

The predictive value of PRDM2 in solid tumor: a systematic review and meta-analysis

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Abstract Background. Many studies have reported the presence of Positive Regulatory/Su(var)3-9, Enhancer-of-zeste and Trithorax Domain 2 (PRDM2) downregulation in cancer. However, its potential as a diagnostic biomarker is still unclear. Hence, a systematic review and meta-analysis were conducted to address this issue.

Introduction. As of 2018, cancer has become the second leading cause of death worldwide. Thus, cancer control is exceptionally vital in reducing mortality. One such example is through early diagnosis of cancer using tumor biomarkers. Having a function as a tumor suppressor gene (TSG), *PRDM2* has been linked with carcinogenesis in several solid tumor. This study aims to assess the relationship between *PRDM2* downregulation and solid tumor, its relationship with clinicopathological data, and its potential as a diagnostic biomarker. This study also aims to evaluate the quality of the studies, data reliability and confidence in cumulative evidence.

Materials & Methods. A protocol of this study is registered at the International Prospective Register of Systematic Reviews (PROSPERO) with the following registration number: CRD42019132156. PRISMA was used as a guideline to conduct this review. A comprehensive electronic search was performed from inception to June 2019 in Pubmed, Cochrane Library, ProQuest, EBSCO and ScienceDirect. Studies were screened and included studies were identified based on the criteria made. Finally, data synthesis and quality assessment were conducted.

Results. There is a significant relationship between *PRDM2* downregulation with solid tumor (RR 4.29, 95% CI 2.58 – 7.13, $P < 0.00001$). The overall sensitivity and specificity of *PRDM2* downregulation in solid tumors is 84% (95% CI 39-98%) and 86% (95% CI 71-94%), respectively. There is a low risk of bias for the studies used. TSA results suggested the presence of marked imprecision. The overall quality of evidence for this study is very low.

Discussion. We present the first meta-analysis that investigated the potential of *PRDM2* downregulation as a diagnostic biomarker in solid tumor. In line with previous studies, our results demonstrated that *PRDM2* downregulation occurs in solid tumor. A major source of limitation in this study is the small number of studies.

Conclusions. Our review suggested that *PRDM2* is downregulated in solid tumor. The relationship between *PRDM2* downregulation and clinicopathological data is still inconclusive. Although the sensitivity

and specificity of *PRDM2* downregulation are imprecise, its high values, in addition to the evidence that suggested *PRDM2* downregulation in solid tumor, hinted that it might still have a potential to be used as a diagnostic biomarker. In order to further strengthen these findings, more research regarding *PRDM2* in solid tumors are encouraged.

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21 Abstract

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42 downregulation in solid tumors is 84% (95% CI 39-98%) and 86% (95% CI 71-94%),
43 respectively. There is a low risk of bias for the studies used. TSA results suggested the presence
44 of marked imprecision. The overall quality of evidence for this study is very low.

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48 limitation in this study is the small number of studies.

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50 relationship between *PRDM2* downregulation and clinicopathological data is still inconclusive.
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52 in addition to the evidence that suggested *PRDM2* downregulation in solid tumor, hinted that it
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54 these findings, more research regarding *PRDM2* in solid tumors are encouraged.

55

56 Introduction

57 Cancer has long been considered a catastrophic public health problem due to its high mortality
58 rates. With an estimate of 9.6 million death, cancer has now become the second leading cause of
59 death worldwide. Moreover, the incidence of cancer is also growing at an alarming rate due to
60 the exponential increase of the aging population and changes in the distribution of cancer risk
61 factors. It was estimated that the incidence of cancer would rise to 18.1 million new cases in
62 2018. To summarise, 1 in 6 women and 1 in 5 men will develop cancer, while 1 in 10 women
63 and 1 in 8 men are dying as a result of cancer (Bray et al. 2018; World Health Organization
64 2018).

65

66 Thus, cancer control is extremely vital in reducing mortality. One example of cancer control is
67 early diagnosis of cancer. This could be achieved through the use of tumor biomarkers.

68 However, despite the potential of biomarkers for early detection of cancer, its implementation in
69 the clinical setting is still lacking (Goossens et al. 2015; Schiffman et al. 2015; World Health
70 Organization 2017). This could be attributed to weak clinical performances, such as low
71 sensitivity, low specificity or low predictive values (Diamandis 2012). Hence, further research to
72 identify novel biomarkers should be performed.

