

Predicting the risk of pancreatic cancer in individuals with newly- diagnosed type 2 diabetes

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Rationale for the study

- Improved outcomes could be possible with earlier detection, but there is no screening programme at present
- Identifying 'higher risk' individuals could enrich existing referral pathways, or inform the development of new pathways that seek to find pancreatic cancers earlier
- One approach could be to target adults with newly diagnosed type 2 diabetes (T2DM)
 - Some studies show that 1% of people with newly diagnosed T2DM are diagnosed with pancreatic cancer within the next 2 years
 - Approximately a quarter of people with pancreatic cancer have a history of T2DM
 - T2DM may in some cases be 'Type 3c' diabetes
- NICE guidelines suggest that adults age 60+ years with newly diagnosed T2DM and weight loss be referred for 'fast track' imaging to assess for a possible tumour in the pancreas
- **Could more nuanced ways to assess risk be better at finding people for referral in primary care?**

Outline of the study approach

- Exploring different approaches that may be able to accurately estimate the risk of an individual with new-onset type 2 diabetes developing pancreatic cancer (PDAC and PNET)
- Statistical and 'machine learning' approaches – which is the most useful?
 - Cox model, XGBoost, neural networks – tuning with Bayesian optimisation
- **QResearch database** – primary care data with individual level linkage to hospital data, cancer registry and ONS in England
- Identified: adults aged 30-85 years at time of T2DM diagnosis (2010-2021)
 - 253,766 individuals with T2DM in the final study database
 - 767 of these were diagnosed with a pancreatic cancer within the next 2 years
- Exploration of different potential predictors – e.g. age, sex, HbA1c, body mass index, creatinine, platelet count, alcohol intake, previous venous thromboembolism, and symptoms such as abdominal pain, weight loss, indigestion (within the previous 6 months)

