

**Amit Dang, Dimple Dang, Vallish BN, Aditya Bhardwaj**

MarksMan Healthcare Communications, Hyderabad, India

## Background

- A large number of modalities for cancer diagnosis that use different artificial intelligence (AI) and machine learning (ML) protocols are currently in various stages of development and validation across the world<sup>(1)</sup>
- Highly encouraging results are reported in terms of sensitivity, specificity, and accuracy of most of these AI/ML protocols during validation studies that are conducted under experimental settings that usually use retrospective patient databases
- We wanted to evaluate to what extent these protocols would perform under real-world conditions, and whether the physicians would routinely adopt these AI/ML modalities for clinical decision-making based on their superlative performance in validation tests

## Objective

- To systematically map the extent of actual use of AI/ ML protocols for diagnosing cancer in prospective settings across the world
- Research question:** 'What is known from published literature about the extent of actual usage of AI/ML protocols in cancer diagnosis in prospective (clinical trial/ real-world) settings, such that the diagnosis by the AI/ML protocol aids in clinical decision making?'

## Methodology

- Type of study:** Systematic Literature Review
- SLR protocol:** drafted as per PRISMA guidelines, and registered prospectively with Open Science Framework on 3rd January 2020 (<https://osf.io/643uq>)
- Databases searched:** PubMed, Google Scholar (first 200 hits)
- Date of search:** From inception till 17<sup>th</sup> May 2021

