

Investigating the feasibility of using A.I. for population-level mammography image quality improvement initiatives at Leeds Teaching Hospitals NHS Trust

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Mammography Image Quality (IQ)

- The 3 core PHE standards in screening:
 1. Achieve **optimum image quality**
 2. Limit radiation dose
 3. Minimise the number of repeat examinations

^Co-dependent: ↑ image quality = ↓ radiation dose^[O'Leary, 2011] & ↓ technical recall^[Salkowski, 2019]

- Higher quality images ...
 - ... ↑ sensitivity^[Taplin, 2002]
 - ... ↓ stage at detection^[Rauscher 2013]
 - ... ↓ interval cancers^[Taplin, 2002]
 - ... ↓ false positive rate^[Guertin 2018]
- The \$\$\$ of mammography IQ
 - Annual direct costs:
 - Technical recalls (2.13%^[2018-2019]) = **~£1.8 million**
 - QA self-reviews (1 shift/month, 40 studies) = **~£2.5 million**
 - Delayed diagnosis and treatment costs
 - Breast cancer stage = strongest predictor of costs.^[Hall 2015]
 - **True annual £ unknown**

Current State of IQ @ LTHT

- IQ monitoring
 - Technical recall/repeat aggregate rates monitored monthly
- Self-evaluations
 - Screening service: 40 mammograms self-reviewed monthly
 - Diagnostic service: no mandated reviews
- IQ improvement initiatives
 - No active IQ educational interventions
 - Until now, population-level IQ data inaccessible
 - Prospective trial planning

Alignment with NHS mandates

- NHS Long Term Plan for Cancer
 - Calls for improvement in national screening programmes through investment in innovative technologies;
- NHSX Strategic priorities
 - Calls for introducing technologies reducing burden on clinicians and staff, to focus on patients

IQ Assessment Challenges

- Time-consuming
 - Infeasible at time of image acquisition (8 min/exam) ← Resource challenges further intensified by mammography workforce pressures
 - Delayed feedback/corrective action

- Visual

- Subjective

UK NHS positioning evaluations

- Nipple in profile ← Some apply literally, others say yes, only when obstructs breast tissue^[Boyce 2015]
- MLO Pectoral muscle to nipple level
- MLO Pectoral muscle at appropriate angle
- Symmetrical images
- MLO IMF shown clearly
- CC Medial border demonstrated
- CC Some axillary tail shown
- CC Back of breast clearly shown with some medial central & lateral

Ambiguous terms



Does the Pectoralis muscle extend within 1cm of the PNL?



1st read: 9/15 reviewers = YES
2nd read: 9/15 reviewers = YES, BUT...

6/15 (40%) reviewers flipped their assessment (3 flipped to present, 3 flipped to absent) ^[Sharma 2020]

Research Study Objectives

Primary: To investigate the current state of mammo IQ in breast imaging services at Leeds Teaching Hospitals NHS Trust (n~60 000 images)

Secondary: To compare the population-based AI prevalence rate with visual prevalence rates in a validation sample (n~200 images).

- *The study was waived by research ethics*
- *The study was approved by institutional quality committee*

Methodology

- Population-based image processing
 - Densitas® IntelliMammo™ was installed
 - Studies acquired over 12/2021 to 03/2022 were processed
 - [N=59 264 images ($n_{CC}=29964$ $n_{MLO}=29300$)]
- Manual data collection (random sample of 50 symptomatic studies)
 - A pair of lead radiographers reviewed together for a consensus
 - 196 images (98 CC & 98 MLO)
- Analysis
 - Event rate per positioning error (stratified by CC/MLO)
 - Weekly average error rate time plots
 - Agreement assessed by Cohen's kappa (validation dataset [n=198])

***Kappa Classifications:**

Less than chance agreement (<0);

Slight agreement (0.01-0.20);

Fair agreement (0.21-0.40);

Moderate agreement (0.41-0.60);

Substantial agreement (0.61-0.80);

Almost perfect agreement (>0.80)