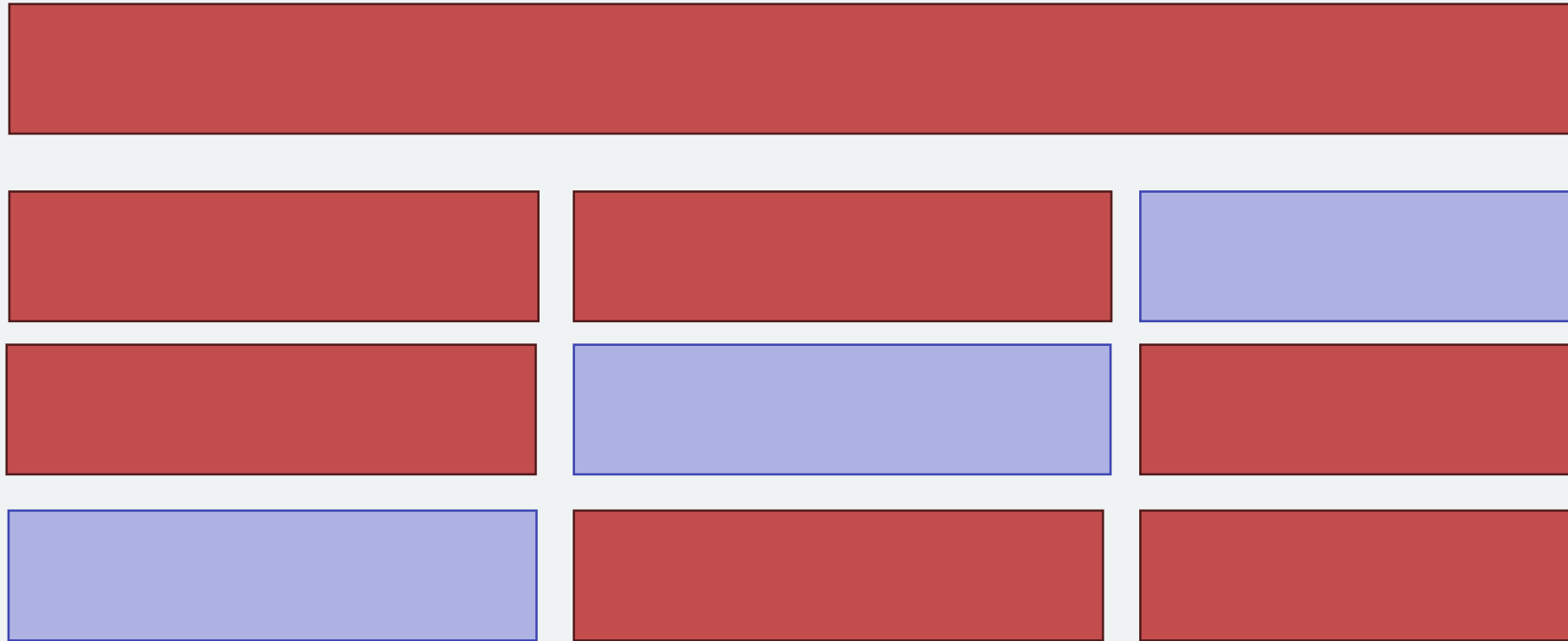
Assessing how well models perform

- Discrimination
 - Does the model distinguish between those that did and did not get a pancreatic cancer diagnosis?
 - Harrell's C – goes between 0.5 and 1
 - 0.5 means model is no better than a coin toss, 1 means 'perfect'
- Calibration
 - Do the probabilities produced by the model align with the observed risk?
- Clinical utility (net benefit)
 - Is the model associated with better clinical decision making?
 - Typically, compare against 'test everyone', 'test nobody', or other models
- Using the geographical coverage of QResearch and the linked datasets
 - Estimate key metrics in each held out region, then pool together with meta-analysis
 - Provides an estimate, and an indication of the expected range of performance if applied to a new population

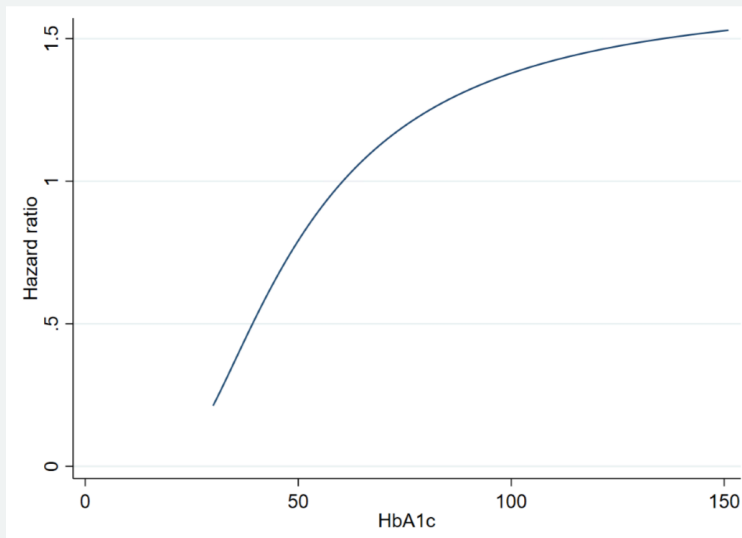
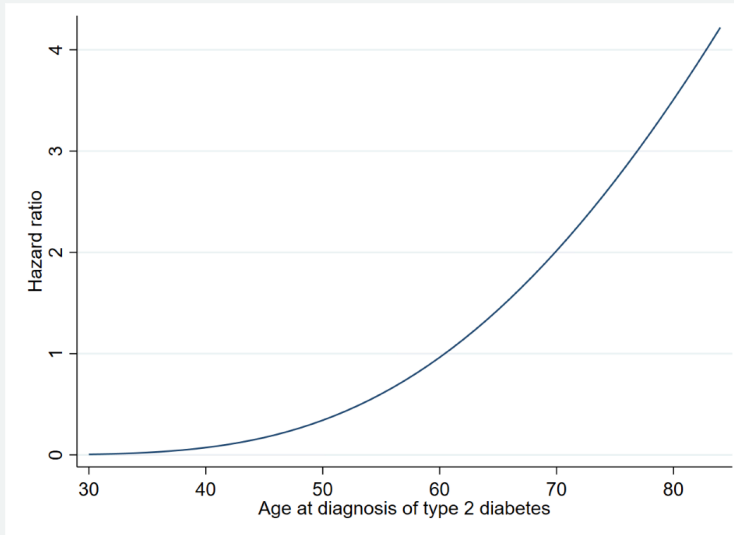
'Train test split'