### Eligibility Criteria

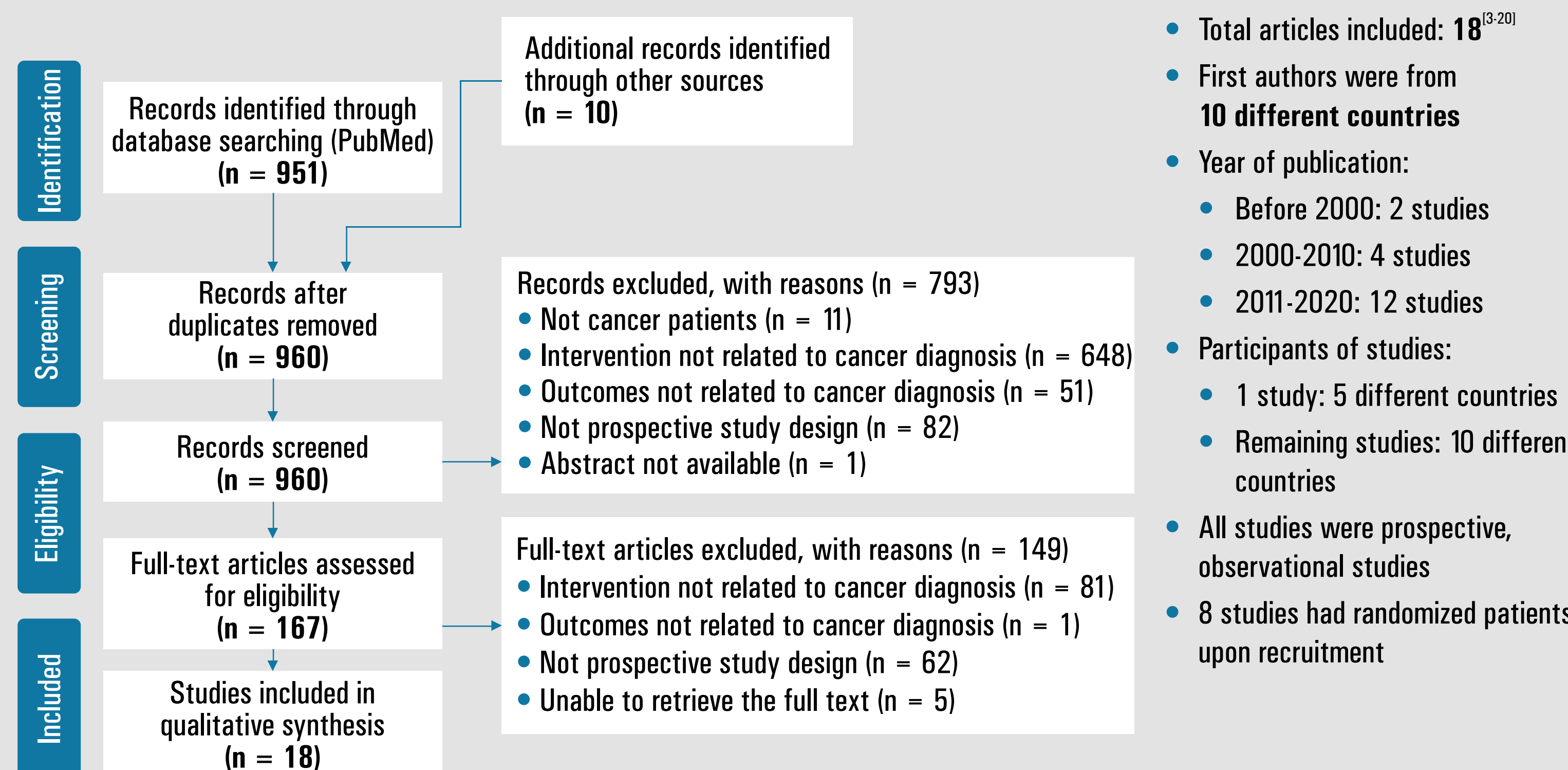
Facet	Inclusion	Exclusion
Population	<ul style="list-style-type: none"> <li>Humans suffering from any type of cancer</li> <li>Any age, any gender</li> </ul>	<ul style="list-style-type: none"> <li>No human subjects</li> <li>No cancer</li> </ul>
Intervention	<ul style="list-style-type: none"> <li>Papers which had described the actual usage of AI/ML protocol for diagnosis of cancer in such a way that the AI/ML diagnosis resulted in or aided in clinical decision making</li> </ul>	<ul style="list-style-type: none"> <li>AI/ ML protocol used for any application apart from cancer diagnosis or staging</li> <li>Robotic surgeries</li> <li>AI/ ML for estimating cancer prognosis</li> </ul>
Comparator	<ul style="list-style-type: none"> <li>Any comparator</li> </ul>	<ul style="list-style-type: none"> <li>No restriction</li> </ul>
Outcome	<ul style="list-style-type: none"> <li>Any outcome which described the application of AI/ML in cancer diagnosis</li> <li>AI/ML protocol has been used to newly diagnose a cancer or performing staging of a patient already diagnosed with cancer, thereby facilitating clinical decision-making</li> </ul>	<ul style="list-style-type: none"> <li>All other outcomes</li> </ul>
Study design	<ul style="list-style-type: none"> <li>Prospective patient enrolment</li> <li>Clinical trial or real-world setting</li> </ul>	<ul style="list-style-type: none"> <li>Retrospective data analysis</li> <li>Studies describing training, testing, or validation of AI/ ML protocols</li> <li>Reviews, editorials, commentaries</li> </ul>