73

74 Positive Regulatory/Su(var)3-9, Enhancer-of-zeste and Trithorax Domain 2 (*PRDM2*) is a tumor
75 suppressor gene (TSG) that regulates protein expression through the methylation of lysine 9 in
76 histone H3. Hence, *PRDM2* also belongs to the nuclear histone/protein methyltransferase
77 superfamily. Its gene products are also involved in DNA-binding and transcription factor
78 binding-activities, implicating its role in carcinogenesis (Sorrentino et al. 2018; Zhang et al.
79 2015). Studies have also reported *PRDM2* downregulation in cancers that exhibit high incidence
80 and mortality, such as bladder cancer, breast cancer, cervical cancer, colorectal cancer,
81 endometrial cancer, esophageal squamous cell carcinoma, gastric carcinoma, hepatocellular
82 carcinoma, lung cancer, pancreatic cancer, prostate cancer, T-cell prolymphocytic leukemia and

83 thyroid carcinoma (Cheng et al. 2010; Cui et al. 2016; Johansson et al. 2018; Lal et al. 2006;
84 Michalak & Visvader 2016; Oshimo et al. 2004; Pandzic et al. 2017; Rossi et al. 2009; Sakurada
85 et al. 2001; Tan et al. 2014; Wu et al. 2016; Yang et al. 2017; Zhang et al. 2016). Furthermore, in
86 a meta-analysis that found a total of 22 genes methylated in hepatocellular carcinoma, *PRDM2*
87 was one of the genes with the most significant result and is on par with the well-known *APC* and
88 *p16* (Zhang et al. 2016). Hence, *PRDM2* might play an important role in malignancies. However,
89 the potential of *PRDM2* as a diagnostic biomarker is still unclear.

90

91 Therefore, we performed a systematic review and meta-analysis that investigated *PRDM2*
92 expression level in solid tumor, as well as its potential as a diagnostic biomarker. If there is
93 sufficient data, we will also investigate if there is any correlation between *PRDM2* expression
94 level with clinicopathological data.

95

96 **Materials and Methods**

97 **Study Registration and Methodology**

98 A protocol of this study is registered at the International Prospective Register of Systematic
99 Reviews (PROSPERO) with the following registration number: CRD42019132156
100 (https://www.crd.york.ac.uk/prospero/display_record.php?RecordID=132156) (National Institute
101 for Health Research). Preferred Reporting Items for Systematic Reviews and Meta-analyses
102 (PRISMA) flow diagram was used as a guideline to conduct our systematic review and meta-
103 analysis (Moher et al. 2009).

104

105 **Search Strategy and Study Selection**

106 A comprehensive electronic search was done in PubMed, Cochrane Library, ProQuest, EBSCO
107 and ScienceDirect from inception to July 2019 using the following search terms: (PRDM2 OR
108 RIZ OR RIZ1 OR RIZ2 OR KMT8 OR KMT8A OR MTB-ZF OR HUMHOXY1) AND (Cancer
109 OR Cancers OR Malignant OR Malignancy OR Malignancies OR Neoplasm OR Neoplasms OR
110 Neoplasia OR Neoplasias OR Tumor OR Tumors OR Tumour OR Tumours). The search was
111 performed by two independent reviewers (Alfredo Bambang and Indra Putra Wendi). Any
112 differences were solved through a discussion with a third reviewer (Anton Sumarpo).

113

114 All of the search outputs were exported into the EndNote software. Duplicates were removed,
115 and screening was performed based on the title and abstract of the study. Probable or included
116 studies were identified and assessed for eligibility according to the criteria above. Finally,
117 included studies were identified, and data extraction was performed.

118

119 A study is included if it meets the following criteria: (1) The study used human subjects; (2) The
120 study investigated the relationship between *PRDM2* expression level and solid tumor through the
121 use of gene expression analysis; (3) The study used histopathological examination as a
122 comparator; (4) The study is a clinical trial or cross-sectional study. A study is excluded if: (1)

123 The study does not have a control group (people without cancer or non-cancer specimens); (2)
124 The study did not use an appropriate or did not state the gene expression analysis method used;
125 (3) The expression level of *PRDM2* in the study is not clearly stated or unquantifiable; (4) The
126 study is a review, case series, conference abstracts, in vitro or in vivo study. (5) The study is not
127 written in English.

128

129 **Data Extraction**

130 The included studies were then analyzed further and the following informations are extracted:
131 First author, publication year, country of origin, age, gender, race, type of cancer, cancer
132 differentiation state, stage of cancer, type of control, number of cases and controls, gene
133 expression analysis method, *PRDM2* expression level and conclusion of the study. In the case of
134 missing data, the authors will be contacted via email to request access to those missing data.

