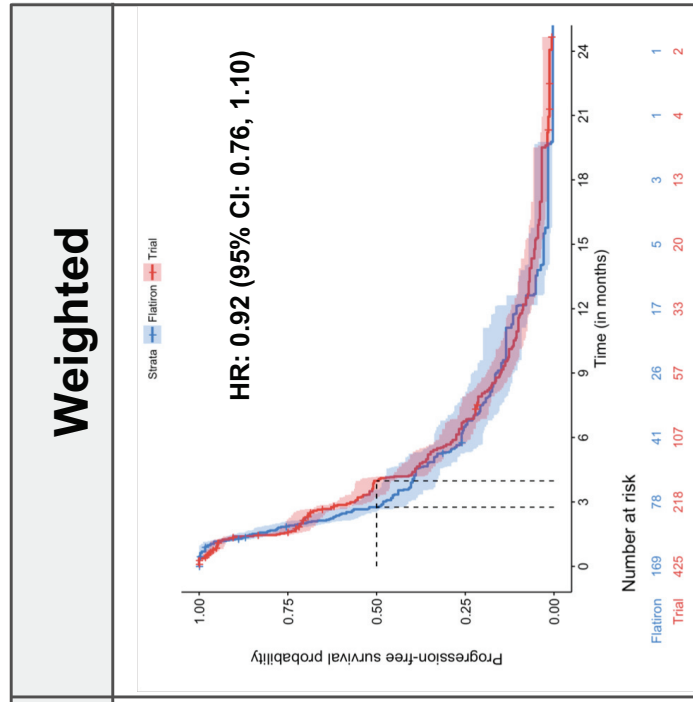
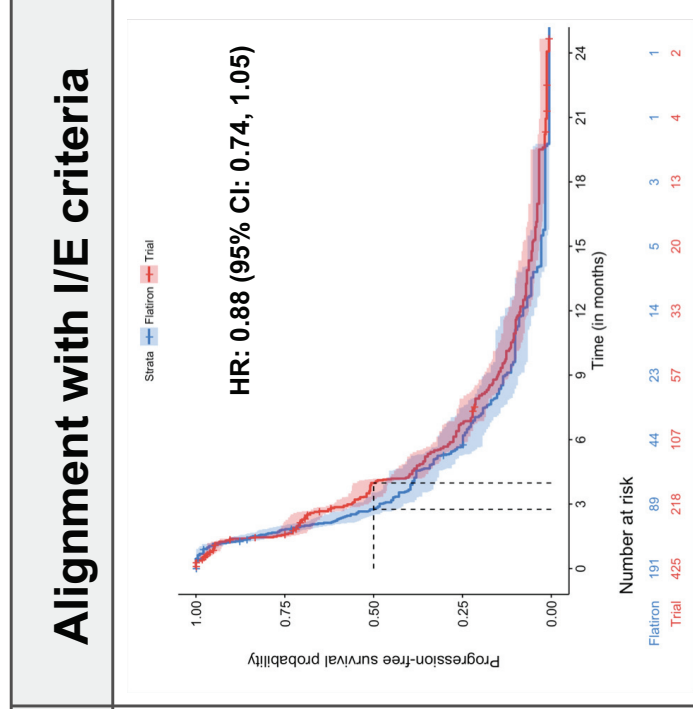
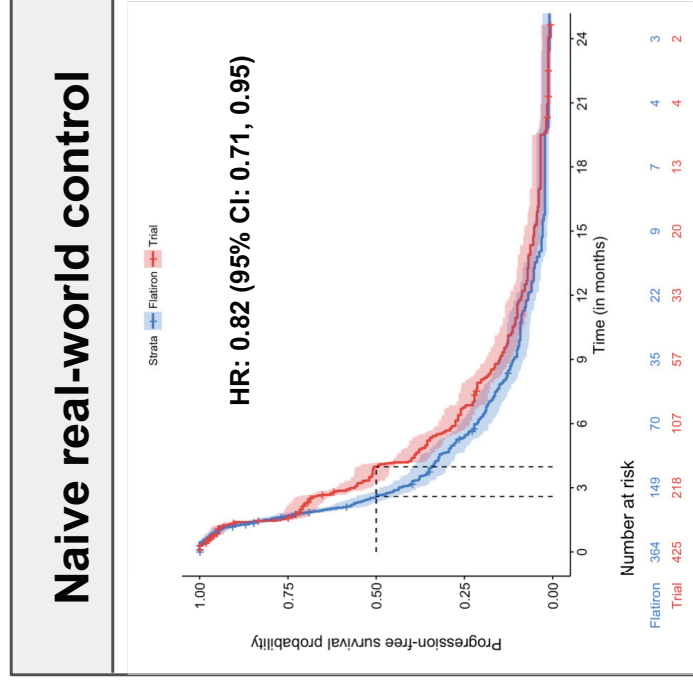
## Alignment with I/E criteria



