

Stop the abuse: A plea for a more principled approach to the analysis of adverse events

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How do we typically analyze «safety» in a clinical trial?



Typical reporting of AEs in a clinical trial

Category	All Adverse Events	
	Obinutuzumab Group (N= 595)	Rituximab Group (N= 597)
Infection*	460 (77.3)	418 (70.0)
Neutropenia	301 (50.6)	269 (45.1)
Infusion-related event†		
Any event	406 (68.2)	349 (58.5)
Antibody-related event	353 (59.3)	292 (48.9)
Tumor lysis syndrome	6 (1.0)	3 (0.5)
Cardiac event‡	78 (13.1)	58 (9.7)
Thrombocytopenia	68 (11.4)	45 (7.5)
Second neoplasm§	43 (7.2)	30 (5.0)
Nonmelanoma skin cancer	18 (3.0)	14 (2.3)
Hematologic event¶	6 (1.0)	0
Other	22 (3.7)	18 (3.0)
Myelodysplastic syndrome	2 (0.3)	0
Gastrointestinal perforation	4 (0.7)	3 (0.5)
Hemorrhagic event	57 (9.6)	62 (10.4)

N Engl J Med 2017;377:1331-44.

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What do these proportions estimate?

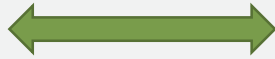
- Incidence proportion for a group of patients in interval from 0 to t:

$$\widehat{IP}(t) = \frac{\text{Number of patients with AE in } [0,t]}{\text{Total number of patients}}.$$

- Estimates probability of AE happening in [0, t] and that **this AE is observed**.
- «this AE is observed»: implies that if we have (administratively) censored patients in our dataset then **$\widehat{IP}(t)$ underestimates true AE probability**.
- Interpretable if all patients have ~ same follow-up.



What do we want to report with safety analyses?

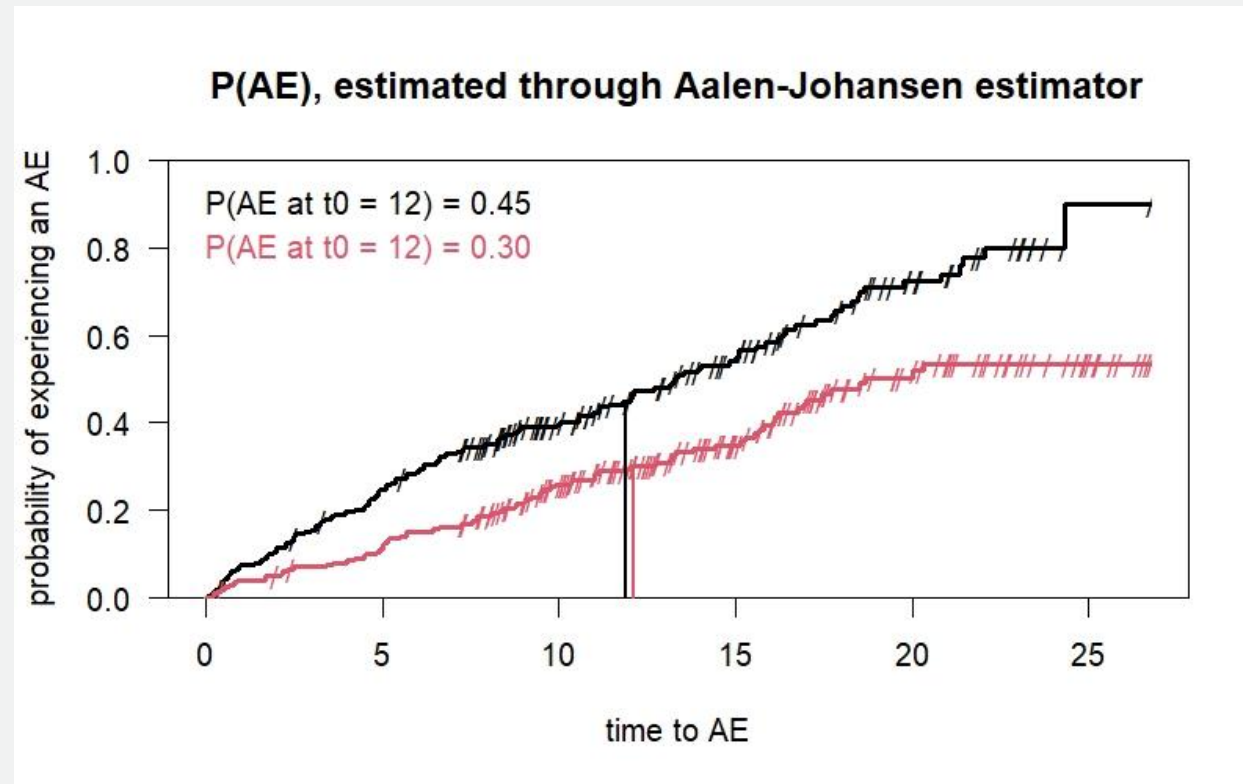
- Signal detection  assessment of AE risk.
- Here: assess AE risk for selected AEs.
- Probability of experiencing an AE, $P(\text{AE})$: **estimand**.

