

# Validation of a CKD progression risk prediction model in the FIDELITY dataset population

N. TANGRI<sup>1,2</sup>, T. FERGUSON<sup>1,2</sup>, S. J. LEON<sup>2,3</sup>, S. D. ANKER<sup>4,5</sup>, B. PITT<sup>6</sup>, P. ROSSING<sup>7,8</sup>, L. M. RUILOPE<sup>9-11</sup>, A. E. FARJAT<sup>12</sup>, Y. M. K. FARAG<sup>13</sup>, P. SCHLOEMER<sup>14</sup>, R. LAWATSCHECK<sup>15</sup>, K. ROHWEDDER<sup>16</sup>, G. L. BAKRIS<sup>17</sup>

<sup>1</sup>Department of Internal Medicine, Max Rady College of Medicine, University of Manitoba, Winnipeg, Manitoba, Canada <sup>2</sup>Seven Oaks Hospital Chronic Disease Innovation Centre, Winnipeg, Manitoba, Canada <sup>3</sup>University of Manitoba, Community Health Sciences, Winnipeg, Canada <sup>4</sup>Department of Cardiology (CVK) of German Heart Center Charité; Institute of Health Center for Regenerative Therapies (BCRT), German Centre for Cardiovascular Research (DZHK) partner site Berlin, Charité Universitätsmedizin, Berlin, Germany <sup>5</sup>Institute of Heart Diseases, Wrocław Medical University, Wrocław, Poland <sup>6</sup>Department of Medicine, University of Michigan School of Medicine, Ann Arbor, MI, USA <sup>7</sup>Steno Diabetes Center Copenhagen, Gentofte, Denmark <sup>8</sup>Department of Clinical Medicine, University of Copenhagen, Copenhagen, Denmark <sup>9</sup>Cardiovascular Translational Laboratory and Hypertension Unit, Institute of Research imas12, Madrid, Spain <sup>10</sup>CIBER-CV, Hospital Universitario 12 de Octubre, Madrid, Spain <sup>11</sup>Faculty of Sport Sciences, European University of Madrid, Madrid, Spain <sup>12</sup>Research and Development, Clinical Data Sciences and Analytics, Bayer PLC, Reading, UK <sup>13</sup>US Medical Affairs, Bayer US LLC Pharmaceuticals, Whippany, NJ, USA <sup>14</sup>Statistics and Data Insights, Bayer AG, Berlin, Germany <sup>15</sup>Cardiology and Nephrology Clinical Development, Bayer AG, Berlin, Germany <sup>16</sup>Cardio-Renal Medical Affairs Department, Bayer AG, Berlin, Germany <sup>17</sup>Department of Medicine, University of Chicago Medicine, Chicago, IL, USA

## INTRODUCTION

- Chronic kidney disease (CKD) affects over 800 million individuals worldwide<sup>1</sup> and is often diagnosed at later disease stages when opportunities to prevent adverse outcomes are limited<sup>2</sup>
- Accurate prediction of individual risk of CKD progression could enhance patient experiences and outcomes
- Klinrisk, a lab-based risk prediction model, has been shown to accurately predict CKD progression in adults at all stages of CKD<sup>3</sup>

## AIM

- To validate the Klinrisk model for the prediction of key composite kidney outcomes up to 4 years post-randomisation in FIDELITY, a dataset combining individual patient-level data from two phase III, multicentre, double-blind trials investigating finerenone – a nonsteroidal mineralocorticoid receptor antagonist

## METHOD



### Population (validation cohort)

This post hoc analysis included all patients from FIDELITY, a prespecified pooled analysis of data from the FIDELIO-DKD (NCT02540993) and FIGARO-DKD (NCT02545049) trials<sup>4</sup>:

- Adults with CKD and type 2 diabetes (T2D) receiving a maximum tolerated dose of a renin-angiotensin system inhibitor
- Randomised 1:1 to finerenone or placebo
- Median follow-up: 3 years



### Key composite kidney outcomes

- Primary:** ≥40% estimated glomerular filtration rate (eGFR) decline or kidney failure
- Secondary:** ≥57% eGFR decline or kidney failure