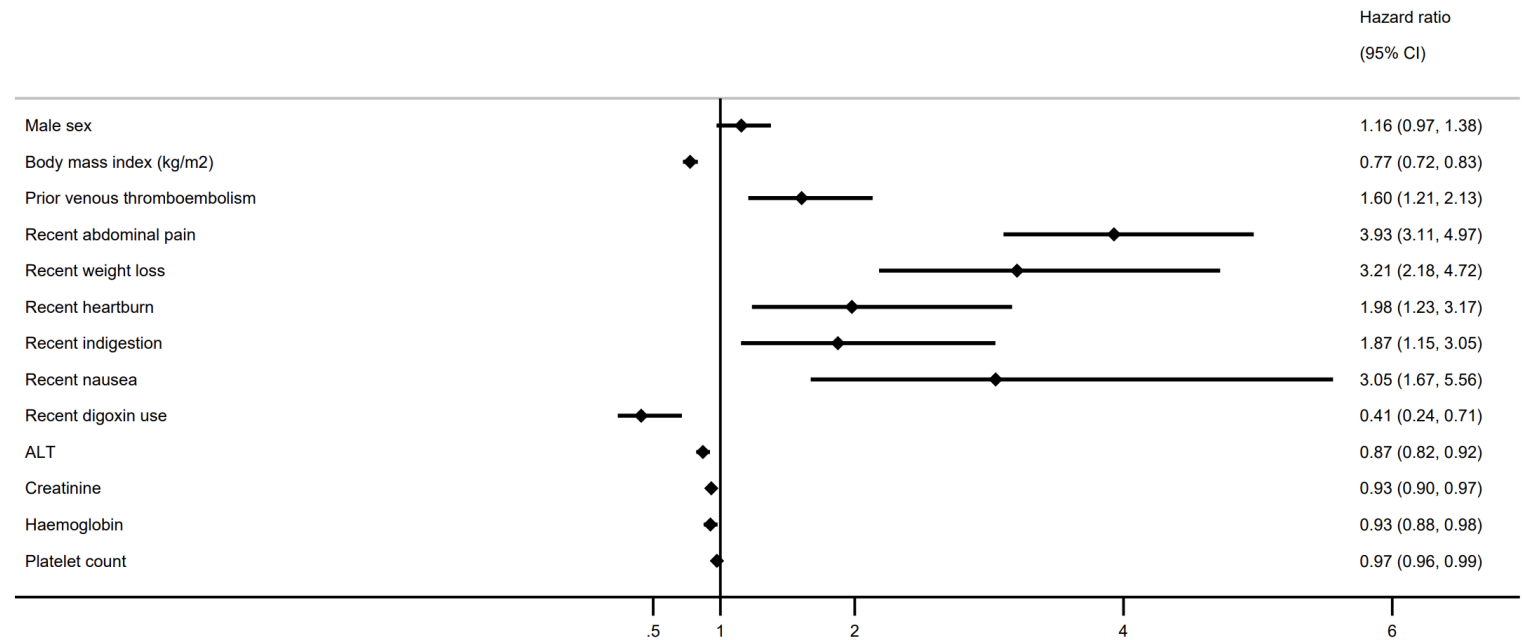
'Cross-validation'



Key results – Cox model



Cox proportional hazards model
Includes FP terms for age (-1), HbA1c (-2) (not shown)



Metric**Estimate
(95% confidence interval)
[95% prediction interval]**

Cox model

XGBoost

Neural Network

Harrell's C-index

0.802

0.723

0.650

(0.787 to 0.817)

(0.689 to 0.756)

(0.516 to 0.784)

[0.766 to 0.839]

[0.628 to 0.817]

[0.202 to 1.000]

Calibration slope

0.980

1.180

1.855

(0.897 to 1.062)

(1.056 to 1.305)

(-0.945 to 4.654)

[0.778 to 1.182]

[0.781 to 1.580]

[-7.552 to 11.261]

Calibration-in-the-large

-0.020

0.180

0.855

(-0.103 to 0.062)

