

Lung cancer screening and tobacco control

Natalie Lui, MD Stanford University School of Medicine Stanford Cancer Center Stanford, CA USA

Natalie Lui, Stanford University School of Medicine, United States

DISCLOSURES

Lung cancer risk assessment

China Lung Cancer Screening (CLUS) version 1.0

- From November 2013 to November 2014
- 6717 eligible participants with high-risk factors
- LDCT vs Nature cases
- Screening interval: biennial
- Screening rounds: three
- Data cut-off date: February 28, 2022

20

60

40

OS(months)

80

Clinical characteristics, stage and histologic features of lung cancers diagnosed

100

Proportion of deaths in lung cancers diagnosed

HUNT Lung-SNP model

Model development

Cohort: HUNT2

- $N = 30$ 746 ever smokers (median follow up 15,26 years).
- $N = 160$ individuals were diagnosed with LC within 6 years

External validation:

- Cohort: Tromsø study
- $N = 3074$ ever smokers (median time to event 3.04 years)
- $N = 39$ individuals were diagnosed with LC within 6 years

Results

MA11.04

The INTEGRAL project

1000+ proteins Proximity extension assay (Olink) 6 cohort studies from 12 countries

731 cases and 731 matched controls with a history of daily smoking

Pre-diagnostic blood collected up to 3y before diagnosis

The Lung Cancer Cohort Consortium (LC3) The Lung Cancer Cohort Consortium, *medRxiv* ²⁰²²

Robbins et al, *medRxiv* 2022

Proteins associated with lung cancer risk

67 proteins Corrected for multiple comparisons 36 'robust' proteins Resampling algorithm

The Lung Cancer Cohort Consortium, *medRxiv* 2022 Robbins et al, *medRxiv* 2022

Nodule detection and reporting

MA11.07

Early Lung Imaging Confederation

- **Participating Sites (Spokes) Provide 100 cases of de-identified high quality screening CT scan images and metadata at 2 time points to IASLC ELIC hub for analysis within a highly secure and strictly controlled environment**
- **All Spoke Provided Data Stays Within Their Country/Region By Using The Amazon Web Services (AWS) Global Cloud Infrastructure**
- **AI Algorithm Developers Can Securely Send a Lung Analysis Algorithm To The Spokes To Run Computational Experiments And Receive Back Analysis Results.**
- **Selected data set for an initial feasibility study**

ELIC Is Now Running Globally Distributed Deep Learning AI and Quantitative Imaging Experiments These First Analyses Show The Potential Of ELIC

400

AUGUST 6-9, 2022 | VIENNA, AUSTRIA

MA11.07

Mean Volume Change from Two CTs

Semi-automated Volumetric Measurements of Change in Solid & Part-solid Nodules in 2 CTs from Same Individual

COV = Coefficient of Variation

STUDY DESIGN

RESULTS – 1) Radiologist's Reading Time

Methods:

Retrospective study based on the reanalysis of LDCT performed in the first lung cancer screening program in Brazil (BRELT1).

LDCT were evaluated by radiologist and analyzed using artificial intelligence software (BOTKIN IA – Russia)

In each exam, LungRADSTM was evaluated.

General methodology primarily focuses on outcomes-based training, full volume approaches, and directly comparable clinical performance evaluation.

ASSESSMENT OF THE MAIN NODULE - LUNG RADSTM

147

272

■ Radiologists ■ Botkin AI Software

LungRADS[™] 3 and 4

Sensibility of 92.5%

Specificity of 78.5%

PPV 50% NPV 97.8%

Overall accuracy of 81.1%

507

632

Cost effectiveness

Methods

We assumed that screening reduces lung cancer mortality (per NELSON and NLST) and calculated the costs and QALYs implied by that assumption.

Modelling procedure

- 1. Identify the eligible population in Australia and model death from lung cancer or other-causes.
- 2. Apply the lung cancer mortality benefit observed in trials to estimate life years gained.
- 3. Estimate lung cancer cases by stage, with and without screening (accounting for overdiagnosis).
- 4. Apply costs and disutilities relating to screening, false positives, and treatment.
- 5. Estimate incremental costs per QALY.

