

# CVCT

#22

**Guiding iterative decision-making in anticoagulant dosing including patient preferences of stroke vs bleeding avoidance: A Bayesian Machine Learning model trained on the ENGAGE AF-TIMI 48 trial**

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# Disclosures

- **CMG: Modest consulting fees (<25K) paid to Boston Clinical Research Institute**

# Background

- **We conduct trials and draw inferences from / about populations**
- **We treat individual patients, however, who have innumerable combinations of subgroups with second and higher order interactions (which can be modeled with ML/AI)**
- **We see patients in clinic and iteratively reassess their risk moving forward based upon the presence or absence of events (including multiple events) & their tolerability of a drug / device (eg a patient who has not bled is at a lower risk of bleeding moving forward)**
- **We should take into account individual patient preferences (not population preferences) with respect to ischemic / bleeding outcomes and side effects**

# Background/Objective

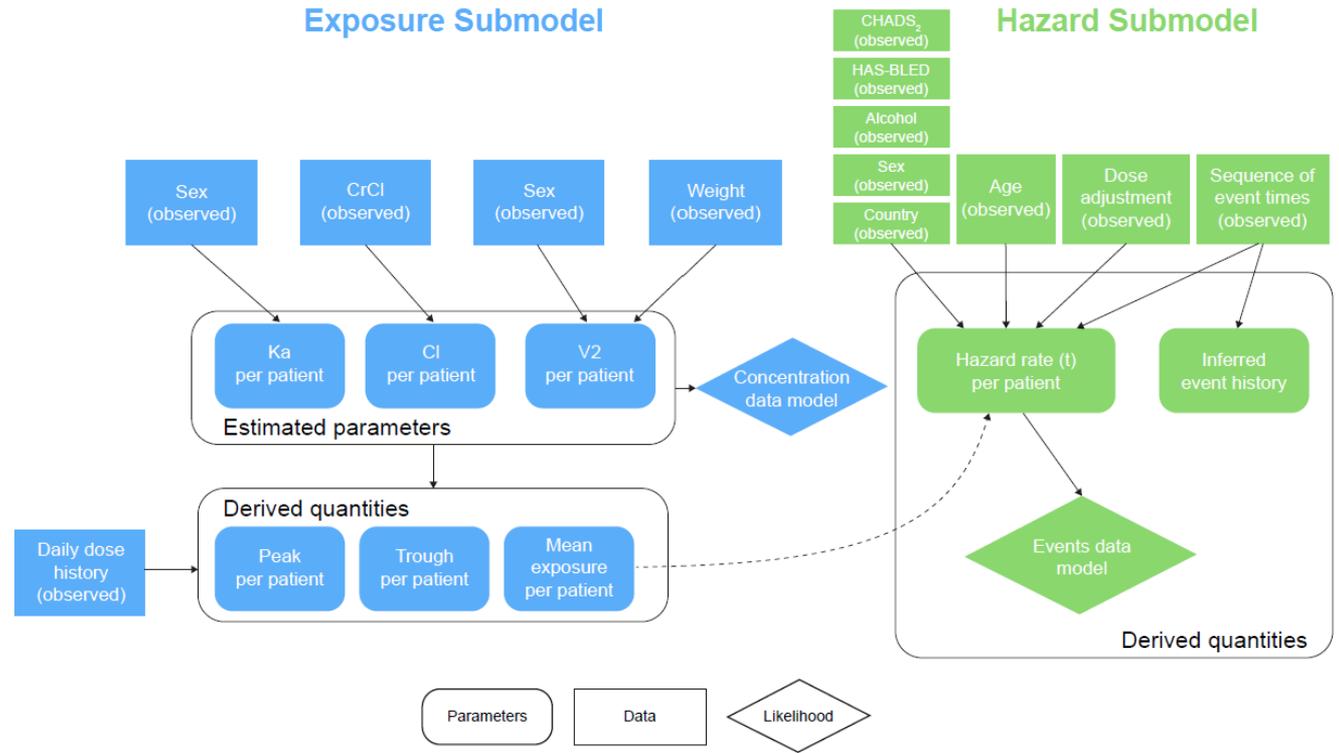
- **Anticoagulation in atrial fibrillation often relies on a single dose, does not consider changes in patient risk during chronic therapy, and does not incorporate patient preferences for risks of stroke, bleeding and death.**
- **Bayesian machine learning (ML) could improve longitudinal risk prediction and patient-centric care, enabling shared decision making.**

**Objective: We developed a Bayesian ML model (Adele\*) to guide individualized long-term dosing accounting for patient preferences and intercurrent events.**

\*Adele was named in honor of Myra Adele Logan, the pioneering African American surgeon who was the first woman to perform successful open-heart surgery.