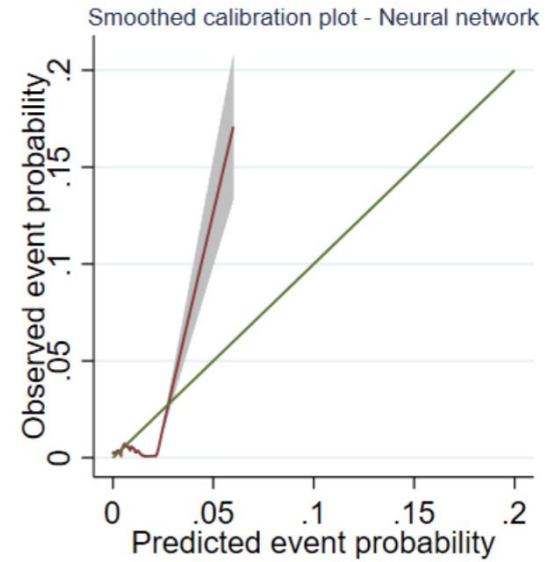
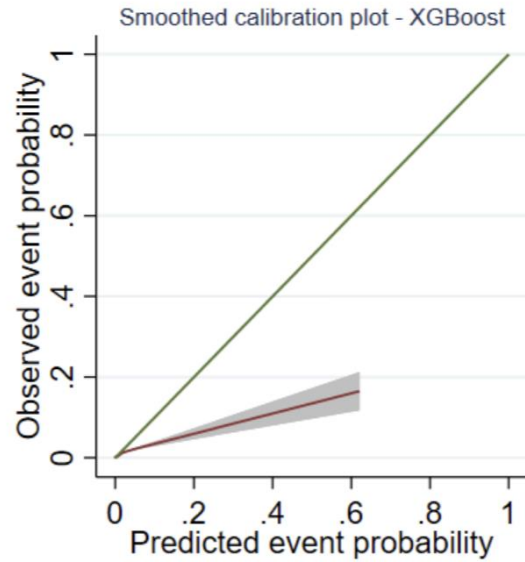
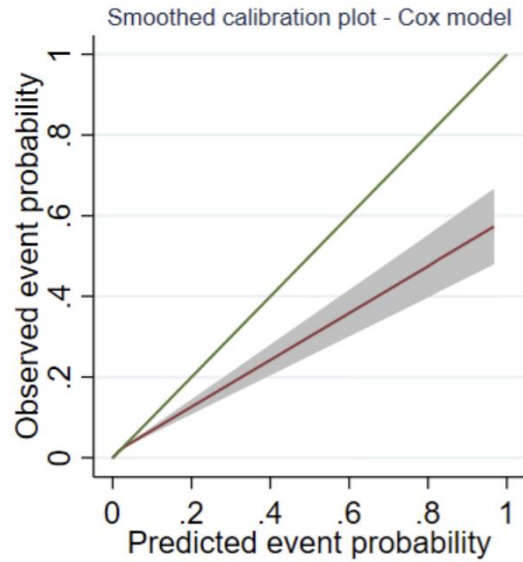
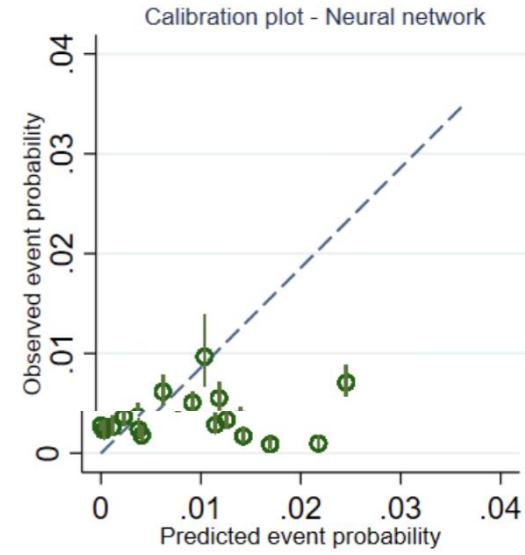
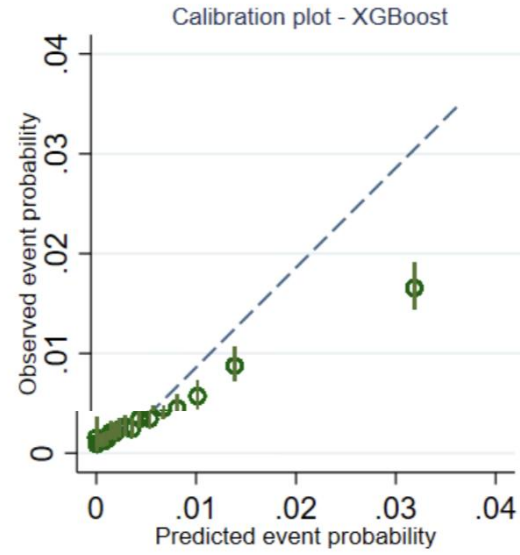
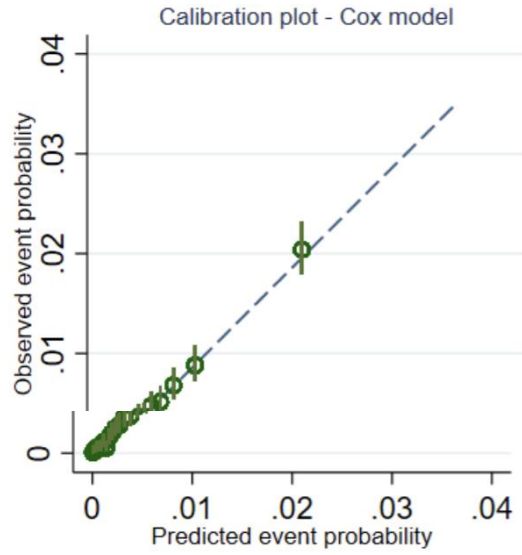
(0.056 to 0.305)