135

136 **Data Synthesis and Statistical Analysis**

137 Sensitivity and specificity of *PRDM2* were assessed in order to elucidate the potential of *PRDM2*
138 expression level as a diagnostic biomarker in solid tumor. Sensitivity and specificity are said to
139 be significant if $>50\%$. Risk ratio (RR) with a 95% confidence interval (CI) was used to
140 determine the relationship between *PRDM2* expression level and risk of cancer, as well as the
141 relationship between *PRDM2* expression level and clinicopathological data. If heterogeneity is
142 present, Random Effects Model (REM) will be used. However, if heterogeneity is absent, Fixed
143 Effects Model (FEM) will be used instead.

144

145 Cochrane's Q test (chi-squared test) and Higgins I^2 statistics were used to assess for the presence
146 of heterogeneity statistically. Heterogeneity is said to be present if $P < 0.10$ or $I^2 > 75\%$ (Higgins &
147 Green 2011; Higgins et al. 2003). To assess for the presence of heterogeneity visually, a forest
148 plot will be generated. Meta-regression and subgroup analysis will be conducted when there are
149 at least 10 studies used in the meta-analysis (Baker et al. 2009). The possible causes of
150 heterogeneity are: Age, gender, ethnicity, country of origin, type of cancer, cancer differentiation
151 state, stage of cancer and genotyping method.

152

153 Funnel plot and Deek's test will be used to assess publication bias when the number of included
154 studies is at least 10. If the funnel plot is asymmetric, publication bias is present. If the P-value
155 for Deek's test is < 0.10 , there is funnel plot asymmetry (Deeks et al. 2005). If publication bias is
156 found, the trim and fill method will be used to correct this bias (Duval & Tweedie 2000).

157

158 Furthermore, sensitivity analysis was performed to elucidate the effect and stability of a single
159 study on the pooled estimates by deleting one study at a time. Additionally, sensitivity analysis
160 was also conducted to compare the pooled estimates using odds ratio (OR) and RR, as well as
161 using REM and FEM. All statistical analyses were generated using RevMan 5.3 and STATA
162 12.0.

163

164 **Quality Assessment and Data Reliability**

165 In order to claim that the meta-analysis conducted has been conclusive, the required information
166 size has to be achieved. Thus, a trial sequential analysis (TSA) was performed using TSA
167 software in order to determine the required information size (Wetterslev et al. 2017). Quality of
168 evidence will be assessed using Quality Assessment of Diagnostic Accuracy Studies - 2
169 (QUADAS-2) which consists of the following key domains: patient selection, index test,
170 reference standard, as well as flow and timing (Whiting et al. 2011).

171

172 **Confidence in Cumulative Evidence**

173 Grading of Recommendations, Assessment, Development, and Evaluations (GRADE) was used
174 to evaluate the confidence in cumulative evidence. Overall certainty of evidence can be written
175 as high, moderate, low or very low (Schünemann et al. 2013).

176

177 **Results**

178 **Search Results**

179 Using variants of the keywords “PRDM2” and “cancer”, we performed a search from inception
180 to July 2019 in PubMed, Cochrane Library, ProQuest, EBSCO and ScienceDirect. After
181 duplicate removal, a total of 3928 records was obtained. Titles and abstracts were screened and
182 58 potential studies were identified. Out of these 58 studies, 52 were excluded due to the studies
183 being unable to meet the inclusion criteria (ineligible), in vitro and/or in vivo, used unsuitable
184 methods, written in non-English, or is a review. The remaining six studies (Akahira et al. 2007;
185 Dong et al. 2012; Ge et al. 2015; Geli et al. 2005; Jiang et al. 1999; Tan et al. 2018) were
186 included in the systematic review while only five studies (Akahira et al. 2007; Dong et al. 2012;
187 Geli et al. 2005; Jiang et al. 1999; Tan et al. 2018) were included in the meta-analysis. This is
188 because Ge et al. (2015) did not mention the number of samples and controls that expressed
189 *PRDM2* downregulation in renal cell carcinoma. Thus, only five studies were included in the
190 meta-analysis. The Preferred Reporting Items for Systematic Reviews and Meta-analyses
191 (PRISMA) flow diagram for this study is shown in Fig. 1.

