



2026 CDISC China Interchange

From SAS to R: A Pioneer's Journey with Admiral

Implementing ADaM for a Prostate Cancer Study with PCWG-Modified RECIST

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Speaker



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- Senior Principal Statistical Programmer at Novartis
- 9+ years of clinical trial programming across multiple therapeutic areas, with a focus on immunology and oncology
- Extensive experience in SAS and R programming, including ADaM development and R/Admiral implementation

Agenda

01

Admiral & Project Context

02

From Learning to Application

03

Why Prostate Cancer Is Different

04

Real Application & Customization

05

Lessons Learned



What you'll take away

- How Admiral adapts to tumor-specific criteria
- Practical ideas for ADRS / ADTTE derivations
- How to customize Admiral without losing structure

Background & Context

Industry Context

- Pharmaverse & open-source R rising in clinical programming
- Growing industry experience with R-based regulatory submissions
- Admiral provides an industry-backed, CDISC-compliant ADaM solution

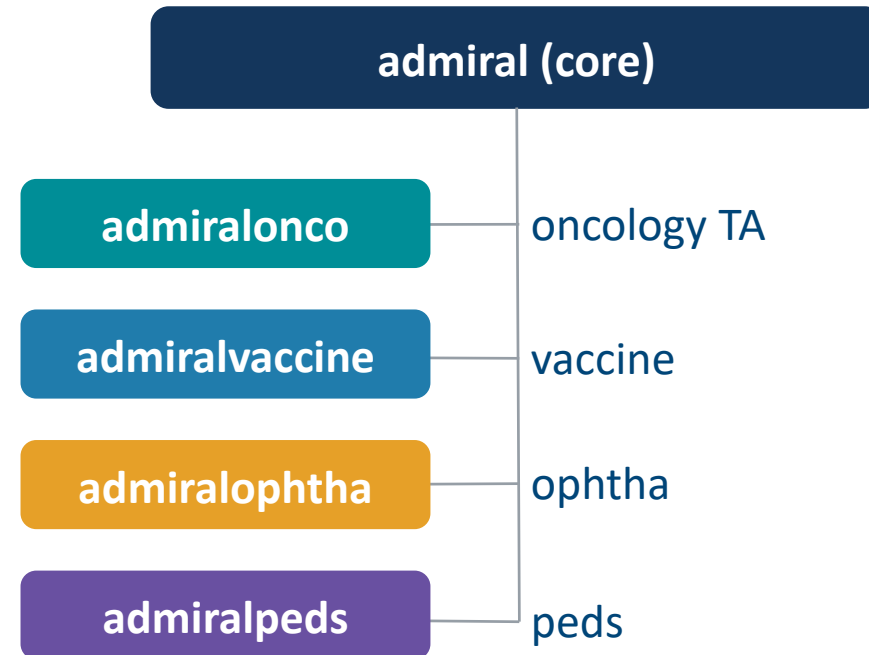
Our Context

- Novartis transitioning from SAS to R
- Selected as a pilot R-based programming project
- Prostate cancer study needed PCWG-modified RECIST
- Standard SAS macro did not fully apply

Admiral: ADaM in R Asset Library

- Open-source R package for ADaM dataset generation
- Developed collaboratively across pharma companies and CROs under the pharmaverse initiative
- CDISC-aligned, Modular, Traceable

Pharmaverse Ecosystem



My Learning Path

STEP 1: Explore

 pharmaverse.github.io

- ✓ Read vignette
- ✓ Study examples
- ✓ Understand design

STEP 2: Replicate

 **Replicate on a pilot study**

- ✓ Worked with known expected results
- ✓ Safe environment to experiment
- ✓ Applied admiral to a standard RECIST scenario