### Search Strategy

No.	Terms	Hits	Facet
#1	Search: (((("artificial intelligence"[MeSH Terms]) OR ("machine learning"[MeSH Terms]) OR (artificial intelligence[Title/Abstract]) OR (machine learning[Title/Abstract])))	144,127	All types of articles dealing with artificial intelligence and/or machine learning
#2	Search: ("neoplasms"[MeSH Major Topic]) AND ("diagnosis"[MeSH Major Topic])	352,175	All types of articles dealing with any type of diagnosis of any type of cancer
#3	#1 AND #2	5,689	All types of articles dealing with AI/ ML AND cancer diagnosis
#4	Search: ("adaptive clinical trial"[Publication Type] OR "clinical study"[Publication Type] OR "clinical trial"[Publication Type] OR "clinical trial, phase i"[Publication Type] OR "clinical trial, phase ii"[Publication Type] OR "clinical trial, phase iii"[Publication Type] OR "clinical trial, phase iv"[Publication Type] OR "comparative study"[Publication Type] OR "controlled clinical trial"[Publication Type] OR "equivalence trial"[Publication Type] OR "multicenter study"[Publication Type] OR "observational study"[Publication Type] OR "pragmatic clinical trial"[Publication Type] OR "randomized controlled trial"[Publication Type])	2,797,020	All clinical trials and related articles as on date
#5	#3 AND #4	983	Studies dealing with AI/ ML AND cancer diagnosis in clinical trial and related settings
	Filters: English	951	Studies dealing with AI/ ML AND cancer diagnosis in clinical trial and related settings, reported in English Language