192

193 The studies that were eligible for systematic review were published from 1999 to 2015. There
194 were a total of 314 samples of solid tumors and 225 controls obtained from patients in China
195 (Dong et al. 2012; Ge et al. 2015; Tan et al. 2018), Japan (Akahira et al. 2007), Sweden (Geli et
196 al. 2005) and United States of America (Jiang et al. 1999). All of these six studies are cross-
197 sectional studies. The solid tumors included in this study are ovarian cancer (Akahira et al.
198 2007), esophageal squamous cell carcinoma (Dong et al. 2012), renal cell carcinoma (RCC) (Ge
199 et al. 2015), pheochromocytoma (Geli et al. 2005), abdominal paraganglioma (Geli et al. 2005),
200 hepatoma (Jiang et al. 1999), lung squamous cell carcinoma (LSCC) (Tan et al. 2018) and lung
201 adenocarcinoma (LAC) (Tan et al. 2018). Out of these six studies, one used
202 immunohistochemistry (IHC) only (Akahira et al. 2007), three used reverse transcription-

203 polymerase chain reaction (RT-PCR) only (Ge et al. 2015; Geli et al. 2005; Jiang et al. 1999) and
204 two used both IHC and RT-PCR (Dong et al. 2012; Tan et al. 2018). A summary of the main
205 characteristics of the included studies for systematic review and meta-analysis is presented in
206 Table 1 and Table 2, respectively.

207

208 **Systematic Review Results**

209 All six studies concluded that *PRDM2* gene expression is significantly decreased in solid tumor
210 compared to control, with the P-value ranging from <0.05 to <0.001 using CI 95%. Akahira et al.
211 (2007) stated that there was a significant correlation between *PRDM2* downregulation with
212 cancer grade ($P<0.0345$) and stage ($P<0.0153$) in ovarian cancer. On the other hand, Ge et al.
213 (2015) stated otherwise, concluding that there was no significant relationship between RCC with
214 tumor progression ($P=0.19$). A study by Geli et al. (2005) reported that decreased *PRDM2* gene
215 expression was not correlated significantly with gender and tumor size, but was found to be
216 weakly correlated with younger age (Spearman rank-order correlations; $R=0.4$). Other
217 clinicopathological data were either absent or not investigated in the studies. Hence, the role of
218 *PRDM2* downregulation in cancer grade, stage, gender, age and other clinicopathological data is
219 still unclear. Due to the lack of sufficient clinicopathological data, only *PRDM2* gene expression
220 and its sensitivity and specificity were further analysed in the meta-analysis.

221

222 **Meta-Analysis Results**

223 Five studies were included in this meta-analysis to further investigate the relationship between
224 *PRDM2* downregulation with solid tumor (Akahira et al. 2007; Dong et al. 2012; Geli et al.
225 2005; Jiang et al. 1999; Tan et al. 2018). The pooled analysis suggested that *PRDM2* gene
226 expression is decreased in solid tumor (RR 4.29, 95% CI 2.58 – 7.13, $P < 0.00001$; Fig. 2).
227 Based on this pooled analysis, three sensitivity analyses were conducted to evaluate the stability
228 of our findings: with and without the deletion of Jiang et al. (1999) (Fig. 3), RR or OR (Fig. 4),
229 and FEM or REM (Fig. 5). All three sensitivity analyses did not have meaningful differences,
230 proving that our results are stable.

231

232 The sensitivity and specificity of *PRDM2* downregulation in solid tumor were also assessed in
233 order to investigate its potential as a diagnostic biomarker. A split forest plot displaying the
234 sensitivity and specificity of the included studies is shown in Fig. 6. As demonstrated in the
235 summary receiver operating characteristic (SROC) curve (Fig. 7), the summary sensitivity and
236 specificity of decreased *PRDM2* gene expression in solid tumor is 84% (95% CI 39-98%) and
237 86% (95% CI 71-94%), respectively. This result is in favor of *PRDM2* downregulation as a
238 potential diagnostic biomarker. However, the confidence interval for *PRDM2* downregulation is
239 wide, suggesting that there is marked imprecision. This was later confirmed on TSA (Fig. 8). In
240 Fig. 8, the line representing the cumulative Z-curve failed to cross the significance boundary and
241 did not reach the required number of studies which is 7743. Therefore, it can be concluded that

242 the usage of *PRDM2* downregulation as a diagnostic biomarker in solid tumor is still
243 inconclusive.