## Weighted



# Progression-free survival (secondary endpoint)





## Planned sensitivity analyses

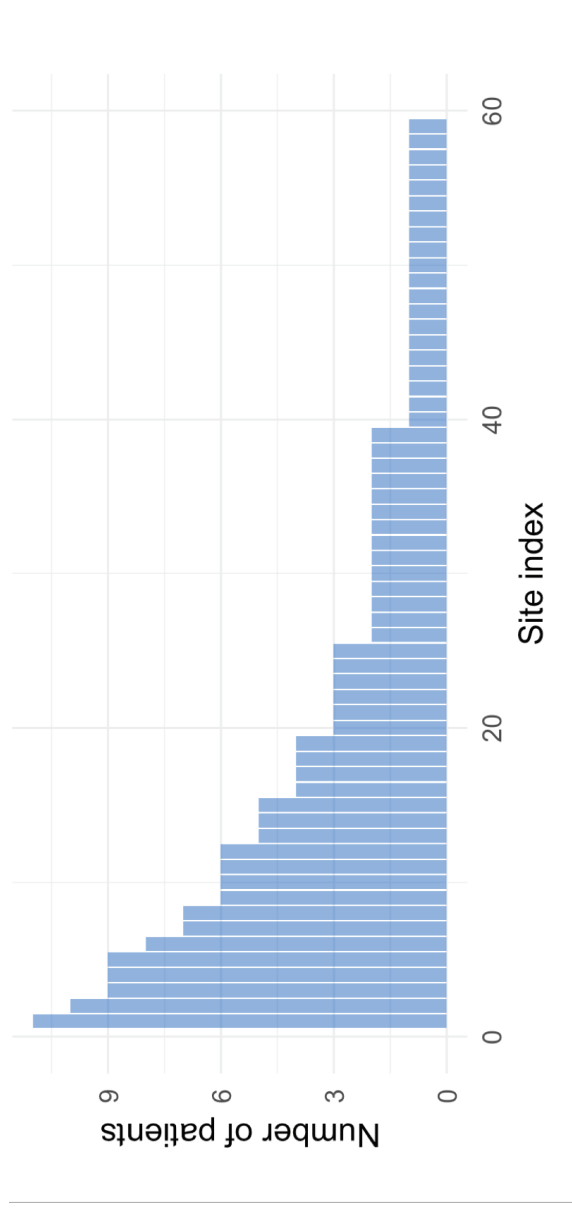
Comparing patients in original trial's control arm versus :

| Trial     | Main analyses<br>(weighted)<br>HR (95% CI) | Excluding<br>patients with<br>missing ECOG<br>performance<br>status<br>HR (95% CI) | Excluding<br>patients with<br>missing<br>laboratory results<br>used to define<br>organ function<br>HR (95% CI) | Excluding<br>patients treated<br>before trial<br>enrollment period<br>started<br>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)