Results

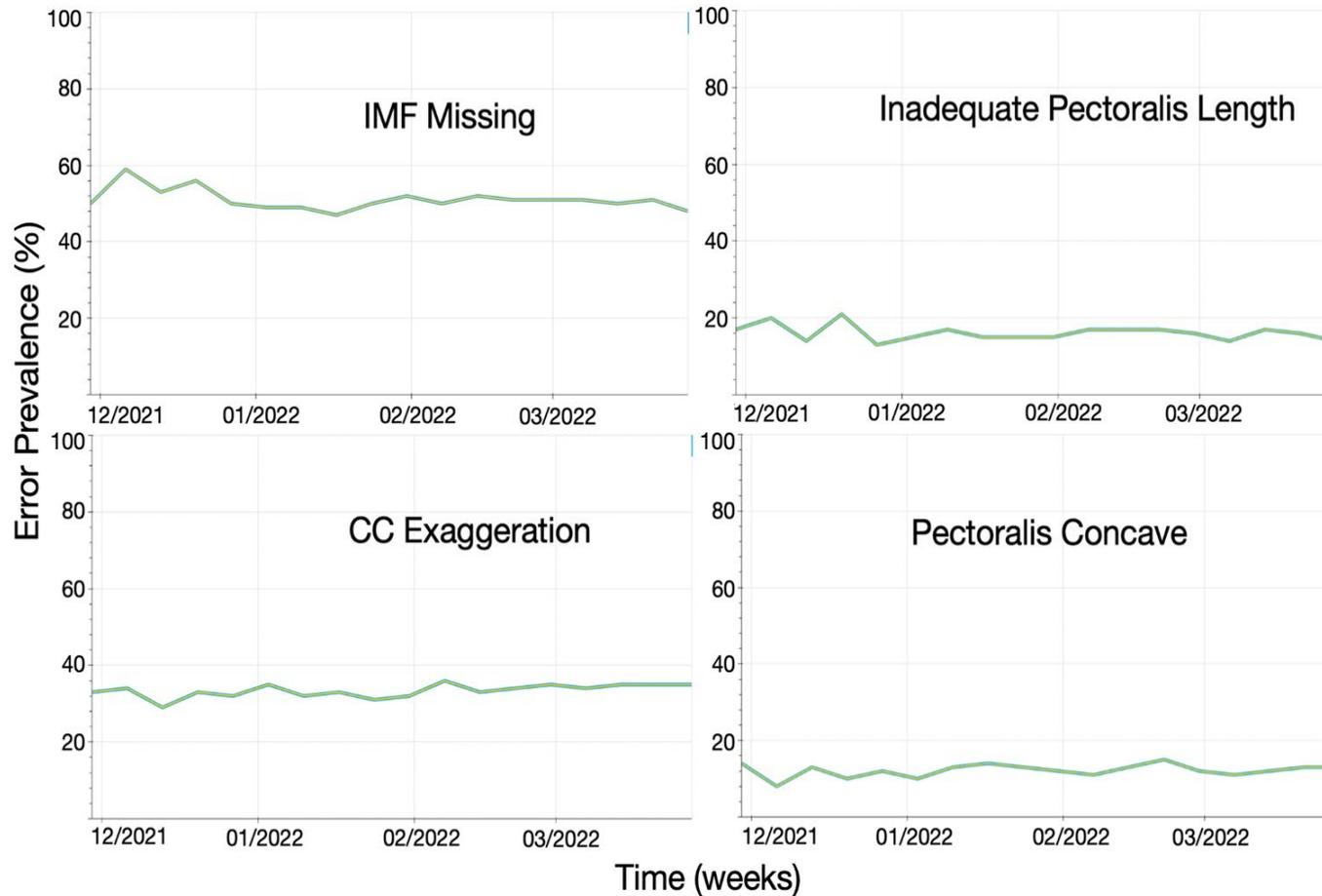
- Population-based error prevalence 12/2021-03/2022
 - As low as **3%** with IR placement error
 - As high as **51%** with IMF missing error
- 3% to 51% by AI in the population-level data, compared to...
 - ... 9% to 56% by expert assessment in the validation set
 - ... 5.5% to 40.4% by AI assessment in the medical literature
- Kappa range from **'substantial'** (>0.60) to **'almost perfect'** (>0.80)

	Leeds Teaching Hospitals NHS Trust			Medical Literature	
	Population (N=59,264)	Validation set (N=198)			
Positioning Error	AI Error Rate	Expert Error Rate	Kappa (95%CI)	AI Event Rate	Population*
Pectoralis Muscle Length	16%	10%	0.89 (0.69, 1.00)	5.5-37.8%	Norway; Canada
Pectoralis Muscle Concave	13%	12%	0.81 (0.61, 1.00)	16.0-19.2%	Norway
IMF Missing	51%	56%	0.71 (0.51, 0.91)	9.0-20.5%	Norway
IR Placement	3%	9%	0.78 (0.59, 0.98)	--	--
MLO Posterior Tissues Missing	11%	11%	0.85 (0.65, 1.00)	16.1%	Canada
CC Posterior Tissues Missing	23%	12%	0.95 (0.76, 1.00)	20.2%	Canada
CC Excessive Exaggeration	34%	17%	0.70 (0.50, 0.90)	24.4-40.4%	Norway

*Source: Norwegian error rate data [Waade 2021]; *Source: Canadian error rate data [Rouette 2021]

Weekly rate variation

- **Stable** weekly error rates



← Provides
‘baseline’
assessment for
monitoring
improvement
initiatives
(e.g. educational
interventions)

Key findings

- These findings support the use of AI for reliable and reproducible quantitative mammography positioning image quality assessments.
- Aligns with other studies suggesting AI may agree with expert assessments
 - Slight 0.06 (pec shape) to substantial agreement 0.69 (nipple not in profile)^[Waade 2021]
- With population-based error rate information at your fingertips, it is possible to evaluate image quality improvement initiatives (such as tailored educational sessions).

Future work

- Investigate Radiographer and Assistant Practitioner error rates
 - Stratification of results by years of experience
- Investigate needs-based image quality improvement initiatives
- Implement interventions and monitor the impacts on baseline error rates
- Stratify error rates by presence of patient associated limitations

Limitations

- Validation data set sample size was small with low positioning error event rates
- Did not stratify analysis by screening and diagnostic mammograms
 - Differences with imaging requirements

Conclusion

These study findings suggest that automated **A.I.** mammography positioning error assessments may provide a **feasible approach** to **measuring and monitoring** the impact of **image quality improvement initiatives** at Leeds Teaching Hospitals NHS Trust.

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