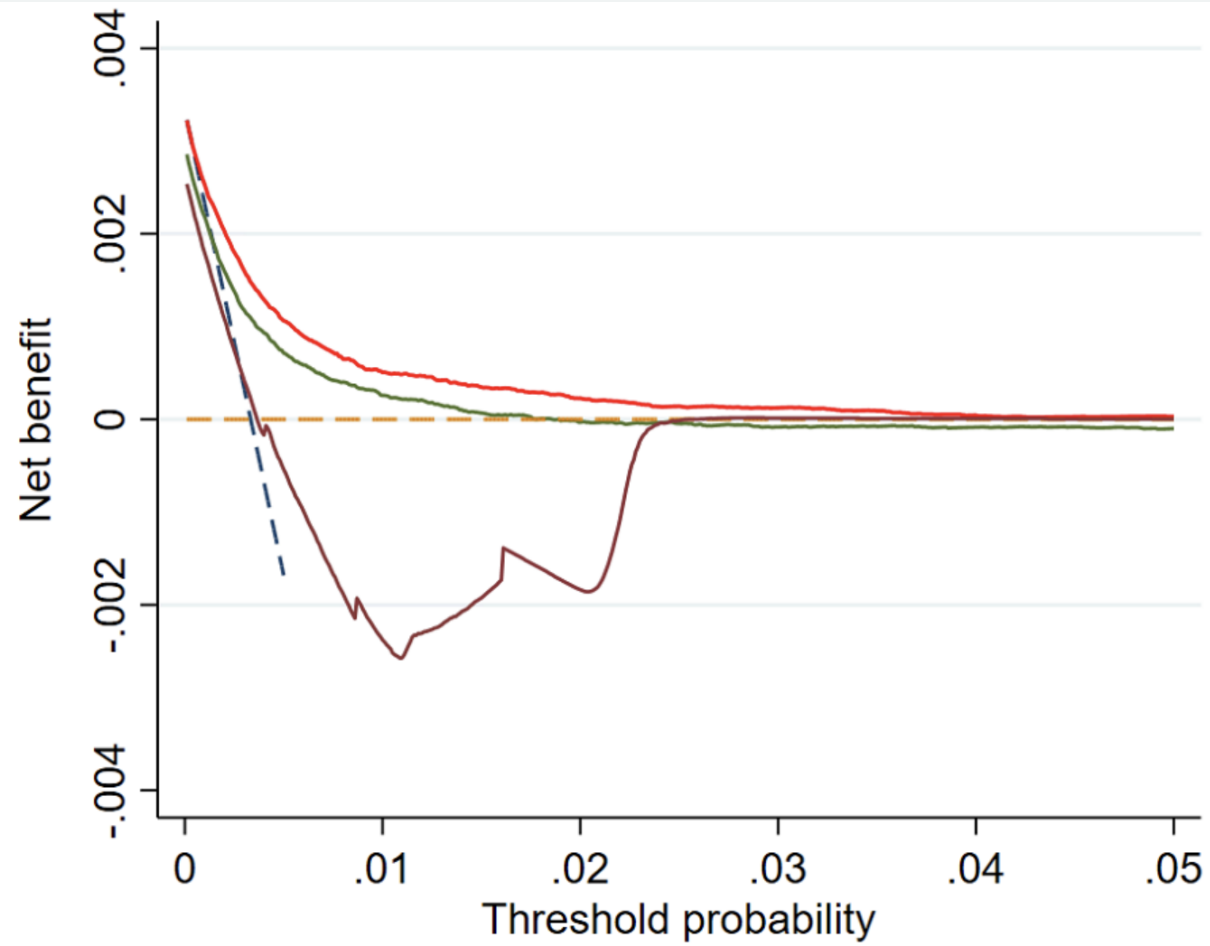
(-1.945 to 3.654)

[-0.222 to 0.182]

[-0.219 to 0.580]