# Methods

Adele is a Bayesian competing-risk, multistate hazard model, developed in 5,380 edoxaban-treated patients with their pharmacokinetic (PK) data from the ENGAGE AF-TIMI 48 trial (NCT00781391)<sup>1-7</sup>



\*Adele was named in honor of Myra Adele Logan, the pioneering African American surgeon who was the first woman to perform successful open-heart surgery.

1. Le-Rademacher JG et al. *Curr Epidemiol Rep.* 2022;9:183–189; 2. de Wreede LC et al. *J Stat Softw.* 2011;38:1–30; 3. Brookmeyer & Abdalla. *Stat Med.* 2019;38:1558–1565; 4. Neumann JT et al. *Lancet Healthy Longev.* 2022;3:e89–e97; 5. Hougaard P. *Lifetime Data Anal.* 1999;5:239–264; 6. Meira-Machado L et al. *Stat Methods Med Res.* 2009;18:195–222.; 7. Giugliano RP et al. *N Engl J Med.* 2013;369:2093–2104.

# Methods

To demonstrate individual patient predictions, we selected 2 patients from the ENGAGE AF-TIMI 48 trial to illustrate the utility of Adele.

## Adele's Applications

### #1: Predicting hazards of subsequent events (3-year)

- Conditional on an observed event at 9 months (ICH or an IS)
- Cumulative hazards were estimated by generating more than 50,000 event sequences for each patient

### #2: Cumulative weighted event rates (3-year) according to hypothetical patient preferences

- Predict optimal dosing across three edoxaban doses (60mg, 30mg, 15mg) based on patient characteristics and hypothetical patient outcome preferences (e.g., to either avoid CV death, avoid disability, avoid MB, or avoid IS/ICH)
- Outcomes were differentially weighted according to a hypothetical patient's preference

## Patient Examples

### 80-year-old female with persistent AF (Patient A)

- body weight: 52 kg
- CrCl: 42 mL/min
- modified HAS-BLED score: 2
- CHADS2 score: 3

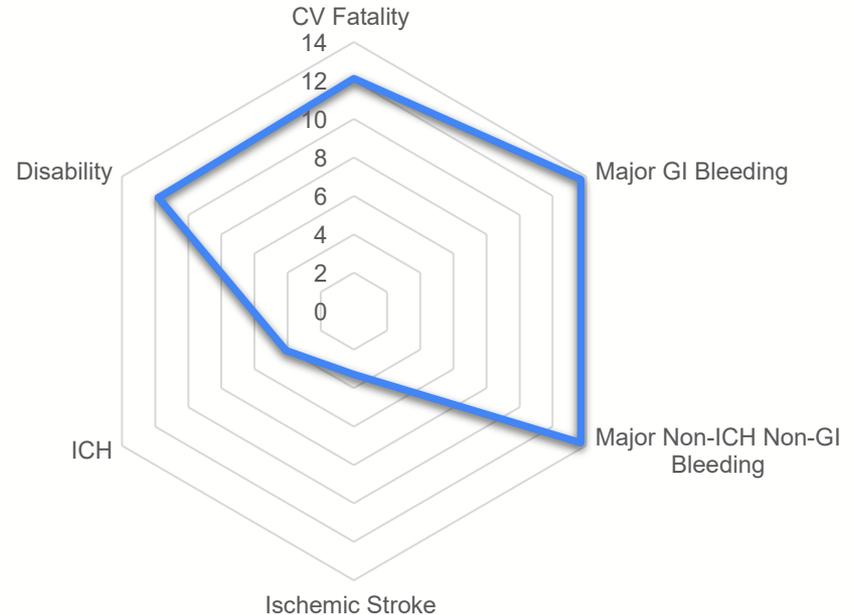
### 72-year-old male with permanent AF (Patient B)

- body weight: 80 kg
- CrCl: 73 mL/min
- modified HAS-BLED score: 2
- CHADS2 score: 4