244

245 **Quality Assessment of Included Studies**

246 The quality of the included studied was evaluated using the QUADAS-2 tool, and a summary of
247 the results can be viewed in Table 3. As shown in Table 3, in the index test domain there are four
248 studies (Ge et al. 2015; Geli et al. 2005; Jiang et al. 1999; Tan et al. 2018) having an unclear risk
249 of bias. These four studies did not directly state whether the index test (gene expression analysis)
250 was interpreted independently from the reference standard (histopathological examination).

251 Thus, we decided that unclear was most fit as the risk of bias. One of the studies, Jiang et al.
252 (1999) also had missing information on how the patients were recruited, leading to an unclear
253 risk of bias for one other domain. In general, the quality of the included studies was robust,
254 ensuring the reliability of our systematic review and meta-analysis.

255

256 **Confidence in Cumulative Evidence**

257 By assessing five domains, including the risk of bias (by using the results from QUADAS-2 risk
258 of bias assessment), indirectness, inconsistency, imprecision (by using the results from TSA) and
259 risk of publication bias, a GRADE evidence profile was constructed as shown in Table 4. To be
260 noted, all of the included studies used diagnostic accuracy test as their design, whereby all of the
261 samples and controls will undergo both the index test and reference standard. Ideally, diagnostic
262 studies should randomize which of the samples and controls will undergo the index test only and
263 which will undergo the reference standard only. Hence, this made the design susceptible to
264 indirectness. In addition, most of the included studies have wide confidence interval and
265 inconclusive TSA results. Thus, serious was placed in the imprecision domain. As for
266 publication bias, since the number of included studies is <10, publication bias could not be
267 evaluated. Unfortunately, this does not entirely rule out the possibility of publication bias being
268 present in our study, and thus we decided to downgrade the quality of evidence further. Overall,
269 we have very low confidence in the pooled estimates obtained for our meta-analysis.

270

271 **Discussion**

272 In this study, we have successfully generated the first meta-analysis that investigated the
273 potential of *PRDM2* downregulation as a diagnostic biomarker in solid tumor. Compared to
274 previous primary studies on *PRDM2* thus far, we investigated the significance of *PRDM2* with
275 solid tumor on the level of a review. This includes the evaluation of quality assessment, data
276 reliability and confidence in cumulative evidence, proving that our study was more
277 comprehensive.

278

279 Meta-regression, funnel plot and Deek's test were not performed due to the small number of
280 studies obtained. Due to the inability to confirm the presence of publication bias, we also could

281 not perform trim and fill method. Since our results indicated that there was no heterogeneity in
282 the studies used, a subgroup analysis was not required.

283

284 In line with previous studies, our results demonstrated that *PRDM2* downregulation occurs in
285 ovarian cancer, esophageal squamous cell carcinoma, hepatoma and lung cancer. According to
286 Sorrentino et al. (2018), *PRDM2* downregulation has also been reported in neuroblastoma, breast
287 cancers, melanoma, parathyroid adenoma and Merkel cell carcinoma. However, our included
288 studies did not investigate those solid tumors. Another notable difference is the inconclusive
289 results linking *PRDM2* downregulation with cancer stage and grade even though *PRDM2*
290 downregulation has been associated with cancer progression (Sun et al. 2011). A possible
291 explanation for these inconsistencies might be due to the fact that our study only accepted human
292 studies, and thus limited the possibility of encountering such studies. Interestingly, all of the
293 individual studies did not have a standardised baseline to define *PRDM2* downregulation.
294 Although this could lead to possible heterogeneity, our study demonstrated otherwise.

295

296 Following these findings, an important question to address is whether *PRDM2* downregulation
297 could be used as a diagnostic biomarker in solid tumor. As described above, the high sensitivity
298 and specificity of *PRDM2* downregulation suggested its potential as a diagnostic biomarker.
299 However, these values have wide confidence intervals and inconclusive TSA results, implying
300 there was marked imprecision (Chai-Adisaksopha et al. 2016; Tan & Tan 2010). Thus, the use of
301 *PRDM2* downregulation as a diagnostic biomarker is still inconclusive. This imprecision might
302 be due to the small number of sample and controls used in the individual studies or low
303 variability in the subjects used (Carlson & Morrison 2009). In addition, there was also a vast
304 difference between the sample and control size, whereby the sample size is much larger. We
305 believe that this was because some of the studies did not obtain their sample and control from the
306 same subject. This made acquirement of control samples, such as normal ovaries or normal
307 adrenal cells, much more difficult when compared to pathological samples that are readily
308 retrieved for examination. Although our present study could not fully prove the potential of
309 *PRDM2* downregulation as a diagnostic biomarker due to its imprecision, it is important to
310 highlight that these results can potentially improve with the addition of new studies. This has
311 been proven by our TSA results whereby the line representing the cumulative Z-curve did not
312 cross the futility boundary.