Challenges for assessment of AE risk

- (Administrative) **censoring**: patients not observed on entire $[0, t]$.
Randomized «too late».
- **Competing events**: Clinical events that preclude the occurrence of AE.
 - Death without prior AE.
 - Treatment discontinuation leading to end of AE recording.
 - «Simply censoring» competing events gives biased estimates of probabilities.
- What is the **process that generates the events** (underlying hazard)?

Potential estimators of $P(\text{AE})$

- Consider «time to AE» → methods from survival analysis.



Can we simply use 1 – Kaplan – Meier?

- For time-to-event endpoints (OS, PFS) we typically use Kaplan-Meier estimates → properly accounts for censoring.
- Do these work also for time to AE?
- Yes – in absence of competing events.
- Competing events: 1 – KM **overestimates** true P(AE).

Potential estimators of AE risk

	Accounts for censoring	Accounts for competing events	Makes no constant hazard assumption
Incidence proportion	No	Yes	Yes
1 – Kaplan-Meier	Yes	No	Yes
Aalen-Johansen estimator	Yes	Yes	Yes

- **Aalen-Johansen:**

- Only estimator that accounts for censoring, competing events, and does not make assumption about underlying hazard.
- Advocated for in statistical literature for decades.
- «Gold standard». Available in any standard software package.



How large can the bias become in a clinical trial?



SAVVY = academia + pharma

- **S**urvival analysis for **AdV**erse events with **VarY**ing follow-up times (SAVVY):
 - **Collaborative** effort from academia and pharma.
 - Goal: improve analyses of AEs in clinical trials through use **survival techniques** that account for varying follow-up times, censoring and competing events.
- Organizational setup:
 - **Centrally developed** R / SAS macros sent to organizations.
 - Summaries computed on clinical trial datasets within organizations → **raw data never left organization**.
 - Summaries centrally meta-analyzed.

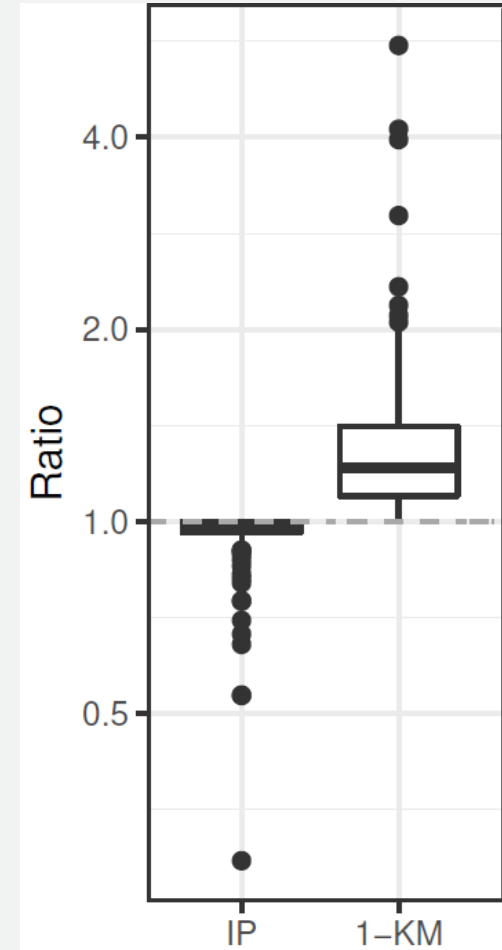
How large can the bias become in a clinical trial?

- Dataset:
 - **Ten** organizations (9 pharma + 1 academic) provided **17 trials** including **186** types of AEs.
 - Trials included between 200 and 7171 patients (median: 443).
- «Gold standard»: Aalen-Johansen estimator of $P(\text{AE})$.
- Maximum follow-up time: compute $P(\text{AE})$ at the last observed AE time.