### Variables used for model predictions

Klinrisk model risk predictions were based on single timepoint (baseline) measures of demographic data and routinely collected laboratory data, including:

- Age and sex
- Complete blood count
- A comprehensive metabolic panel
- Urine albumin-to-creatinine ratio (UACR)
- eGFR



### Statistical analyses

- Model **discrimination ability** and **calibration** were calculated using area under the curve (AUC) values, Brier scores and calibration plots in the overall population
- Sensitivity analyses** examined the accuracy of the models in predicting the secondary outcome and the change in risk score over time
- Kidney Disease: Improving Global Outcomes (KDIGO) heatmap categories were used as the **reference standard**

## RESULTS

### Baseline characteristics

- Selected baseline characteristics for the 13,026 patients included in the analysis are shown in **Table 1**

**Table 1.** Selected FIDELITY baseline characteristics

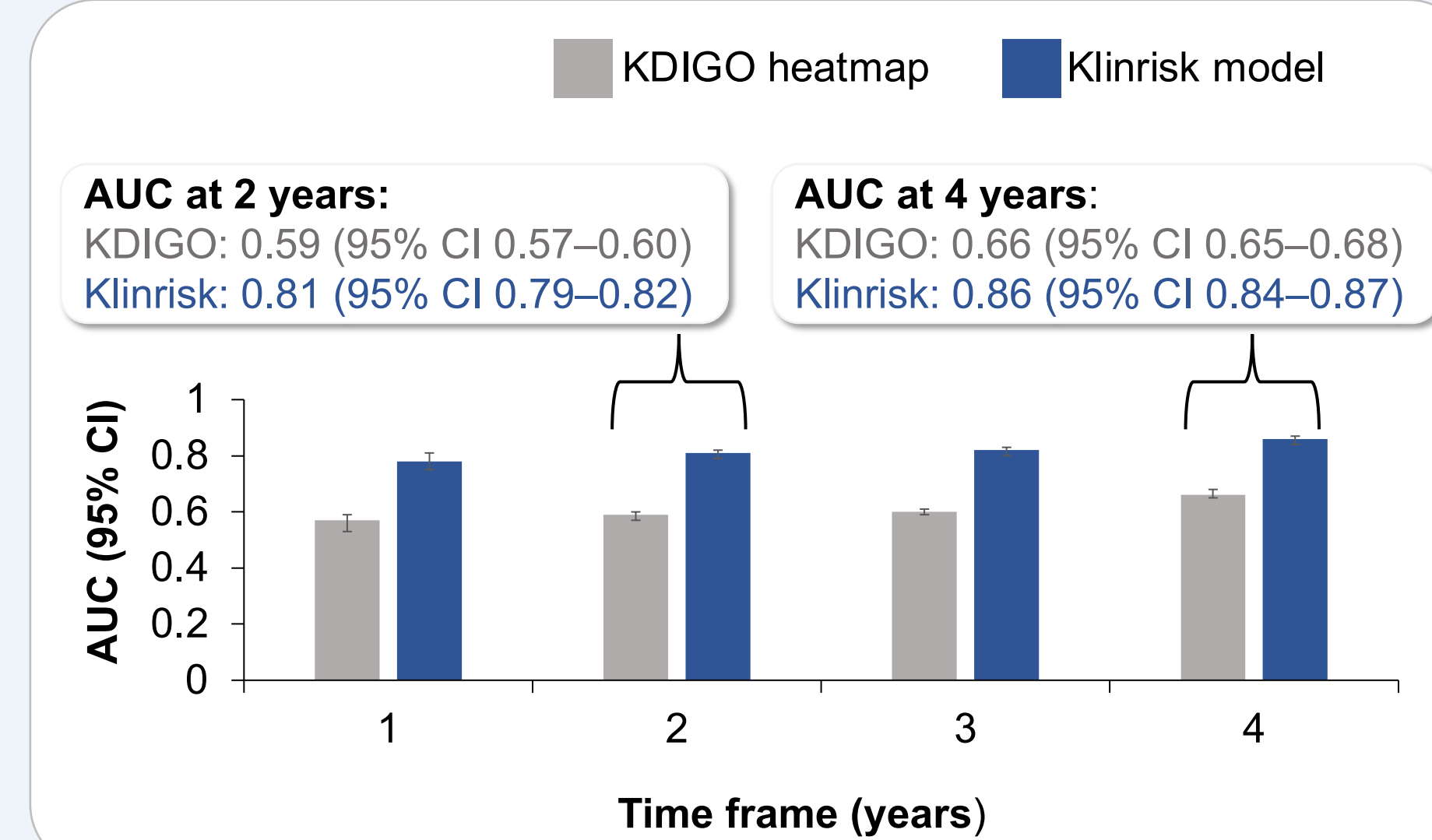
Characteristic, n (%) <sup>*</sup>	Total, N=13,026
Age, mean (SD)	64.8 (9.5)
Sex, male	9088 (69.8)
Sex, female	3938 (30.2)
Mean eGFR, ml/min/1.73 m <sup>2</sup> (SD)	57.6 (21.7)
Median UACR, mg/g (IQR)	515 (198–1147)