- Methodological quality assessment: using QUADAS-2 tool<sup>(2)</sup>
- Post-hoc analysis: After completing planned data extraction, a post-hoc analysis of all the retrieved records was performed to identify studies that described the validation of AI/ML protocol (either using standardized patient databases or prospectively enrolled patients) without their actual usage. Data pertaining to the types of cancer studied, the nature of AI/ML protocol being employed, the year of publication of the study, the country of the first author, the location of the study site, and the number of patients/ lesions/ images being used for the validation of the AI/ML protocol were extracted.
- Inter-rater reliability: Through Cohen's kappa statistic;  $\leq 0.20$  = slight agreement;  $0.21-0.40$  = fair agreement;  $0.41-0.60$  = moderate agreement;  $0.61-0.80$  = substantial agreement;  $0.81-0.99$  = near-perfect agreement; and  $1.00$  = perfect agreement

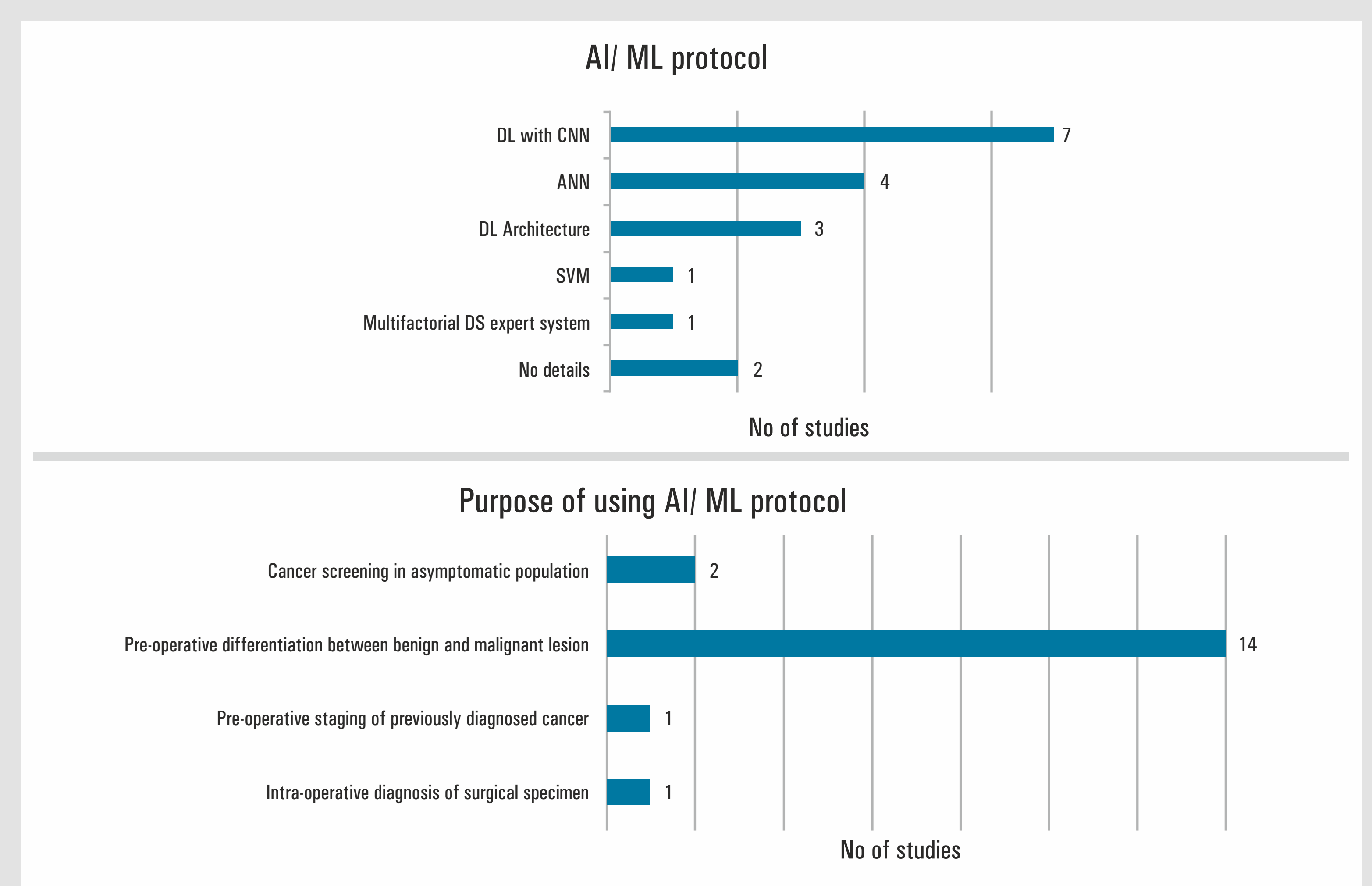
## Result



### Main Characteristics of Included Studies

No.	Study, Year <sup>(3,20)</sup>	1 <sup>st</sup> author country	Cancer studied	Type of lesions studied	AI/ML protocol	No. of patients	Male (%)	Female (%)	No. of lesions studied
1	Mori Y et al, 2018	Japan	Colorectal cancer	Colorectal Polyps	ML, SVM	325	235 (72.3%)	90 (27.7%)	466
2	Wang P et al, 2019*	China	Colorectal cancer	Colorectal Polyps	DL architecture	1,058	512 (48.4%)	546 (51.6%)	767
3	Su JR et al, 2019*	China	Colorectal cancer	Colorectal polyps	CNN, DL	623	307 (49.3%)	316 (50.7%)	442
4	Wang P et al, 2020*	China	Colorectal cancer	Colorectal polyps	DL	369	179 (48.5%)	190 (51.5%)	811