STEP 3: Apply

 **Real project: Prostate cancer study**

- ✓ Adapt to PCWG
- ✓ Iterate & refine

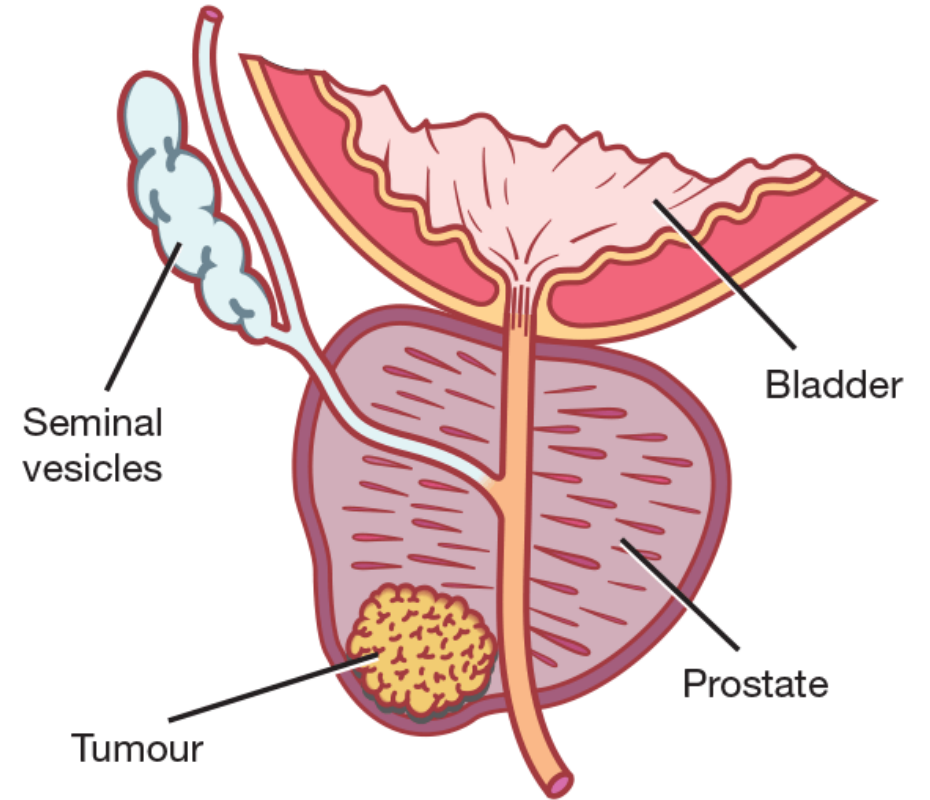
Prostate Cancer: Disease Characteristics

- **Epidemiology*:**

- The most-commonly diagnosed cancer
- The fourth most common cause of death in worldwide

- **Disease Characteristics:**

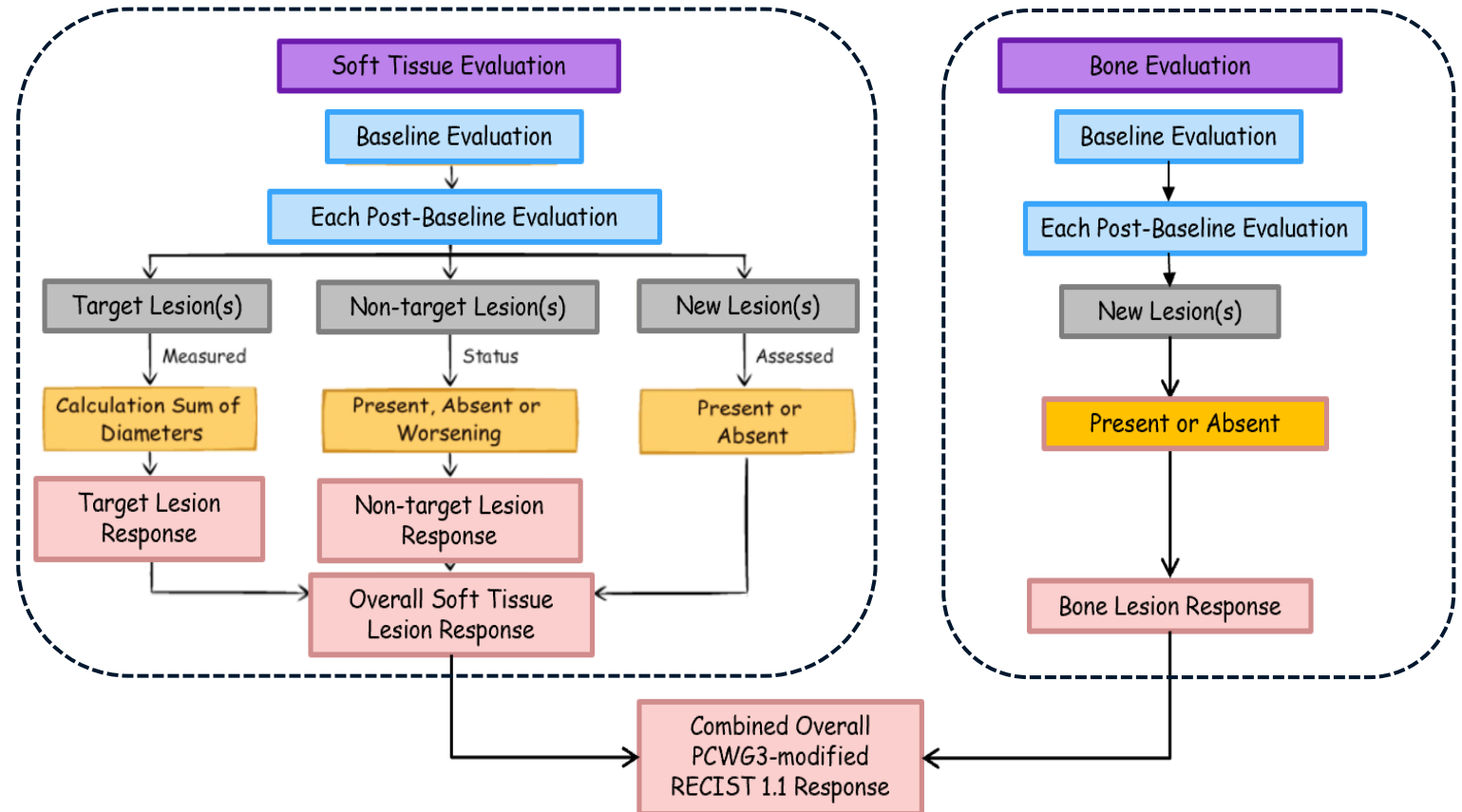
- Prostate cancer commonly metastasizes to both soft tissue and bone
- Bone lesions are a major component of disease burden and disease progression
- Assessment of bone disease differs from assessment of measurable soft-tissue lesions



