Table S1. Quality metrics of included studies according to QUADAS-2 score with each cell containing assessments from each Author.

Ref.	Au.	Risk of bias				Concerns regarding applicability		
		PATIENT SELECTION	INDEX TEST	REFERENCE STANDARD	FLOW AND TIMING	PATIENT SELECTION	INDEX TEST	REFERENCE STANDARD
[1]	1	Unclear	Unclear	Low	Low	Unclear	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[2]	1	Low	Low	Unclear	Unclear	Low	Low	Unclear
	2	Low	Unclear	Low	Low	Low	Low	Low
[3]	1	Low	Unclear	Unclear	Unclear	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[4]	1	Low	Unclear	Low	Low	Unclear	Low	Low
	2	Low	Unclear	Low	Low	Unclear	Low	Low
[5]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Unclear	Unclear	Low	Low	Unclear
[6]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Low	Unclear	Low	Low	Unclear	Low	Low
[7]	1	Unclear	Low	Low	Low	Low	Low	Unclear
	2	Low	Low	Unclear	Unclear	Low	Low	Low
[8]	1	Low	Low	Low	Low	Unclear	Low	Low
	2	Unclear	Unclear	Low	Low	Low	Low	Low
[9]	1	Low	Low	Low	Unclear	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[10]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Unclear	Unclear	Low	Low	Unclear	Low	Low
[11]	1	Low	Unclear	Low	Low	Low	Low	Low
	2	Unclear	Low	Low	Low	Low	Low	Low
[12]	1	Low	Unclear	Unclear	Unclear	Unclear	Low	Unclear
	2	Low	Unclear	Unclear	Low	Low	Low	Unclear
[13]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[14]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Low	Unclear	Low	Unclear	Unclear	Low	Low
[15]	1	Low	Unclear	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[16]	1	Low	Low	Unclear	Low	Unclear	Low	Unclear
	2	Low	Unclear	Low	Low	Low	Low	Low
[17]	1	Low	Low	Low	Unclear	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[18]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Low	Unclear	Unclear	Low	Low	Low	Unclear
[19]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Unclear	Low	Low
[20]	1	Unclear	Low	Low	Unclear	Low	Low	Low
	2	Low	Unclear	Low	Unclear	Low	Low	Unclear
[21]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[22]	1	Unclear	Low	Unclear	Unclear	Unclear	Low	Unclear
	2	Low	Unclear	Low	Low	Low	Low	Low
[23]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[24]	1	Unclear	Low	Unclear	Unclear	Unclear	Low	Low
	2	Low	Unclear	Unclear	Unclear	Low	Low	Unclear
[25]	1	Low	Low	Low	Low	Low	Low	Low
	2	Unclear	Low	Low	Low	Low	Low	Low

Ref.	Au.	Risk of bias				Concerns regarding applicability		
		PATIENT SELECTION	INDEX TEST	REFERENCE STANDARD	FLOW AND TIMING	PATIENT SELECTION	INDEX TEST	REFERENCE STANDARD
[26]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Low	Unclear	Low	Low	Low	Low	Unclear
[27]	1	Low	Low	Unclear	Unclear	Low	Low	Low
	2	Low	Low	Unclear	Unclear	Low	Low	Low
[28]	1	Low	Low	Low	Low	Low	Low	Low
	2	Unclear	Unclear	Low	Low	Unclear	Low	Unclear
[29]	1	Low	Low	Unclear	Unclear	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[30]	1	Low	Low	Low	Low	Low	Low	Unclear
	2	Unclear	Unclear	Unclear	Low	Unclear	Low	Low
[31]	1	Low	Low	Low	Low	Low	Low	Low
	2	Unclear	Low	Unclear	Low	Low	Low	Low
[32]	1	Low	Low	Unclear	Low	Low	Low	Low
	2	Unclear	Unclear	Low	Unclear	Unclear	Low	Unclear
[33]	1	Low	Low	Low	Unclear	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[34]	1	Low	Low	Low	Low	Low	Low	Low
	2	Unclear	Unclear	Unclear	Unclear	Low	Low	Low
[35]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[36]	1	Unclear	Low	Unclear	Unclear	Low	Low	Unclear
	2	Low	Unclear	Unclear	Unclear	Unclear	Low	Low
[37]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Low	Low	Low	Low
[38]	1	Low	Low	Unclear	Low	Low	Low	Low
	2	Unclear	Unclear	Low	Low	Unclear	Low	Unclear
[39]	1	Low	Low	Low	Unclear	Low	Low	Low
	2	Unclear	Low	Low	Low	Low	Low	Low
[40]	1	Low	Low	Unclear	Low	Low	Low	Low
	2	Unclear	Unclear	Low	Low	Unclear	Low	Low
[41]	1	Low	Low	Low	Low	Low	Low	Low
	2	Low	Low	Low	Unclear	Low	Low	Low
[42]	1	Low	Low	Unclear	Unclear	Low	Low	Low
	2	Unclear	Unclear	Low	Unclear	Low	Low	Unclear
[43]	1	Low	Low	Low	Low	Low	Low	Low
	2	Unclear	Low	Low	Low	Low	Low	Low
[44]	1	Unclear	Low	Unclear	Unclear	Low	Low	Unclear
	2	Low	Unclear	Unclear	Unclear	Low	Low	Low
[45]	1	Unclear	Low	Low	Low	Low	Low	Low
	2	Unclear	Low	Low	Low	Low	Low	Low

Abbreviations: Au= Author

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