## ML model for Shared Decision-Making Tool for CRC screening

Daiane M. Seibert and Karen Feyen

## What will Orient try to do?

Provide the general practitioners (GPs)
with the tools to discuss
colorectal cancer screening and its benefits and harms with patients belonging to vulnerable groups, so that they are well-informed and can take a deliberate decision to participate or not in screening.

## Key messages about SDM tools

- SDM tools for cancer screening are effective in increasing knowledge on cancer and cancer screening of vulnerable people.
- SDM tools are effective in reducing decision conflicts among people considering cancer screening.
- SDM tools are effective in increasing screening intentions among vulnerable people, especially those who are considering colorectal cancer screening.
- Given the complexities of patients' and clinicians' preferences in SDM tool characteristics, fostering collaboration between patients and clinicians during the creation of an SDM tool for cancer screening is essential.


## Key messages about SDM tools

- Out of numerous risk prediction models for CRC with good model performance, only a few of them can be potentially integrated into practical healthcare settings.
- It is crucial to establish a standardized reporting of risk prediction models in CRC that accounts for the model's interpretability, generalizability, and potential clinical utility.


## The constraints

- The tool need to be simple in a way to be used in a $15-\mathrm{min}$ appointment.
- Features based in laboratory tests do not fit our objectives. (Conventional models)
- Information needed to feed into the model will be collected by the doctor.
- Things to keep in mind:
- The language barrier over the vulnerable population.
- Lack of knowledge about CRC screening.

Study ID
Sample size
Asia-Pacific Colorectal Screening risk scores

| Yeoh 2011 | 1,892 C | Asia | -- |
| :---: | :---: | :---: | :---: |
| He 2019 | $1,201{ }^{\text {B }}$ | China | $\longmapsto-1$ |
| He 2019 (modified) | 1, 201 ${ }^{\text {D }}$ | China | - |
| Sung 2018 | 1,915 ${ }^{\text {D }}$ | Hong Kong | $\longmapsto-1$ |
| Sekiguchi 2018 | $5,218{ }^{\text {B }}$ | Japan | +o-1 |
| Kaminski's risk score |  |  |  |
| Kaminski 2014 | 17,939 A,C | Poland | -or |
| Ruco 2015 | 5,137 ${ }^{\text {B }}$ | Canada | Ho-1 |
| Other de novo models |  |  |  |
| Cai 2011 | 7,541 ${ }^{\text {B }}$ | China | $0-1$ |
| Chen 2014 | 905 A | China | - 0 |
| Hong 2017 | 48, $950{ }^{\text {A }}$ | Korea | -0-1 |
| Jung 2017 | 96, $235{ }^{\text {A }}$ | Korea | -0- |
| Kim 2015 | 3,561 ${ }^{\text {A }}$ | Korea | $\longmapsto \sim$ |
| Kim 2019 | 17, 873 A | Korea | $\mapsto$ |
| Ma 2010 | 18, $256{ }^{\text {A }}$ | Japan | 1 |
| Musselwhite 2019 | $3,121^{\text {B }}$ | U.S. | -O- |
| Schroy III 2015 | 3,543 A | U.S. | -0-1 |
| Sekiguchi 2018 | 5,218 A | Japan | -0-1 |
| Sharara 2020 | $980{ }^{\text {A }}$ | U.S. | $\stackrel{-1}{ }$ |
| Shin 2014 (male) | 547, $874{ }^{\text {B }}$ | Korea | Юr |
| Shin 2014 (female) | 415, $875{ }^{\text {B }}$ | Korea | Or |
| Sutherland 2021 | 3, $035{ }^{\text {A }}$ | Canada | $\longmapsto-1$ |
| Tao 2014 | 7, 891 ${ }^{\text {A }}$ | Germany | $\mapsto-1$ |
|  |  | $\Gamma$ | 1 1 |
|  |  | 0.5 | $\begin{array}{lll}0.6 & 0.7 & 0.8\end{array}$ |