313

314 Another issue that should be addressed in the future is whether the quality of our evidence is
315 satisfying enough. The quality of evidence is judged based on five domains: risk of bias,
316 indirectness, inconsistency, imprecision and publication bias. It should be noted that all of the
317 studies used in this review are diagnostic accuracy studies which are considered a proxy to
318 randomised-controlled trials. Hence, indirectness is present, and this could lead to overestimation
319 of sensitivity and specificity, resulting in the downgrading of the quality of evidence (Schmidt &
320 Factor 2013). As discussed before, imprecision is present, and publication bias could not be

321 assessed, leading to further downgrading. Together, these three domains led to the downgrading
322 of the quality of evidence from high to very low. Although there is very low confidence for our
323 results, it is important to highlight once again that these results can improve if new studies are
324 added.

325

326 Limitations of our study are the lack of RCTs as part of our included studies which made it
327 difficult to evaluate the internal validity of our results (Carlson & Morrison 2009). As mentioned
328 before, our study also lacks clinicopathological data in order to assess the potential of *PRDM2*
329 further. Interestingly, none of the included studies investigated *PRDM2* gene expression in the
330 same type of solid tumor. Hence, we were unable to evaluate in which type of solid tumor is
331 *PRDM2* downregulation most suitable to be used as a biomarker. Furthermore, there was no
332 standardised baseline among studies. Another limitation of this study involves the issue of only
333 using studies written in English, leading to the possibility of language bias. Most of the
334 individual studies have a wide confidence interval and inconclusive TSA results, indicating there
335 is insufficient knowledge about the effect and that further research should be done. Based on the
336 points above, it can be concluded that a major source of limitation is due to the small number of
337 studies.

338

339 **Conclusions**

340 In conclusion, our review suggested that *PRDM2* gene expression is decreased or downregulated
341 in solid tumor. Due to insufficient data, we are unable to determine the relationship between
342 *PRDM2* downregulation and clinicopathological data. Although the sensitivity and specificity of
343 *PRDM2* downregulation are imprecise, its high values, in addition to the evidence that suggested
344 *PRDM2* downregulation in solid tumor, hinted that it might still have a potential to be used as a
345 diagnostic biomarker. Furthermore, its imprecision could potentially be solved through the
346 addition of new studies. Thus, we suggest more research to be conducted, especially those with
347 RCT as their design, to fully elucidate the potential of *PRDM2* downregulation in solid tumor.
348 More study is urgently needed to determine a standardised baseline for *PRDM2* downregulation
349 level. We would also recommend more research regarding the relationship between *PRDM2*
350 gene expression with clinicopathological data to further evaluate the potential of *PRDM2* gene
351 expression in solid tumor. Finally, once there is sufficient data available, we suggest a new
352 systematic review and meta-analysis to be done in order to renew the findings of our study.

353

354 **Abbreviations**

355 **CI:** Confidence interval

356 **Df:** Degree of freedom

357 **F:** Female

358 **FN:** False negative

359 **FP:** False positive

360 **GRADE:** Grading of recommendations, assessment, development, and evaluations

361 **IHC:** Immunohistochemistry
362 **LAC:** Lung adenocarcinoma
363 **LSCC:** Lung squamous cell carcinoma
364 **M:** Male
365 **M-H:** Mantel-Haenszel
366 **ND:** Not determined
367 **OR:** Odds ratio
368 **PRDM2:** Positive regulatory/su(var)3-9, enhancer-of-zeste and trithorax domain 2
369 **PRISMA:** Preferred reporting items for systematic reviews and meta-analyses
370 **PROSPERO:** International prospective register of systematic reviews
371 **qRT-PCR:** Quantitative reverse transcription-polymerase chain reaction
372 **QUADAS-2:** Quality assessment of diagnostic accuracy studies - 2
373 **RCC:** Renal cell carcinoma
374 **REM:** Random effects model
375 **RR:** Risk ratio
376 **RT-PCR:** Reverse transcription-polymerase chain reaction
377 **SROC:** Summary receiver operating characteristic
378 **TN:** True negative
379 **TP:** True positive
380 **TSA:** Trial sequential analysis
381 **TSG:** Tumor suppressor gene

382

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386

387 **Reference**

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Figure 1

PRISMA flow diagram for selection of included studies.