# Results

**Model Performance: Adele outperformed standard Kaplan-Meier estimators across all clinical events, improving 3-year prediction (concordance-index) in the test sample.**

- **CV fatality: +12.1%**
- **Disability: +11.8%**
- **Major GI bleeding: +13.7%**
- **Major non-ICH non-GI bleeding: +13.7%**
- **Ischemic stroke: +3.3%**
- **ICH: +4.1%**



**% Improvement vs Kaplan-Meier**

# Results

## Predicting hazards of subsequent events (3-year)

### Patient A:

#### At randomization:

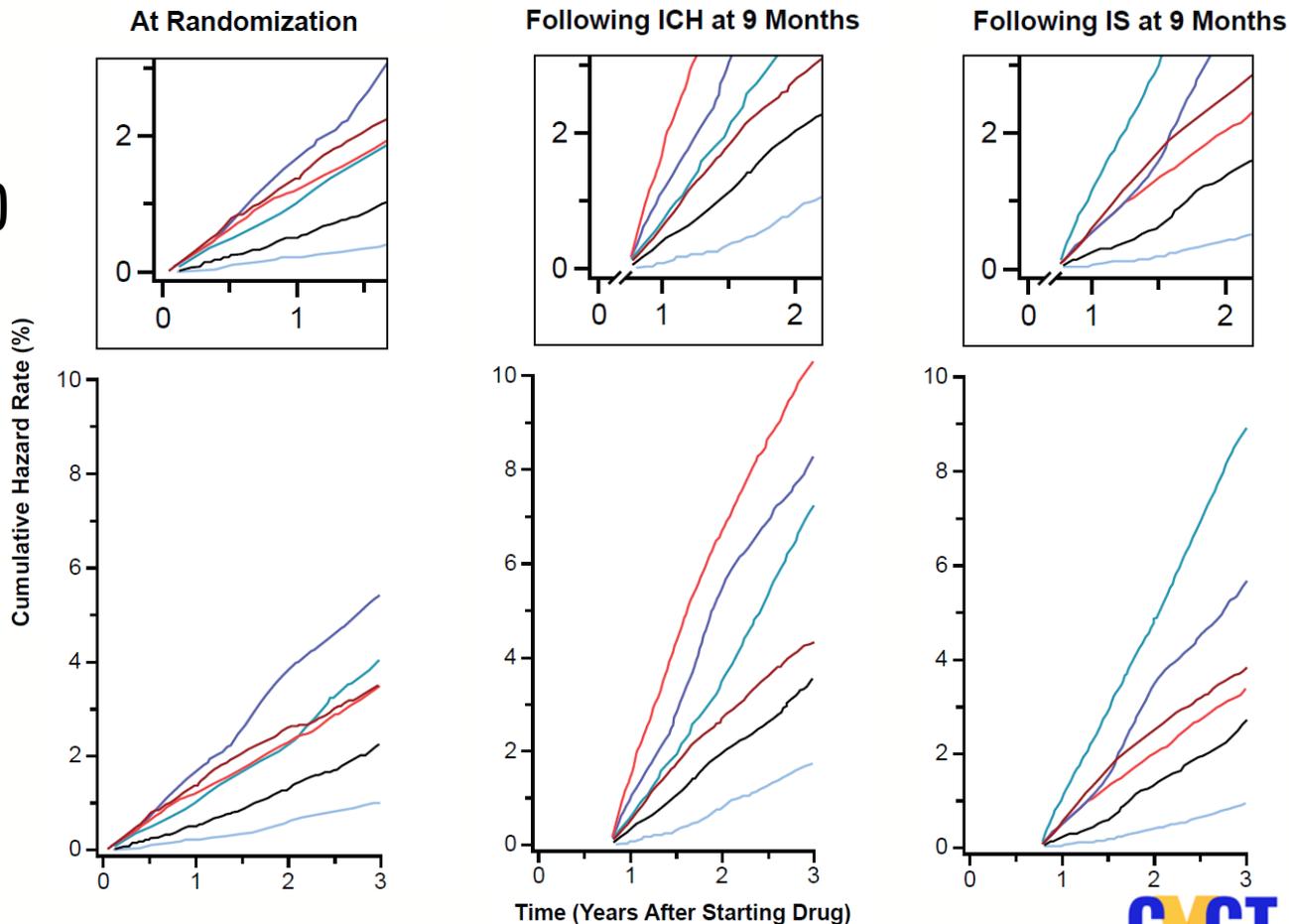
- IS represented the highest baseline risk of ~5% over three years