AUC 95\% CI

|  | A Internally validated model <br> $0.64(0.57-0.71)$ |
| :--- | :--- |
| $0.67(0.57-0.71)$ | B External validation |
| $0.69(0.65-0.73)$ | C Original model |
| $0.65(0.61-0.69)$ |  |
|  |  |

0.62 (0.60-0.64) 0.64 (0.61-0.67)
0.74 (0.70-0.78) 0.75 (0.70-0.82) 0.71 (0.69-0.74) 0.67 (0.65-0.70) 0.68 (0.61-0.76) 0.66 (0.65-0.67) 0.70 (0.68-0.72) 0.60 (0.57-0.63) 0.69 (0.66-0.72) 0.70 (0.67-0.73) 0.73 (0.66-0.79) 0.76 (0.75-0.77) 0.71 (0.70-0.72) 0.69 (0.65-0.72) 0.71 (0.67-0.75)

| A Internally validated model |
| :--- |
| ${ }^{\text {B }}$ External validation |
| c Original model |
| D Updated model |

ORIZNT

Model performance of conventional risk models for colorectal cancer with 95\% confidence interval (CI) Info collected during the literature review performed in the project.

## Al for ORIENT



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Inference

## Al for ORIENT - Inference example

## Patient X :

- 65 years old
- Former smoker
- Diabetes
- No Hypertension



## Data: PLCO - The Prostate, Lung, Colorectal and Ovarian Cancer

PLCO study aimed to evaluate the impact of screening exams on reducing mortality from prostate, lung, colorectal, and ovarian cancers. Participants assigned to the control arm received usual care, whereas participants assigned to the intervention arm were invited to receive screening exams.

Enrolment period: 1993-2001
Data size: ~154,000
Screening period: 1993-2009
Data Collection: +500 risk factors, collected through 6 questionnaires


## Al model: feature engineering

US dataset


- $\pm 150000$ patients
- $500+$ risk factors

First risk factor selection.
E.g. factors removed:

- Screening \& diagnosis features
- US specific
- Too specific

Second risk factor \& patient selection.
E.g. removed:

- Patients with too many missing values
- Factors statistically too insignificant

US dataset


- $\pm 116000$ patients
- 98 risk factors
- 1 label

US dataset


- $\pm 150000$ patients
- 248 risk factors
- 1 label

- Balance $\pm 1: 70$


## Al model: feature engineering

US dataset


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Third risk factor
selection:
E.g. factors kept:

- Already known in literature
E.g. factors removed:
- Low importance in the model


## US dataset



- $\pm 116000$ patients
- 20 risk factors
- 1 label

US dataset

E.g. factors removed:

- Too many assumptions needed to be calculated.
- $\pm 116000$ patients
- 12 risk factors
- Too difficult to be collected by
- 1 label GP (vulnerable population)


## Selected risk factors (Features)

1. Sex: Male, Female
2. Age: Numeric
3. Height: Numeric
4. Weight: Numeric
5. BMI: Numeric
6. Hypertension: Did the participant ever have high blood pressure? Yes/No Question modified: Do you take medication for hypertension? Yes/No
7. Heart problems: Did the participant ever have coronary heart disease or a heart attack? Yes/No Question modified: Do you take medication for heart problems? Yes/No
8. Diabetes: Did the participant ever had diabetes? Yes/No
9. Family History of Colorectal Cancer: Colorectal cancer family history in first-degree relatives. Includes parents full-siblings and children. Yes/No/Possible, cancer type not clear
10. Smoke history: Participant's current cigarette smoking status. Former smoker/ Current smoker / Never smoked
11. Smoke quantity: In the time that you smoked, how many cigarettes per day (approximately)? 0 / 1-10/11-20/21-30 / More than 30
12. Alcohol drink history: Alcoholic Beverages (drinks/day) - DHQ. Does not drink / Drinks, but less than 7 drinks per week / Drinks, more than 7 drinks per week
Alcohol from Beer Wine and Liquor - Age 40-54 (drinks/day) - DHQ. Did not drink / Drank, but less than 7 drinks per week / Drank, more than 7 drinks per week

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\#9
This factor is an exception factor in our tool.