<sup>\*</sup>Data shown as n (%) unless otherwise specified  
eGFR, estimated glomerular filtration rate; IQR, interquartile range; SD, standard deviation; UACR, urine albumin-to-creatinine ratio

### Model performance and calibration

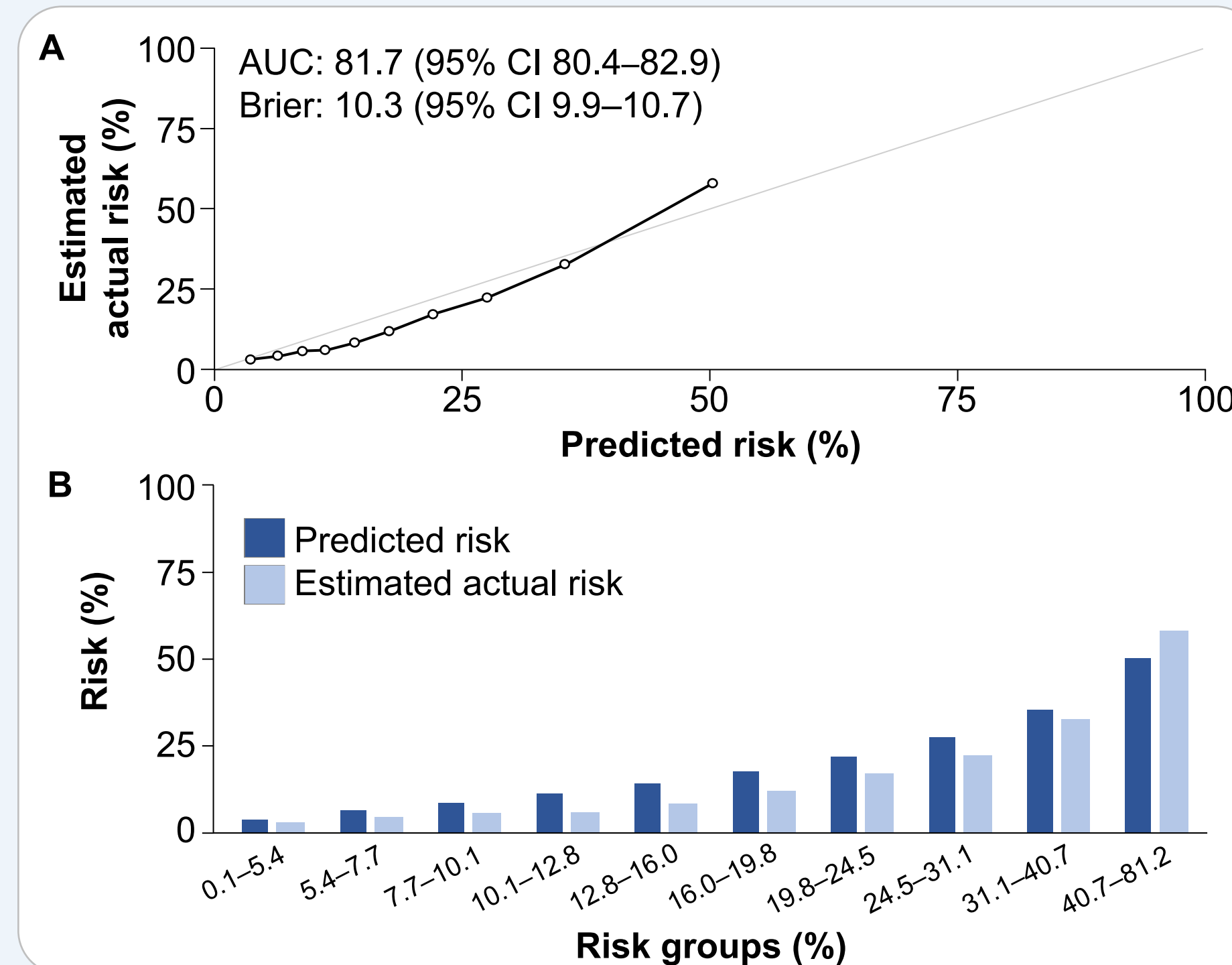
- At 2 and 4 years, 984 and 1795 patients had experienced a primary outcome event, respectively
- The Klinrisk model accurately predicted the primary outcome and outperformed KDIGO heatmap categories (**Figure 1**)
- Calibration was found to be appropriate:
  - Brier score 0.067 (95% confidence interval [CI] 0.064–0.070) at 2 years and 0.115 (95% CI 0.109–0.120) at 4 years
  - Calibration at 3 years is shown in **Figure 2**
- Discrimination accuracy for the secondary outcome was similar to that obtained for the primary outcome (C-statistic 0.88; 95% CI 0.87–0.90 at 3 years)