Why Response Assessment Is Different in Prostate Cancer

- **Impact on Response Evaluation:**

- Standard RECIST primarily focuses on measurable soft-tissue lesions
- Bone disease requires a separate assessment approach
- Soft-tissue and bone assessments must be combined to determine overall response



Our Approach: From Decision to Implementation

The Situation

- Team moving toward R
- PCWG-modified RECIST needed
- Existing SAS RECIST macro → optimized for standard RECIST, not designed for PCWG

Options We Considered (team discussion)

A Deeper Realization

- Starting with R + Admiral gave us the flexibility
 - design ADaM around PCWG logic
 - rather than force PCWG into a macro built for a different scenario

Option 1: Adapt existing SAS macro

- Macro optimized for standard RECIST
- Extending to PCWG would require significant rework

Option 2: Build from scratch in R

- Maximum flexibility, but reinvents the wheel

Option 3: Adopt Admiral + customize

- CDISC-compliant foundation
- Modular & flexible for PCWG logic
- R-native, aligned with team's direction



Real Application: The Three Datasets I Worked On

ADRS

Response Analysis

- ✓ Best Overall Response (BOR)
- ✓ PCWG-modified RECIST logic
- ✓ admiral event() framework

ADTR

Tumor Results

- ✓ Diameter-based
- ✓ Custom code (not covered today)