Inputs

Australian data

- Composition of eligible population estimated from the *45 and Up Study*, a longitudinal cohort study (n = 267,153).
- Hazard ratios for LC and all-cause mortality in the *45 and Up Study* using linked records on cancer diagnoses and deaths.
- Lung cancer $costs¹$ and cost of CT scan (\$307).
- $SF-6D$ utility values²⁻⁴

Trial-related outcomes

- LC mortality reductions (by length of follow-up)
- Stage-shift
- False positive rates
- Overdiagnosis rates
- 1. Goldsbury et al. Health services costs for lung cancer care in Australia: Estimates from the 45 and Up Study. PLOS ONE. 2020 Aug;15(8):e0238018.
- 2. Ngo et al. Health utilities for participants in a population-based sample who meet eligibility criteria for lung cancer screening. Lung Cancer. 2022 May 13.
- 3. Ngo et al. Large-Scale Population-Based Surveys Linked to Administrative Health Databases as a Source of Data on Health Utilities in Australia. Value in Health. 2022 May 6.
- 4. Tramontano et al. Catalog and Comparison of Societal Preferences (Utilities) for Lung Cancer Health States. Medical Decision Making. 2015 Apr;35(3):371–87.

Results

NELSON

Base case: AU**\$39,250**/QALY 95% CI: AU**\$18,150-108,300**/QALY P(ICER < AU\$30,000/QALY) = **15%** P(ICER <AU\$50,000/QALY) = **60%**

NLST

Base case: AU**\$76,300**/QALY 95% CI: AU**\$41,750–236,500**/QALY P(ICER < AU\$30,000/QALY) = **0.5%** P(ICER < AU\$50,000/QALY)= **6.7%**

Figure 1. (A) Scatter plot of incremental costs (in AU\$/person) vs incremental QALYs/person obtained from the PSA for the NELSON and NLST settings. (B) Corresponding estimated cost-effectiveness curve given the ICER distributions obtained from the PSA. (C and D) Histograms showing the ICER distributions obtained from the PSA for the NLST and NELSON settings, respectively. MA11.03

Surveillance

@alexandra_p_24 @PriyankaSenth16 @chifujeffyang

Incidence of Metachronous Primary Lung Cancer from the Date of Initial Primary Lung Cancer Diagnosis

Smoking cessation

PL03.03

cigarettes dispensed from the co-located service

Opt-In vs Opt-Out Tobacco Treatment in Hospital

Changing the Default (N=1,000)

- Randomized clinical trial
- Primary outcome: Verified quit
	- 1 month post dx

Richter & Ellerbeck, 2015; Faseru et al., 2017; Faseru et al. 2022

Babalola Faseru, MD, MPH, University of Kansas Cancer Center, USA NGI Cancer Center

Medication & Counseling Use (N=739)

Smoking Cessation Programs

Results

Smoking cessation program participation:

4,451 had baseline LDCT scan 3,063 (68.8%) current smokers 2,736 (89.3%) attended counselling on day of LDCT screening

Program results:

1,689 had a 12-month follow-up LDCT with complete data Quit rate (30-day abstinence): 15.5% (range 10.5%–20.0%) Relapse rate 6.3%: (3.1%–7.3%)

Trial design is feasible 31/36 (86.1%) Enrolled 25/36 (69.4%) First clinic visit 6/36 (16.7%) saw smoking cessation, received \$50 and quit

12/36 (33.3%) quit rate amongst both groups

- Reward Group: 5/15 (33.3%)
- Banked Money Group: 7/16 (43.8%)
-

Take home points

- Lung cancer screening implementation is challenging globally – every step needs analysis and optimization, with goals of equitable care and using technology
- Smoking cessation programs require creativity more intensive programs, opt out instead of opt in, integrating with screening, financial incentives