Following the local guidelines, who has family history of colorectal cancer, should be advised to go through colonoscopy directly.
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## \#6 and \#7

Heart and Hypertension factor in our model decreased the risk. We discovered that previous studies indicated the use of standard medication for these conditions also reduced the risk. As a result, we decided to modify our research question.
7. Heart problems: Did the participant ever have coronary heart disease or a heart attack? Yes/No Question modified: Do you take medication for heart problems? Yes/No
8. Diabetes: Did the participant ever had diabetes? Yes/No
9. Family History of Colorectal Cancer: Colorectal cancer family history in first-degree relatives. Includes parents full-siblings and children. Yes/No/Possible, cancer type not clear
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Alcohol from Beer Wine and Liquor - at your forties (drinks/day) - DHQ. Did not drink / Drank, but less than 7 drinks per week / Drank, more than 7 drinks per week

## Data change

## 1. Alcohol drink history:

- Alcoholic Beverages (drinks/day) - DHQ. Does not drink / Drinks, but less than 7 drinks per week / Drinks, more than 7 drinks per week
- Alcohol from Beer Wine and Liquor - At your forties (drinks/day) - DHQ. Did not drink / Drank, but less than 7 drinks per week / Drank, more than 7 drinks per week

A regular beer is considered as one drink.

The threshold of 7 biers was defined based in the local dietary guidelines and, also the proportion of cancer incidence in our data.
2. Age: The age was collected when all patients were enrolled.

- We have the age that the cancer was discovered, for positive cases.
- And what about negative cases?
- Age that colonoscopy was performed (just for intervention arm)
- Ages based in the colonoscopy date was added for negative cases.

Cases where a person died from colorectal cancer but was negative, were excluded.

Model

## US dataset

- $\pm 116000$ patients
- 12 risk factors
- 1 label


## Explainable AI

remainder_former_smoker


Output $=0.4$


Base rate $=0.1$
to_norm__age_cancer
to_norm__bmi_curr
to_norm__weight_f to_norm__cigpd_f to_norm__height_f
remainder_hyperten_f
to_norm__dhq_alc_drinks_day
remainder__curr_never_smoke
remainder sex
remainder__heart_f
dhq_mped_alc_bev_alc_drinks40
remainder__diabetes_f

## Measures of prediction model performance

| Terms | Definition |
| :--- | :--- |
| AUC | Area under the curve, in this case the receiver operating characteristic curve. A measure <br> of discrimination. For prediction models based on logistic regression; this corresponds to <br> the probability that a randomly selected diseased patient had a higher risk prediction <br> than a randomly selected patient who does not have the disease |
| Calibration | Correspondence between predicted and observed risks is usually assessed in calibration <br> plots or by calibration intercepts and slopes. |
| Sensitivity | The proportion of true positives in truly diseased patients |
| Specificity | The proportion of true negatives in truly non-diseased patients. |
| Positive predictive value | The proportion of true negatives in patients classified as negative. |
| Negative predictive value |  |

Info from the literature reviews performed in the project.

## Model's performance

Receiver Operating Characteristic
AUROC of our model is 0.716 .
It is a metric used to evaluate the performance of binary classification models.

The Receiver Operating Characteristic (ROC) curve is a graphical representation of the true positive rate (sensitivity) plotted against the false positive rate (1specificity) as the discrimination threshold of the model varies.
The area under this ROC curve, the AUROC, is a single scalar value that summarizes the overall performance of the model across various threshold settings.


Model's performance


How to present the risk


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## Thank you!

Questions?

You can contact us by e-mail: daiane.seibert@thomasmore.be karen.feyen@thomasmore.be

## THOMAS