5	Repici A et al, 2020*	Italy	Colorectal cancer	Colorectal polyps	CNN, DL	685	337 (49.2%)	348 (50.8%)	493
6	Gong D et al, 2020*	China	Colorectal cancer	Colorectal polyps	CNN, DL	704	345 (49.0%)	359 (51.0%)	369
7	Wang P et al, 2020*	China	Colorectal cancer	Colorectal polyps	DL	962	495 (51.5%)	467 (48.5%)	809
8	Liu WN et al, 2020*	China	Colorectal cancer	Colorectal polyps	CNN, DL	1026	551 (53.7%)	475 (46.3%)	734
9	Dreisheitl S et al, 2009	Austria	Skin cancer	PSL	ANN-based DS tool	458	NA	NA	3,021
10	Fink C et al, 2017	Germany	Skin cancer	PSL	Not specified	111	59 (53.2%)	52 (46.8%)	346
11	Walker BN et al, 2019	USA	Skin cancer	PSL	CNN, DL	63	34 (54.0%)	29 (46.0%)	63
12	Kok MR et al, 1996	Netherlands	Cervical cancer screening	Cervical smear	ANN-based DS tool	91,294	0	91,294 (100%)	91,294
13	Nieminen P et al, 2002*	Finland	Cervical cancer screening	Cervical smear	ANN-based DS tool	108,686	0	108,686 (100%)	108,686
14	Hollon TC et al, 2020	USA	Brain cancer	Intra-op surgical specimen	CNN, DL	278	NA	NA	278
15	de Veld DC et al, 2004	Netherlands	Cancer of Oral Cavity	Oral mucosal lesion	PCA; ANN	155	NA	NA	176
16	Li L et al, 2019	China	Lung cancer	Lung nodules	CNN, DL	346	221 (63.9%)	125 (36.1%)	1916
17	Lucidarme D et al, 2010	France	Ovarian cancer	TVS image of ovary	Not specified	264	0	264 (100%)	375
18	Chang PL et al, 1999	Taiwan	Prostate cancer	Multiple parameters	Multifactorial DS system	43	43 (100%)	0	43

Note: \*Randomization was done in these studies; ANN: Artificial neural network; CNN: Convolutional neural network; DL: Deep learning; DS: Decision support; ML: Machine learning; PCA: Principal Component Analysis; PSL: Pigmented skin lesions; SVM: support vector machine