#### Following ICH (at 9 months):

- Major GI bleeding emerged as the highest risk, reaching ~10%

#### Following IS (at 9 months):

- Disability became the predominant risk, peaking at ~8%-9%



# Results

## Predicting hazards of subsequent events (3-year)

### Patient B:

#### At randomization:

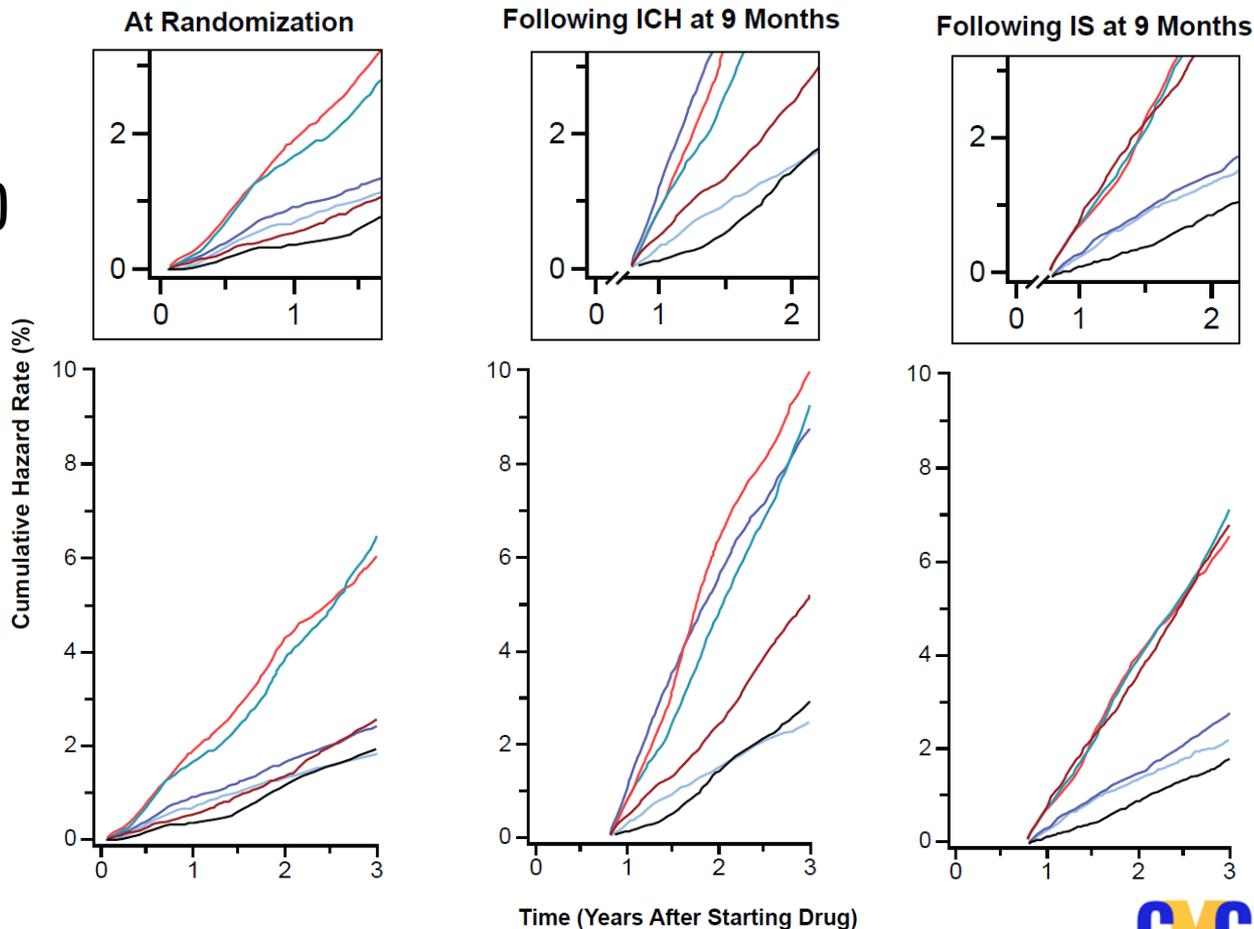
- Major GI bleeding and disability were the highest baseline risks (~6%)