How large can the bias become in a clinical trial?

Incidence proportion (IP)

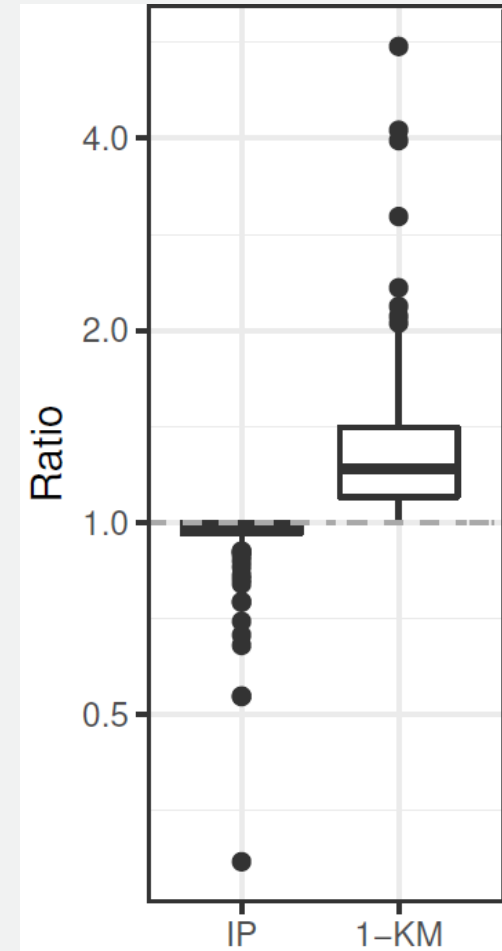
- Every observation in boxplot corresponds to ratio of IP to gold standard for a type of AE.
- Ratio = 1: IP gives same estimate as gold standard.
- **Underestimation of P(AE) up to factor THREE!**
- Overall reasonable performance.
- IP accounts for CEs but not censoring → datasets have many CEs → little censoring.



How large can the bias become in a clinical trial?

I – Kaplan-Meier (I - KM)

- Every observation in boxplot corresponds to ratio of (1 – KM) to gold standard for a type of AE.
- Ratio = 1: (1 – KM) gives same estimate as gold standard.
- **Overestimation of P(AE) up to factor FIVE!**
- (1 – KM) accounts for censoring but not CEs.



Does it impact decisions?

- SmPC frequency categories: very rare (< 0.01%), rare (< 0.1%), uncommon (< 1%), common (< 10%), very common ($\geq 10\%$).
- Compare frequency category from IP or (1 – KM) to those from gold standard.

Does it impact decisions?

		gold-standard Aalen-Johansen				
		very rare	rare	uncommon	common	very common
incidence proportion	very rare	6				
	rare		0			
	uncommon			6		
	common				86	2
	very common					86
1-Kaplan-Meier	very rare	6				
	rare		0			
	uncommon			4		
	common			2	72	
	very common				14	88

- **Potential impact on labeling!**

How about estimation of relative effects?

- Categorization according to German Institute for Quality and Efficiency in Health Care (IQWiG).

		HR Cox for AE			
		(0) no effect	(a) minor	(b) considerable	(c) major
RR gold-standard Aalen-Johansen	(0) no effect	42	3	3	1
	(a) minor	9	2	1	
	(b) considerable	4	1	3	2
	(c) major	2		4	17

- **Potential impact on labeling!**

Conclusions

- Start with the **scientific question**: Signal detection? Assessment of AE risk?
- Respect **data structure**: censoring? Competing events?
- Estimand!
- **Select appropriate estimator.**
- Estimation of $P(\text{AE})$ in one cohort: Incidence proportion and $(1 - \text{KM})$ **biased** in presence of censoring or competing events.
- Comparing AE risk between two arms in RCT: differences between estimators become more emphasized.
- Template for:
 - Academia – pharma partnership.
 - Pool results of 17 RCTs without need for individual patient data sharing.



SAVVY: next steps and resources

- Estimate $P(\text{AE})$ in selected indications accounting for data structure.
- Statistical Analysis Plan:
 - Stegherr et al (2021) Biom J <https://doi.org/10.1002/bimj.201900347>
- Methods:
 - Preprint <https://arxiv.org/abs/2001.05709>
- Results:
 - One-sample case: <https://arxiv.org/abs/2008.07883>
 - Two-sample case: <https://arxiv.org/abs/2008.07881>
- Example R code (markdown file):
https://numbersman77.github.io/AEprobs/SAVVY_AEprobs.html



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