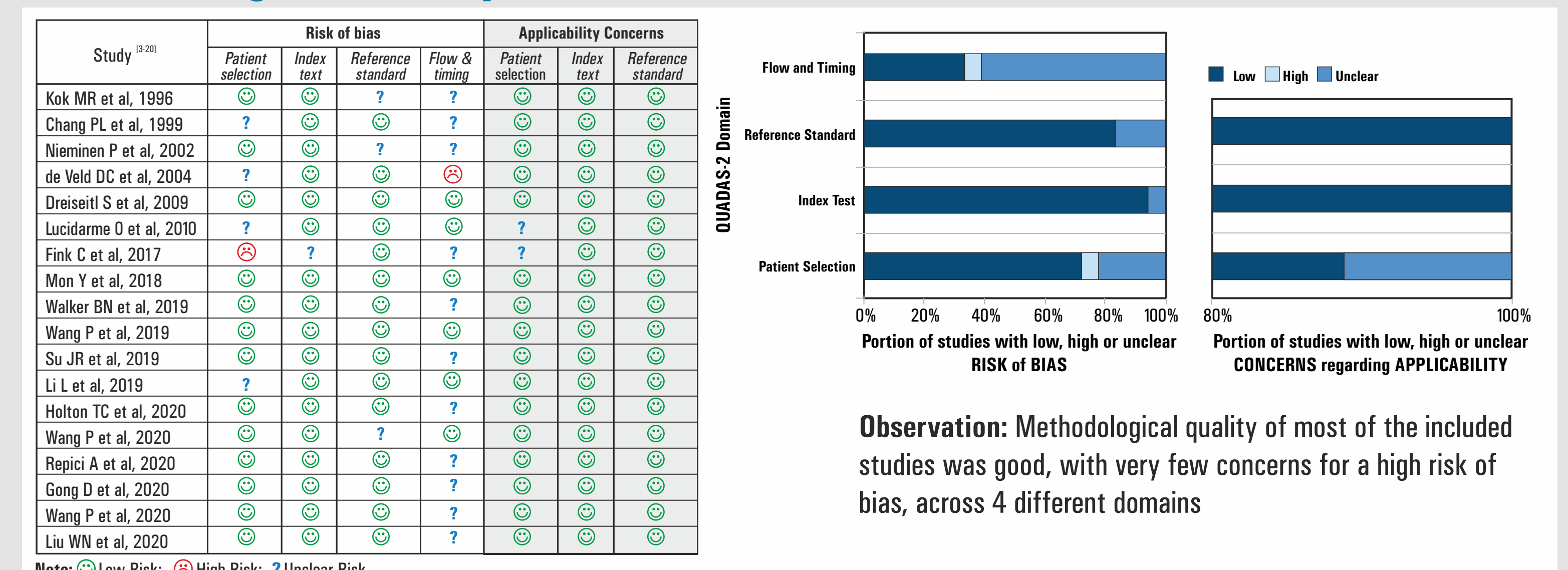


### Diagnostic Performance of Included Studies

No.	Study, Year <sup>(3,20)</sup>	Performance of AI/ML diagnosis as compared to human diagnosis	Sensitivity of the AI/ML protocol	Specificity of the AI/ML protocol	Accuracy of the AI/ML protocol	PPV of the AI/ML protocol	NPV of the AI/ML protocol
1	Chang PL et al, 1999	AI improves human diagnosis	92%	84%	88.40%	NA	NA
2	Lucidarme D et al, 2010	AI improves human diagnosis	98%	88%	NA	NA	NA
3	Wang P et al, 2019	AI improves human diagnosis	NA	NA	NA	NA	NA
4	Su JR et al, 2019	AI improves human diagnosis	NA	NA	NA	NA	NA
5	Repici A et al, 2020	AI improves human diagnosis	NA	NA	NA	NA	NA
6	Gong D et al, 2020	AI improves human diagnosis	NA	NA	NA	NA	NA
7	Wang P et al, 2020	AI improves human diagnosis	NA	NA	NA	NA	NA
8	Liu WN et al, 2020	AI improves human diagnosis	NA	NA	NA	NA	NA
9	Mori Y et al, 2018	AI is better than human diagnosis	NA	NA	98.10%	NA	93.7% to 96.5%
10	Li L et al, 2019	AI is better than human diagnosis	86.20%	NA	NA	57.00%	NA
11	Hollon TC et al, 2020	AI is better than human diagnosis	NA	NA	94.60%	NA	NA
12	Wang P et al, 2020	AI is better than human diagnosis	NA	NA	NA	NA	NA
13	Kok MR et al, 1996	AI is similar to human diagnosis	NA	NA	NA	NA	NA
14	Nieminen P et al, 2002	AI is similar to human diagnosis	NA	92.50%	NA	55%	NA
15	de Veld DC et al, 2004	Comparison not performed	NA	NA	NA	NA	NA
16	Fink C et al, 2017	Comparison not performed	100%	68.50%	2.30%	2.80%	100%
17	Walker BN et al, 2019	Comparison not performed	86% (system B); 91% (system A) <sup>1</sup>	69% (system B) <sup>1</sup>	NA	88.90%	88.90%
18	Dreisheitl S et al, 2009	Depends on the user's background	72%	82%	NA	NA	NA