#### Following ICH (at 9 months):

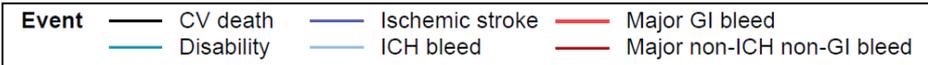
- Major GI bleeding and disability emerged as the highest risks (~9-10%)

#### Following IS (at 9 months):

- Major GI bleeding, disability, and major non-ICH non-GI bleeding became the predominant risks (6.5-7%)



GI, gastrointestinal; ICH, intracranial hemorrhage; IS, ischemic stroke.

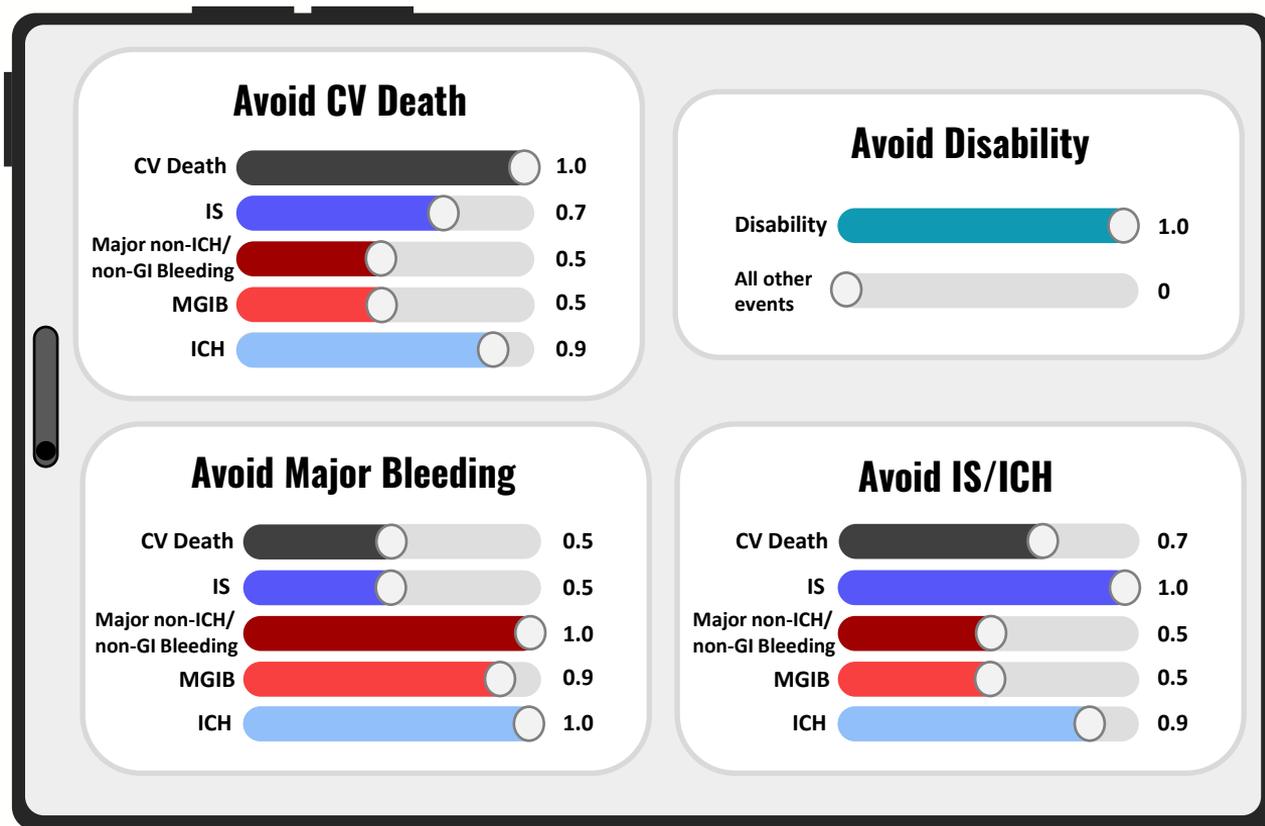
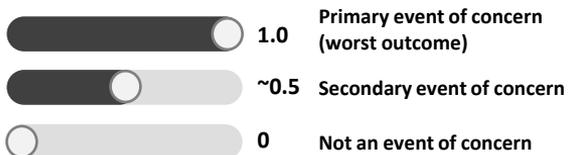