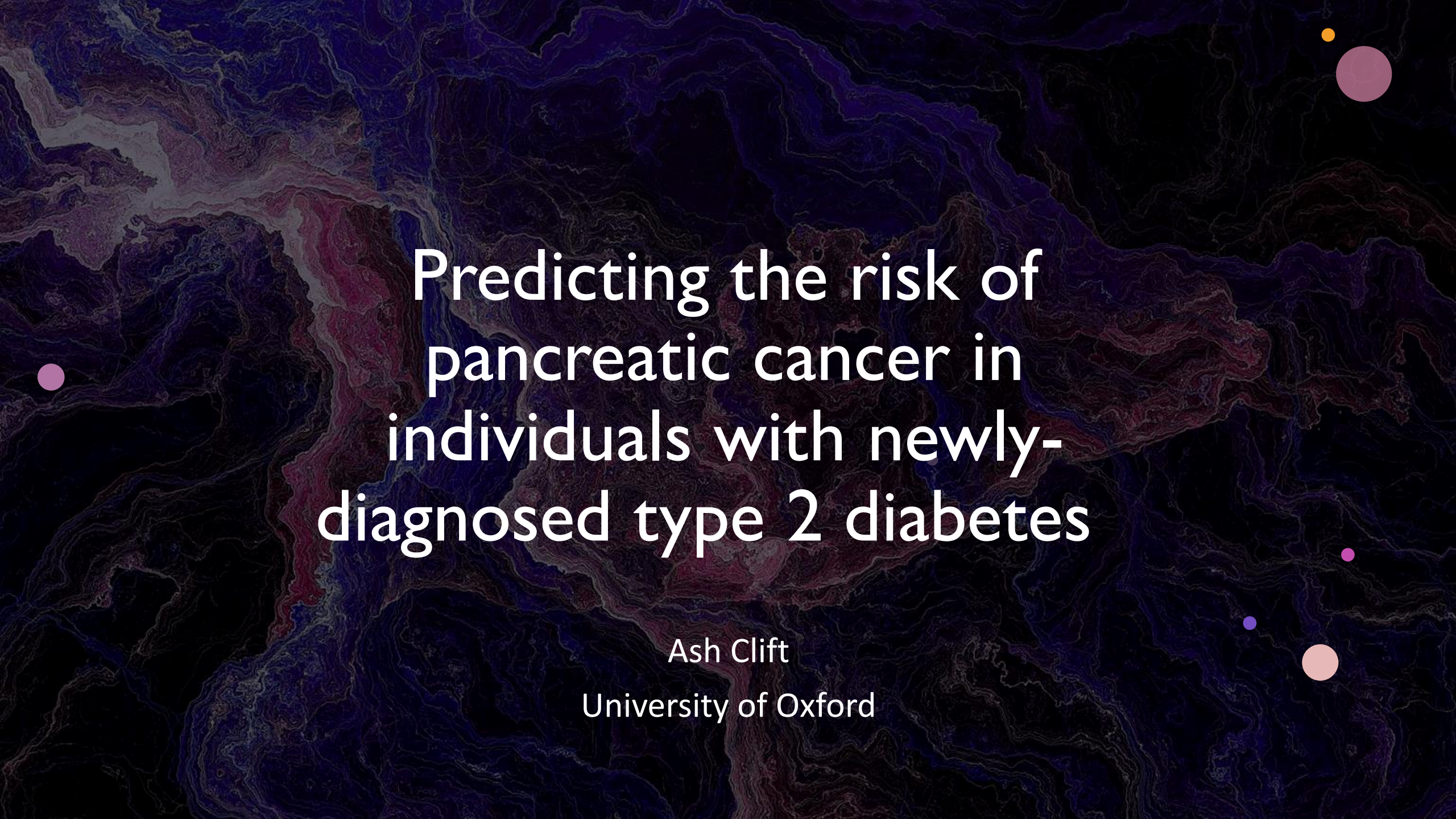
[-8.552 to 10.261]





What does it all mean?

- In this study, the sensitivity of the current NICE guidelines was 3.53% overall, and 3.95% in the over 60s
 - The sensitivity for the highest 1% group as assessed by the new model = 12.51%
- The Cox model -> discriminates well, is well calibrated, and associated with better decision making
- We developed a model, understand how well it works, but next we need:
 - External evaluation, including a comparison with other tools
 - Health economic simulations to understand the clinical and cost-effectiveness of new strategies that could be informed by the model's outputs
 - What is the best way to use it?
- No tool is perfect, and no tool in this setting can probably 'find every case', but *progress*



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