ADTTE

Time-to-Event

- ✓ PFS, OS, and other endpoints
- ✓ admiral + custom wrappers

ADRS: PCWG-Modified RECIST

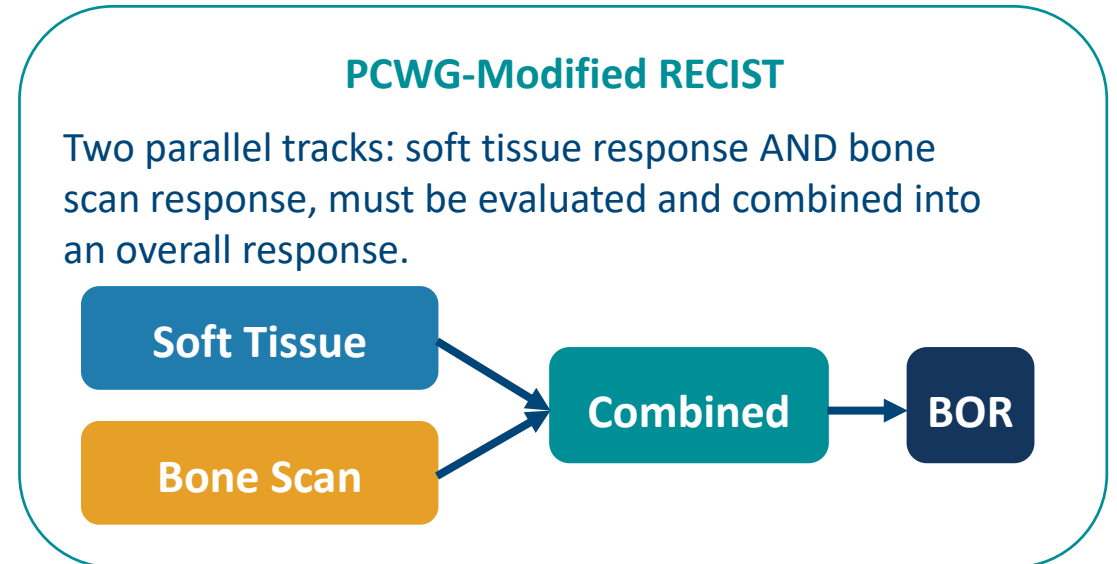
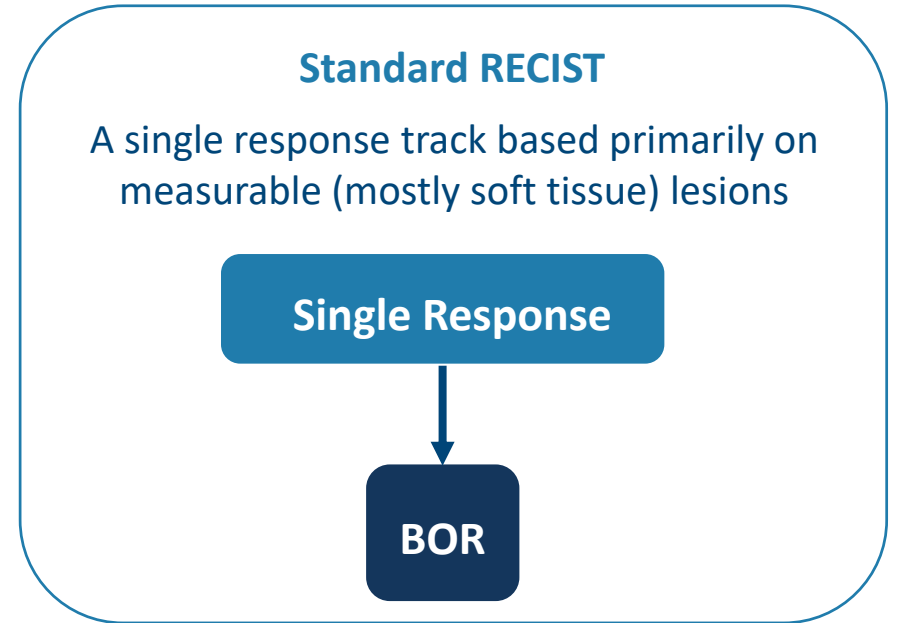
- **Why is response derivation harder for prostate cancer?**

💡 Two response tracks must be evaluated separately, then combined

- a logic beyond standard RECIST

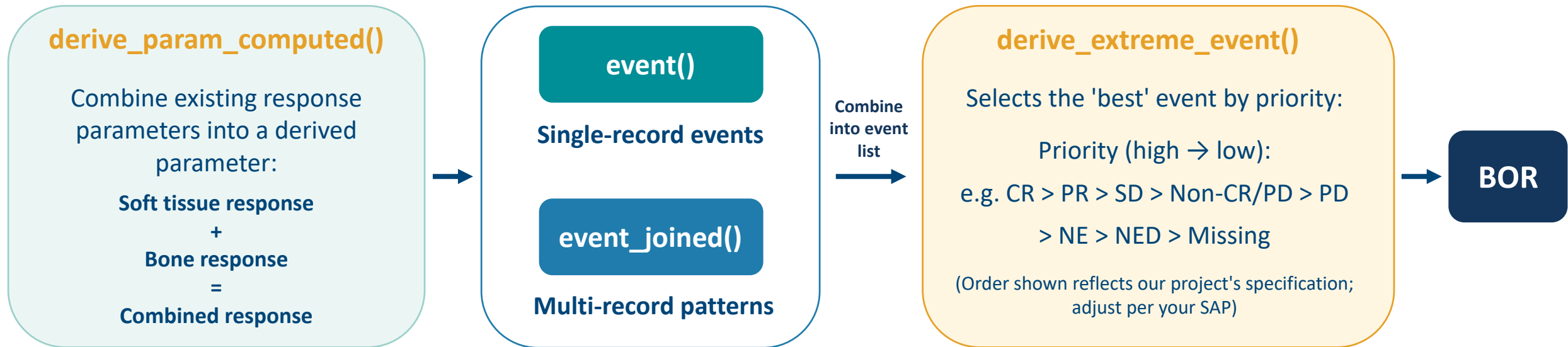
👉 The modular framework in admiral fits this pattern well

- Combine response assessments, define each response outcome as an event, then apply the project-defined priority to derive BOR



ADRS: A Modular Derivation Framework

- The Design Pattern: Define components → Define events → Combine into BOR



💡 Where the project-specific logic lives:

- Response combination → `derive_param_computed()`
- Event definitions → `event()` & `event_joined()`
- BOR prioritization → `derive_extreme_event()`