# Methods

## Patient Preference: Weighting Events

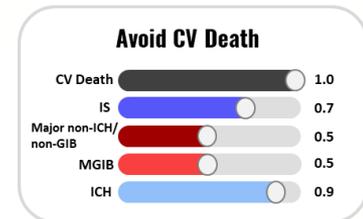
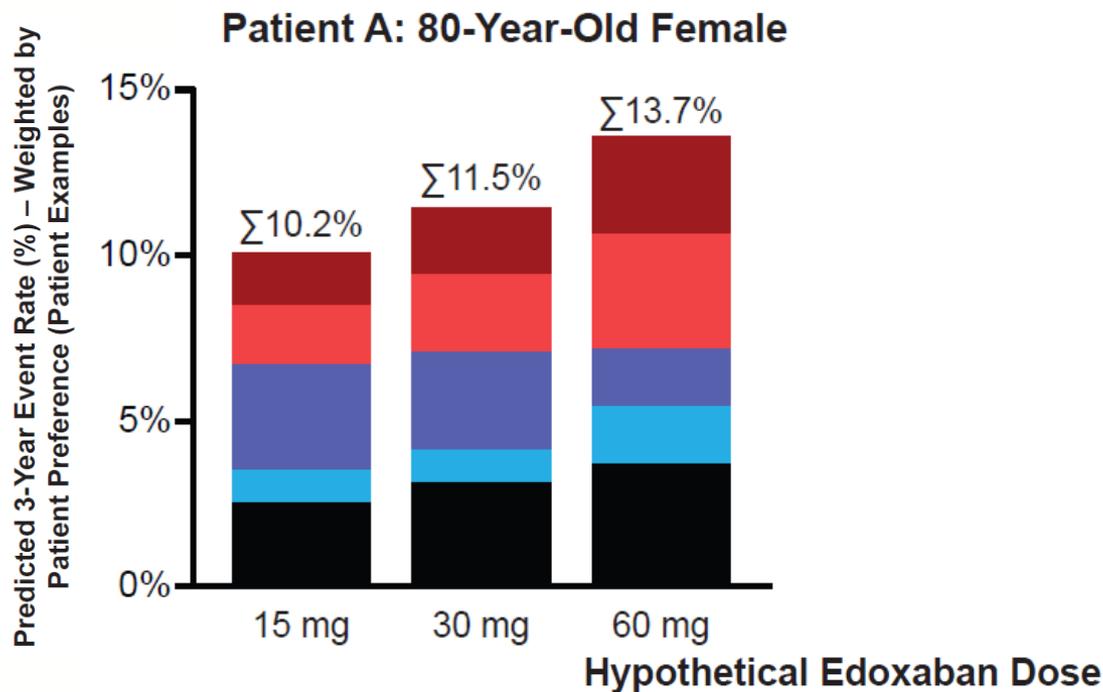