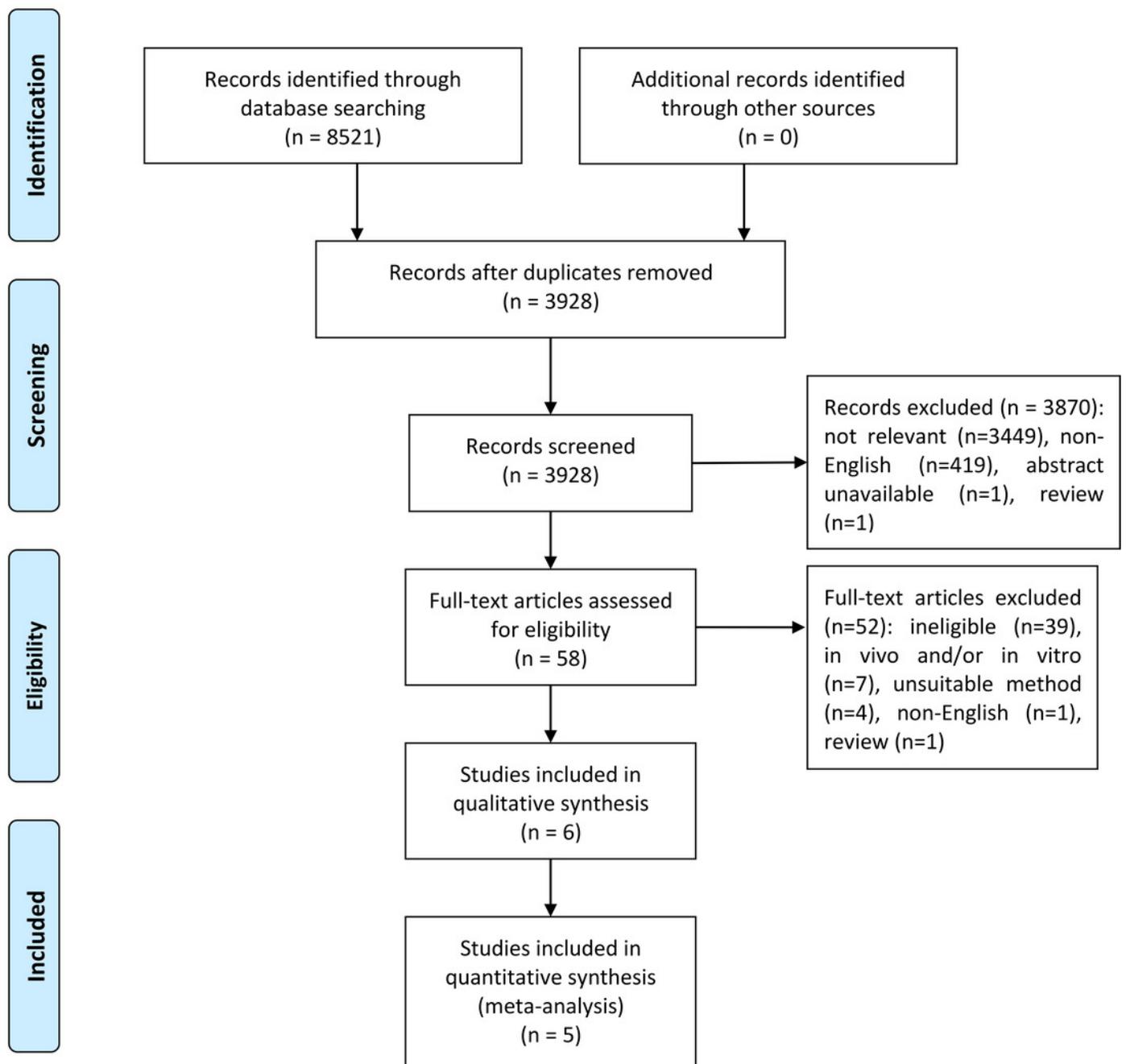


Figure 2

Forest plot of *PRDM2* downregulation in solid tumors and control.

Studies with notable weights are Tan et al. (2018) (46.2%) and Dong et al. (2012) (31.2%). The results from this forest plot demonstrated that *PRDM2* downregulation occurs more often in solid tumor when compared to control (RR 4.29, 95% CI 2.58-7.13, $P < 0.00001$). There was no significant heterogeneity in this analysis ($X^2 = 2.85$, $I^2 = 0\%$). The horizontal line represents 95% CI. The blue box is the result of each individual study. The black diamond at the bottom of the plot is the pooled analysis of all studies. CI = Confidence interval. df = Degree of freedom. I^2 = Test of heterogeneity. M-H = Mantel-Haenszel.