Note: <sup>1</sup>System A is a deep learning classifier whose outputs from image processing of pigmented skin lesions were converted into sound waves, which were once again classified by System B. PPV: Positive predictive value; NPV: Negative predictive value

### Methodological Quality of Included Studies

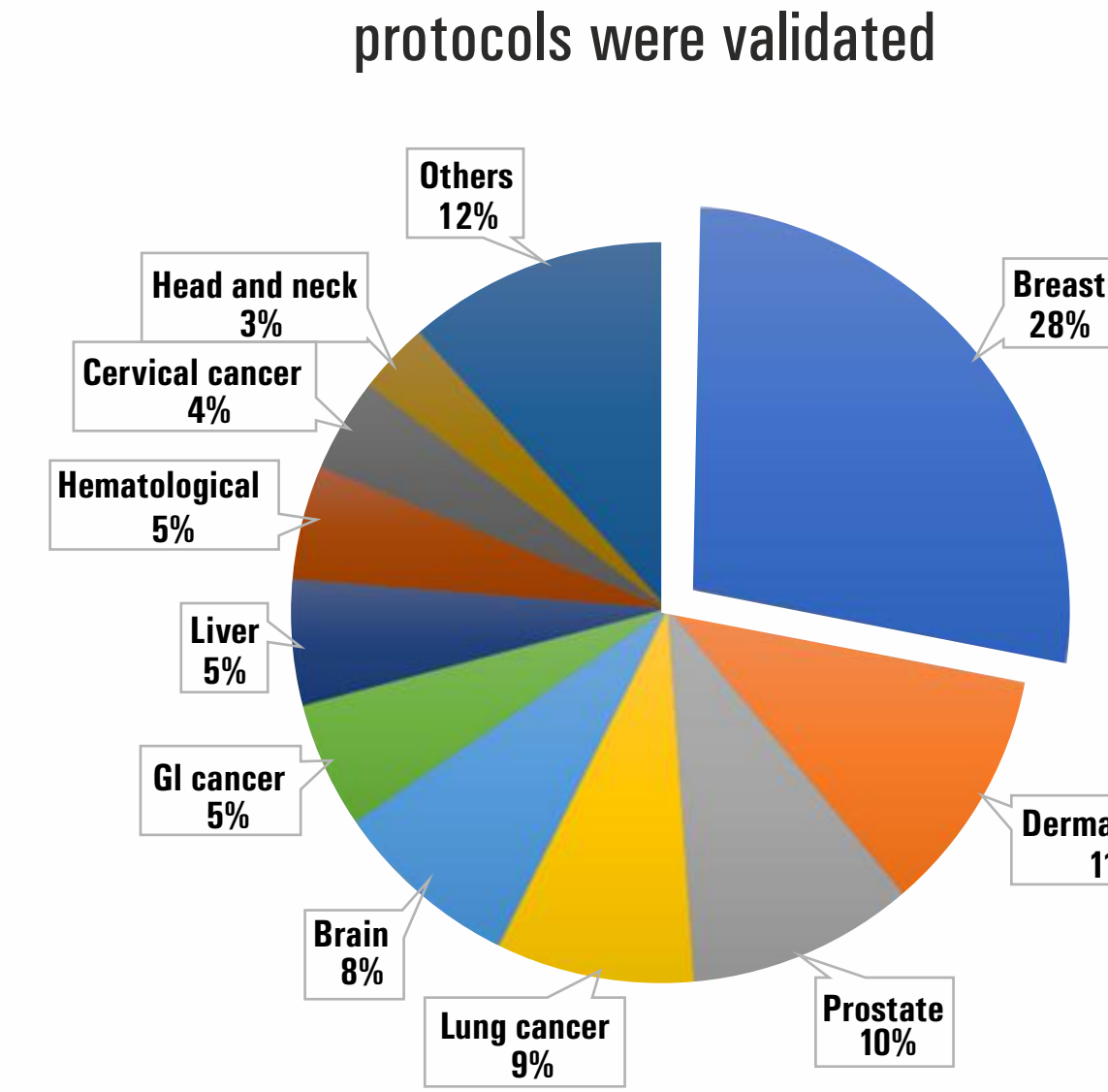


Observation: Methodological quality of most of the included studies was good, with very few concerns for a high risk of bias, across 4 different domains