### Interpreting Hypothetical Patient Weighting



# Results – Avoid CV Death

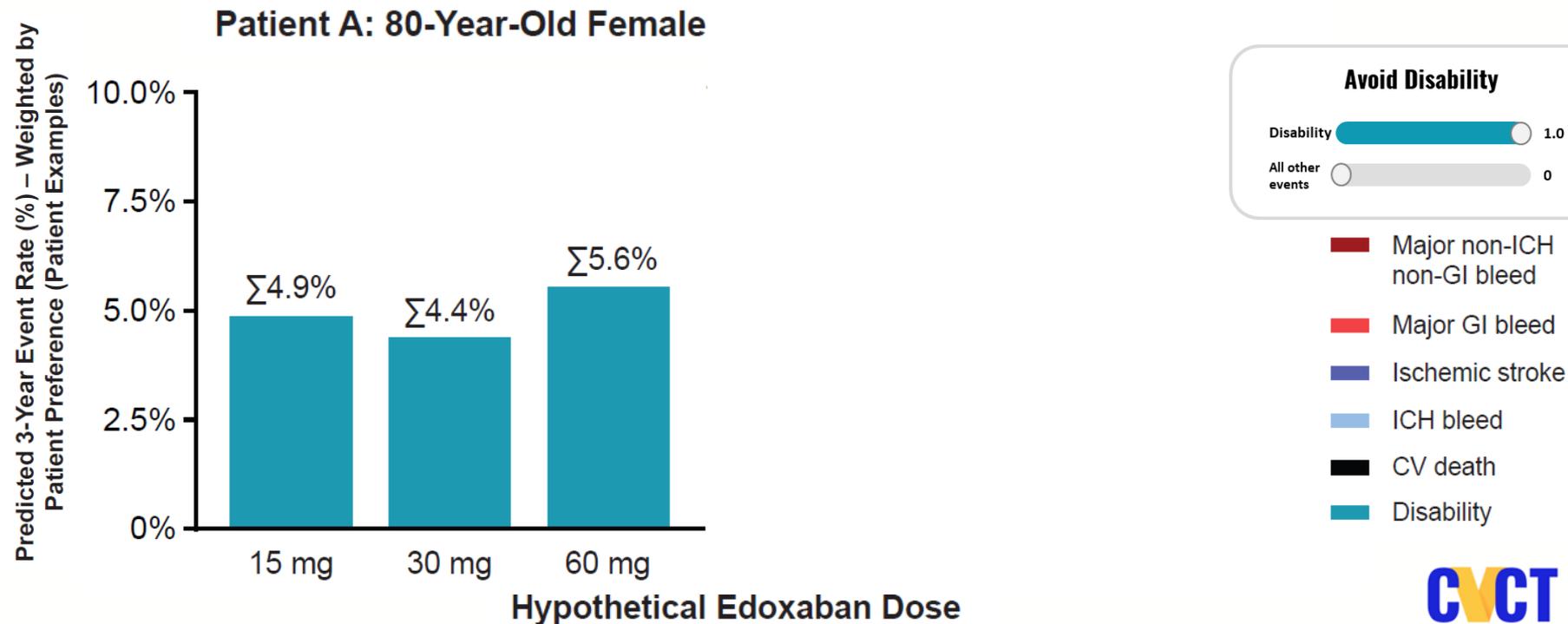
Cumulative weighted event rates (3-year) according to hypothetical patient preferences



- Major non-ICH non-GI bleed
- Major GI bleed
- Ischemic stroke
- ICH bleed
- CV death
- Disability