**Figure 1.** AUC scores for the primary outcome for years 1 to 4 with Klinrisk model and KDIGO heatmap categories



AUC, area under the curve; CI, confidence interval; KDIGO, Kidney Disease: Improving Global Outcomes

**Figure 2.** Calibration plots for Klinrisk prediction model for the primary composite outcome at 3 years



AUC, area under the curve; CI, confidence interval

## CONCLUSIONS

- Based on routinely collected lab data, the Klinrisk machine learning model accurately predicted CKD progression events in a well-characterised population pooled from two global clinical trials
- Prospective implementation of the model in clinical trial enrolment, as well as clinical care pathways, may allow for earlier intervention and improve clinical outcomes for patients with CKD

## ACKNOWLEDGEMENTS

Funded by Bayer AG. Medical writing assistance provided by Chameleon Communications International and funded by Bayer AG.

## REFERENCES

- Jager, KJ et al. A single number for advocacy and communication –worldwide more than 850 million individuals have kidney diseases. *Kidney Int* 2019; 96: 1048-1050
- Szczzech LA et al. Primary care detection of chronic kidney disease in adults with type-2 diabetes: the ADD-CKD Study (awareness, detection and drug therapy in type 2 diabetes and chronic kidney disease). *PLoS One* 2014; 9: e110535.
- Ferguson T et al. Development and external validation of a machine learning model for progression of CKD. *Kidney Int Rep* 2022; 7: 1772-1781
- Agarwal R et al. Cardiovascular and kidney outcomes with finerenone in patients with type 2 diabetes and chronic kidney disease: the FIDELITY pooled analysis. *Eur Heart J* 2022; 43: 474-484

## CONTACT INFORMATION

**Navdeep Tangri**  
Max Rady College of Medicine, Internal Medicine, Section of Nephrology, University of Manitoba, 2LB19 – 2300 McPhillips Street, Seven Oaks General Hospital, Winnipeg, Manitoba R2V 3M3, Canada  
Phone: +1 204 -631-3834; Fax: +1 204-632-3660; ntangri@sogh.mb.ca