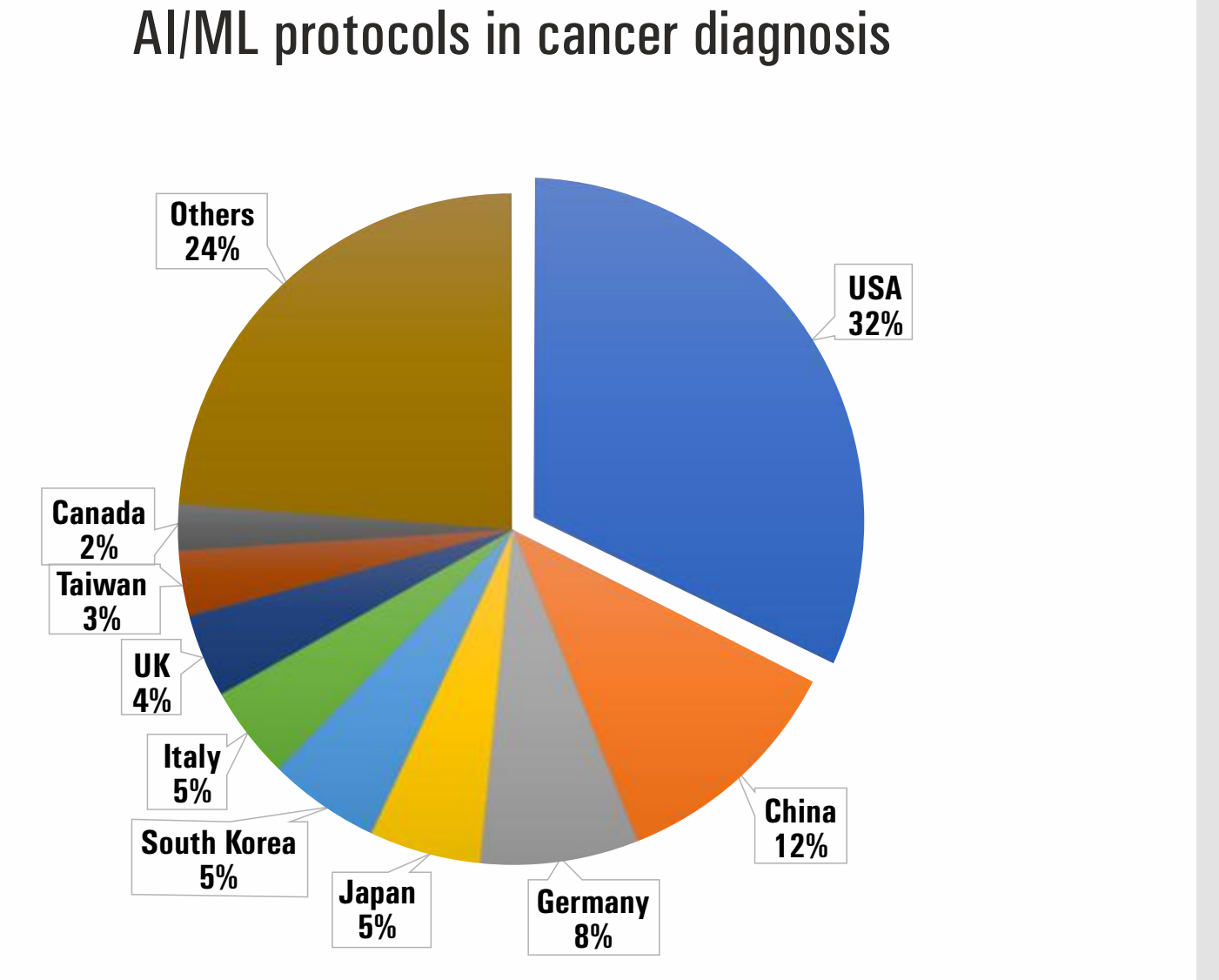
### Post-hoc Analysis

- 223 studies described validation of an AI/ML protocol in cancer diagnosis
- A huge variation in the number of samples/ patients/ lesions/ images included for validation of the AI/ML protocol was observed
  - Patient numbers ranged from 8 to 84,424
  - Image/ lesion numbers ranged from 15 to 10,36,496
- Most frequent cancer for which:
  - AI/ ML validation was done: Breast cancer
  - AI/ ML protocol was actually used: colorectal cancer

#### Type of cancers in which AI/ML protocols were validated



#### Country of first author of studies validating AI/ML protocols in cancer diagnosis



## Discussion

- Only 18/96 (1.9%) of initial hits on AI/ ML have actually used AI/ML protocols for diagnostic decision making in cancer; most excluded studies focused on validation of AI/ ML protocols
- Most studies concluded that AI/ML protocol is able to improve the human diagnosis, especially that made by the less experienced clinician: AI/ML protocols have a potential to significantly improve upon the prevailing diagnostic capabilities
- Meaningful translation of AI/ ML research into oncology diagnosis is lacking
- Performance of AI/ML protocols in validation studies is much better than that in real world studies
- Large number of validation tests, but few number of actual usage studies
- Disconnect between most frequent cancer in validation studies (breast) vs actual use (colorectal cancer)
- Large variations in the number of sample sizes in validation tests: lack of regulation in new diagnostic tests, unlike the stringent drug approval regulations

### Study limitations

- Literature search restricted to PubMed and English language articles

## Conclusions

- A meaningful translation from validation of AI/ML protocols to their actual usage in cancer diagnosis is lacking. Development of regulatory framework specific for AI/ML usage in healthcare is essential

### References

1 Kamm BH et al. Oncology (Williston Park). 2019;33:46-53. 9 Lucidarme D et al. Eur Radiol. 2010;20:1822-1830. 17 Repici A et al. Gastroenterology. 2020;158:512-520.e7.

2 Whiting PF et al. Ann Intern Med. 2011;155:529-536. 10 Fink C et al. J Dtsch Dermatol Ges. 2017;15:414-419. 18 Gong D et al. Lancet Gastroenterol Hepatol. 2020;5:352-61

3 Mori Y et al. Ann Intern Med. 2018;169:357-366. 11 Walker BN et al. EBMedicine. 2019;40:178-183. 19 Wang P et al. Lancet Gastroenterol Hepatol. 2020;5:343-51

4 Dreisheitl S et al. Melanoma Res. 2009;19:180-4. 12 Wang P et al. Gut. 2019;68:1813-1819. 20 Liu WN et al. Saudi J Gastroenterol. 2020;26:13-19.

5 Kok MR, Boon ME. Cancer. 1996;79:112-117. 13 Su JR et al. Gastrointest Endosc. 2020;91:415-424.e4. 14 Li L et al. Thorax. 2019;74:183-192.

6 Chang PL et al. Med Decis Making. 1999;19:419-427. 15 Hollon TC et al. Nat Med. 2020;26:52-58. 16 Wang P et al. Gastroenterology. 2020;158:1252-1261.e5.

7 Nieminen P et al. Int J Cancer. 2003;103:422-426. 8 de Veld DC et al. J Biomed Opt. 2004;9:940-950.