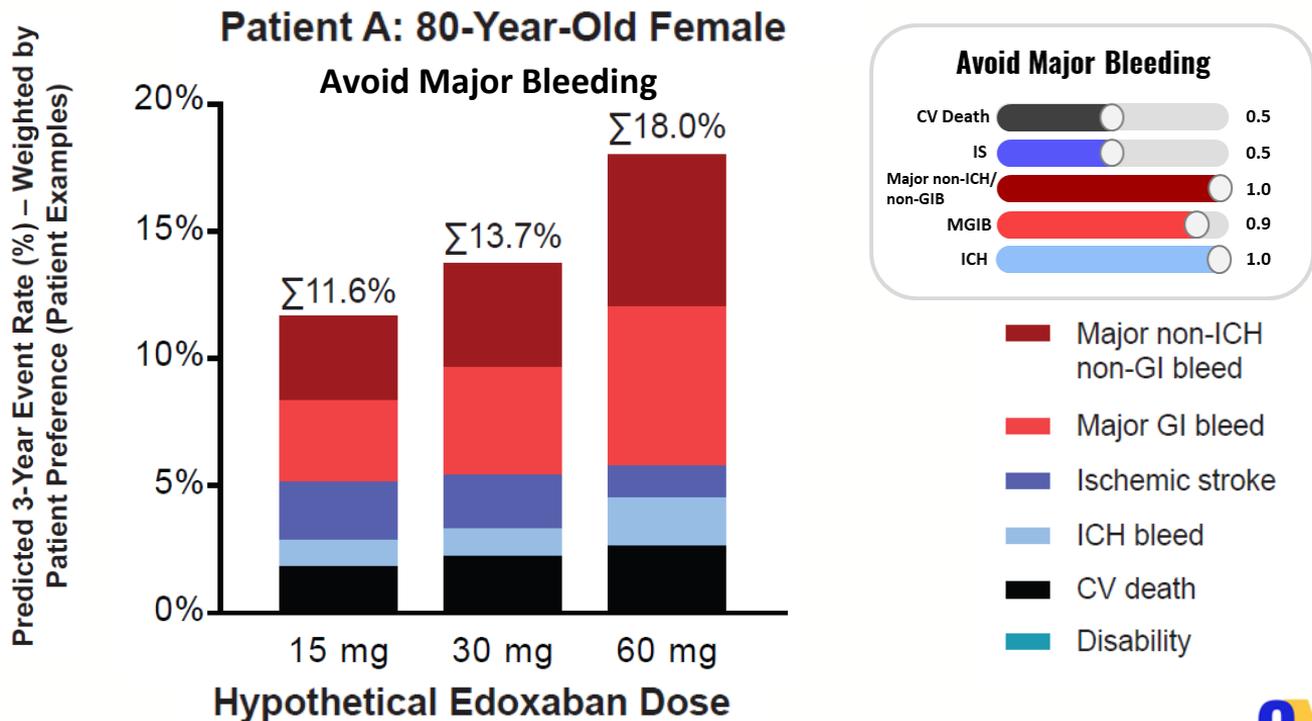
# Results – Avoid Disability

Cumulative weighted event rates (3-year) according to hypothetical patient preferences



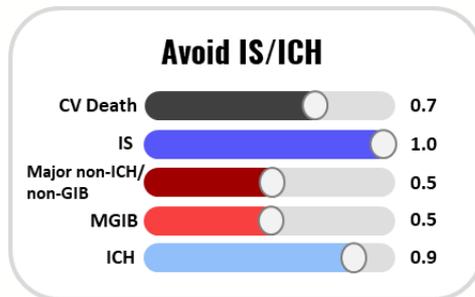
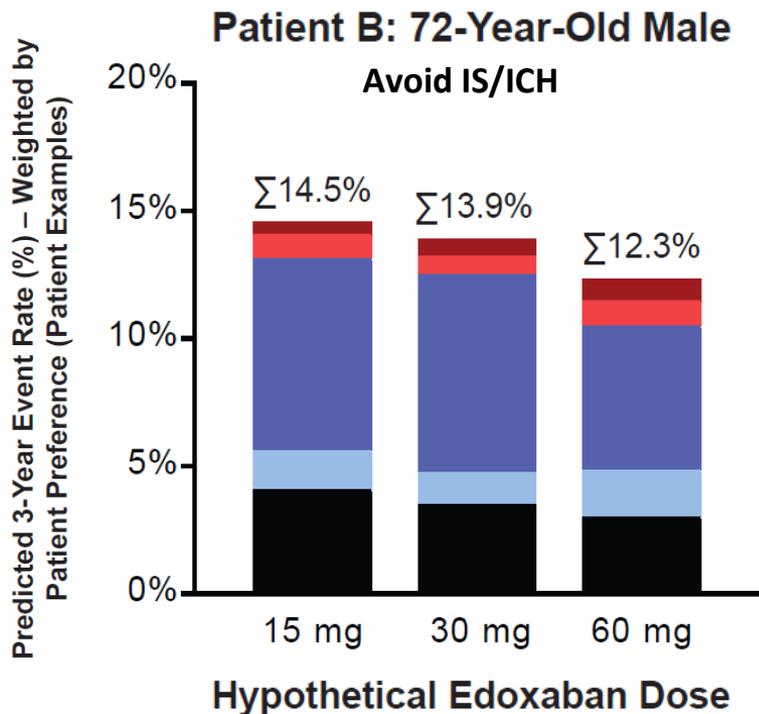
# Results – Avoid Major Bleeding (Patient A)

Cumulative weighted event rates (3-year) according to hypothetical patient preferences



# Results – Avoid IS/ICH (Patient B)

Cumulative weighted event rates (3-year) according to hypothetical patient preferences



- Major non-ICH non-GI bleed
- Major GI bleed
- Ischemic stroke
- ICH bleed
- CV death
- Disability

# Conclusions

- **Adele represents a state-of-the-art Bayesian ML framework, leveraging clinical factors and PK data, to enable longitudinally adaptive, preference-weighted, event predictions, and thus patient centric care.**

**This paradigm supports iterative, shared decision-making, offering a path toward potentially safer and more effective anticoagulation therapy and should be applicable to other chronic treatments.**

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Daiichi-Sankyo



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## Journal Pre-proof



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