Distribution of Number of patients from each sites in the Flatiron network

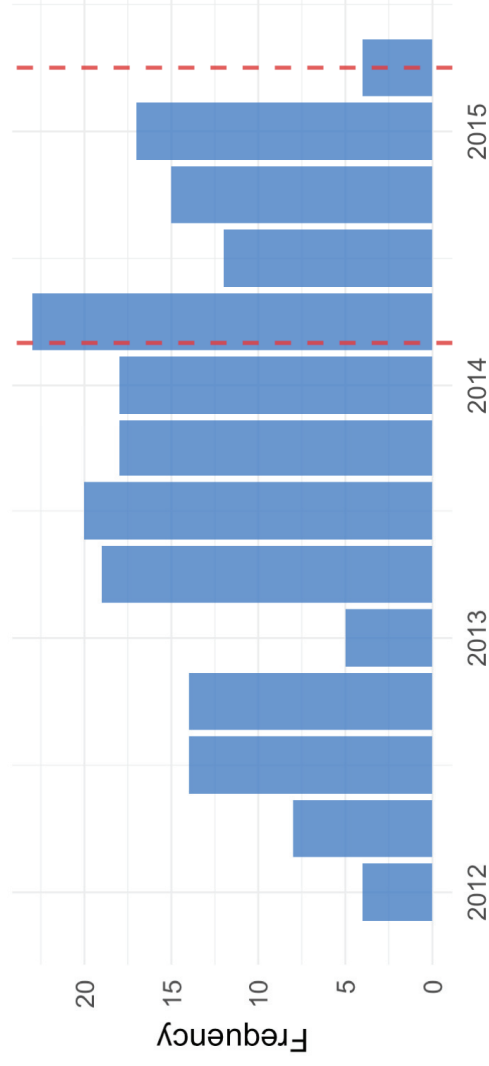


- Total 59 sites
- Both academic and community are well represented

# Diagnostics

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

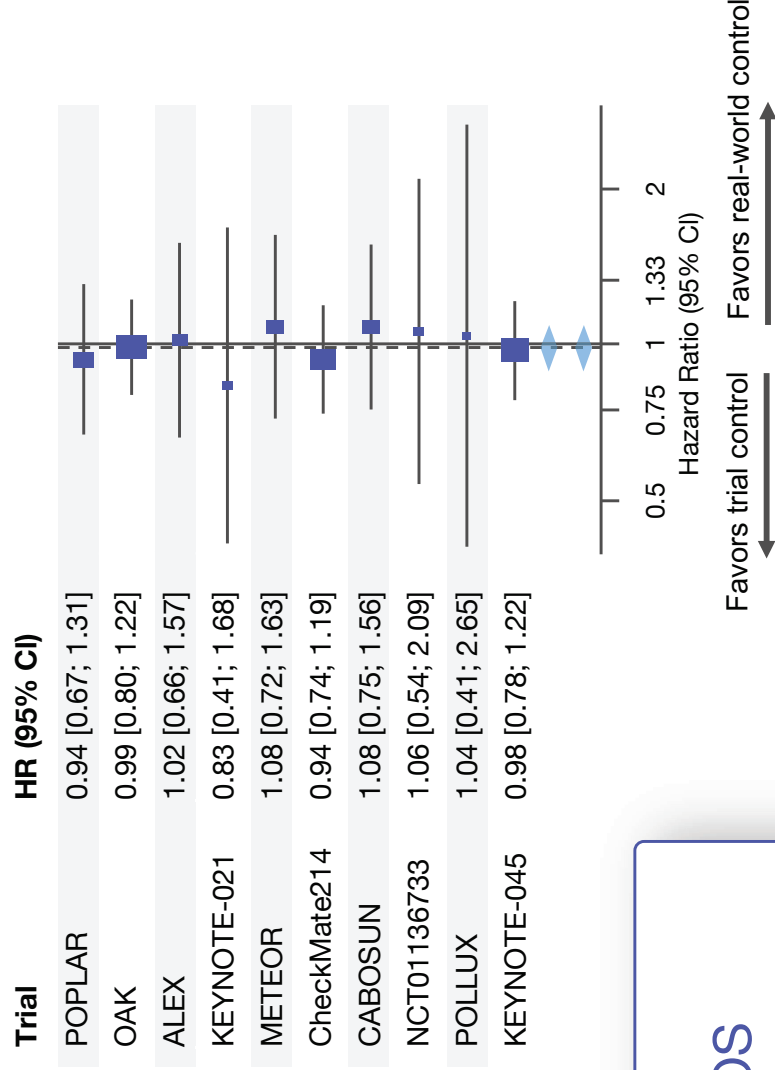
## Follow-up Comparisons



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

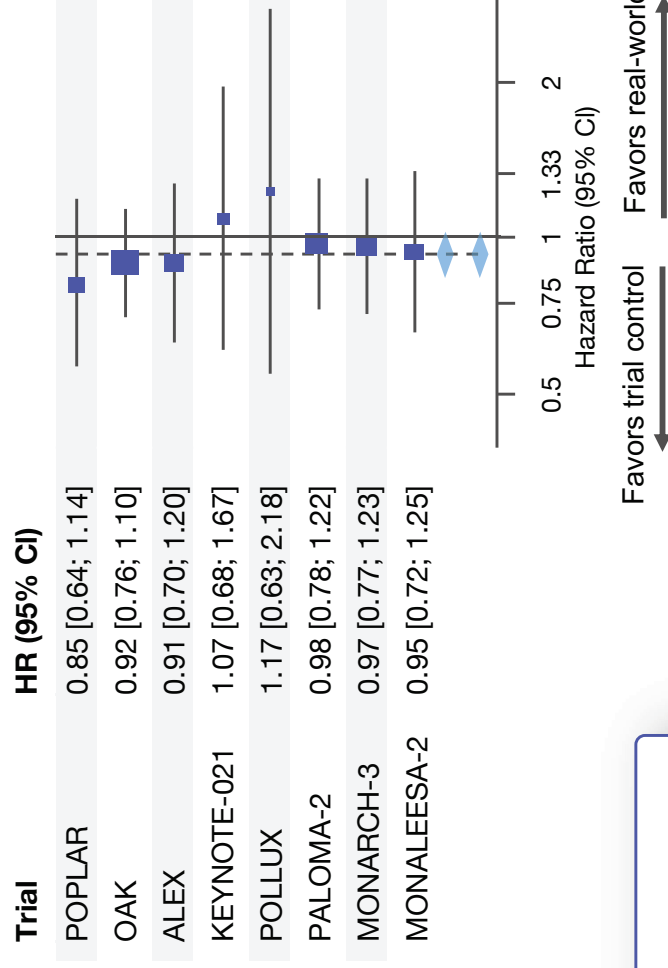
- Concurrent patient cohort
- Possible due to large sample size
- HR=1.01 if cohort is restricted

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



### Aligning with a trial's eligibility criteria and timing of enrollment

- Requires transparent application of trial eligibility criteria to real-world patients

## Weighting or Matching



### Matching or weighting real-world patients to trial patients

- Predefined approach to improve balance between non-randomized groups

## Comparing Outcomes



### Comparing real-world and clinical endpoints

- Validation of real world endpoints
- Diagnostics and sensitivity analyses

Cohort  
selection

Weighting or  
Matching

Comparing  
Outcomes

### Comparability of enrollment timeframe and follow-up

- 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

### 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

## Weighting or Matching

Comparing  
Outcomes

### Selection of covariates

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

### 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

Weighting or  
Matching

Comparing  
Outcomes

### Validity of real-world endpoints and analysis

- 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

### “Threshold crossing” framework,

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

## Conclusion

- 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

Thank you