ADRS: Customizing the Framework

Example 1: Confirmation with a project-specific tolerance rule

- ✓ Our project allows one intervening SD between two PRs, per SAP
- ✓ Sequence: PR → SD → PR → PR
- 👉 event_joined() + count_vals() can match this pattern → ☒ Confirmed PR

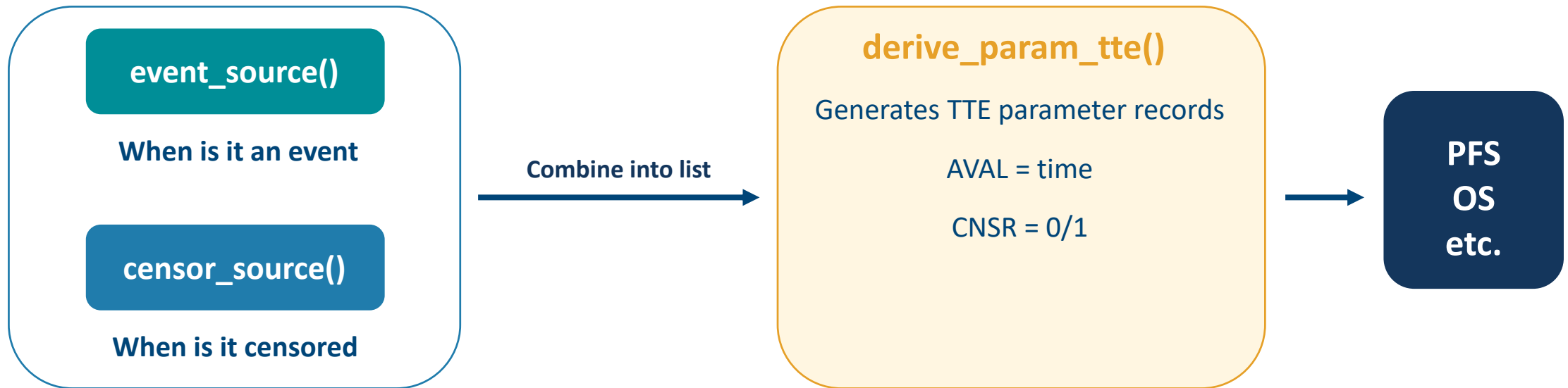
Example 2: Special responses (event)

- ✓ PDu - Unconfirmed Progressive Disease
- ✓ NED - No Evidence of Disease
- 👉 Each defined as its own event() with tailored conditions
→ All events plug into the same derive_extreme_event() pipeline



ADTTE: Reusing the Modular Pattern

- Similar design philosophy as ADRS: Define sources → Derive parameters



- ✓ Standard, well-documented pattern
- ✓ Flexible for defining any TTE endpoint - but the source setup is repetitive across endpoints

Going Further: Building Wrappers on Top of admiral

Opportunity for a Convenience Layer:

- `derive_param_tte()` requires:
 - Each event/censor defined separately
 - Manually organized into TWO separate lists:
(events = list(...), censors = list(...))
- With many events/censors, maintaining two parallel source lists created an avoidable organizational risk for our team

Wrapper 1

Build source (event/censor)

- One unified function
- Type-tagged: event or censor
- All definitions live in ONE combined list

Wrapper 2

- Auto-splits by Type
- Routes to correct arguments of `derive_param_tte()`

Going Further: Building Wrappers on Top of admiral



Why we built

- ✓ Multiple TTE endpoints meant repeatedly organizing large numbers of event and censor definitions



What improved

- ✓ Fewer chances to misplace event/censor definitions
- ✓ Cleaner code organization (one list, not two)



What did not change

- ✓ Individual definitions are still verbose
- ✓ They are convenience layers, not new capabilities



Why it matters

- ✓ Modular design allows targeted workflow improvements without changing the underlying derivation approach

Key Lessons Learned

Structure is a strength

For standard, well-defined endpoints (like oncology response), admiral's structure promotes consistency, traceability, and CDISC alignment

Complex logic is manageable

Even for complex cases like PCWG, Admiral gave us enough flexibility to adapt, without rebuilding everything from scratch

✓ The sweet spot: constrained yet adaptable

Modularity comes with trade-offs

Admiral works like building blocks: define each piece clearly, then assemble.

✓ Each step is transparent and extensible

⚠ Code volume increases: especially for endpoint-heavy datasets like ADTTE



Acknowledgement

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- Study team members for the opportunity to apply Admiral in a complex prostate cancer study
- Programming and statistics colleagues for their partnership and feedback
- Everyone who contributed ideas, reviews, and encouragement along the way

Thank You!