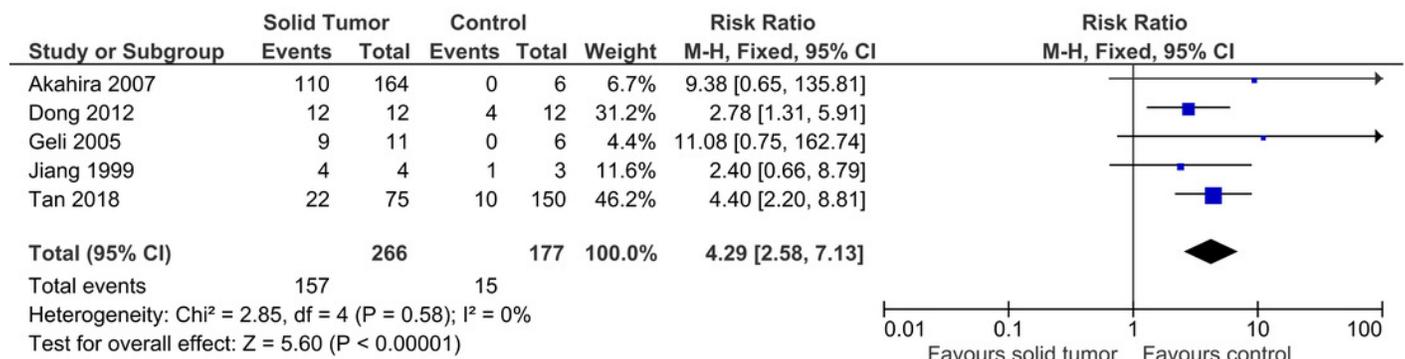
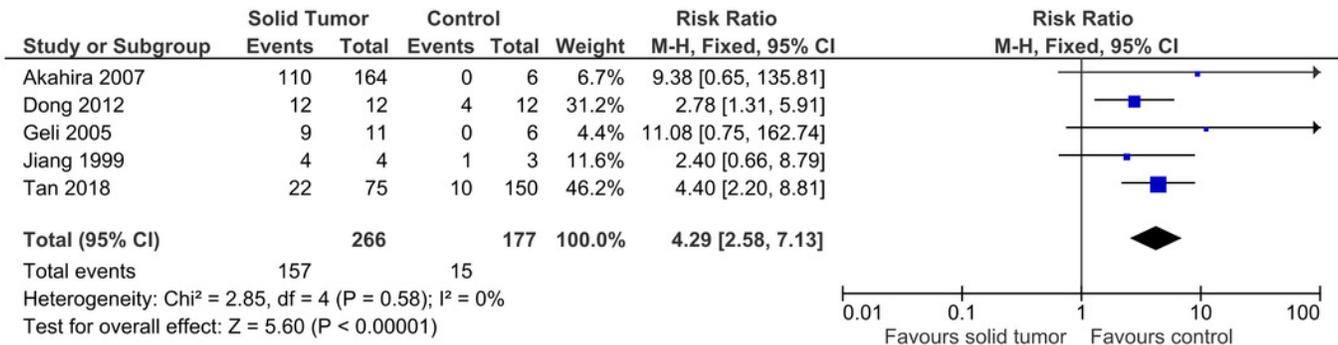


Figure 3

Sensitivity analysis to compare the use of all studies with deletion of a study.

The deleted study, Jiang et al. (1999), is a study that has the most questionable results based on the risk of bias assessment. There was a significant result for both analyses: (A) Without deletion of Jiang et al. (1999): RR 4.29, 95% CI 2.58-7.13, $P < 0.00001$, $X^2 = 2.85$, $P < 0.58$, $I^2 = 0\%$; (B) With deletion of Jiang et al. (1999): RR 4.53, 95% CI 2.63-7.82, $P < 0.00001$, $X^2 = 2.33$, $P < 0.51$, $I^2 = 0\%$. The deletion of Jiang et al. (1999) increased RR by 1.1 times higher with the 95% CI 1.2 times wider. The deletion of study also slightly lowered heterogeneity. This sensitivity analysis proved that the results were stable. The horizontal line represents 95% CI. The blue box is the result of each individual study. CI = Confidence interval. df = Degree of freedom. I^2 = Test of heterogeneity. M-H = Mantel-Haenszel.

A



B

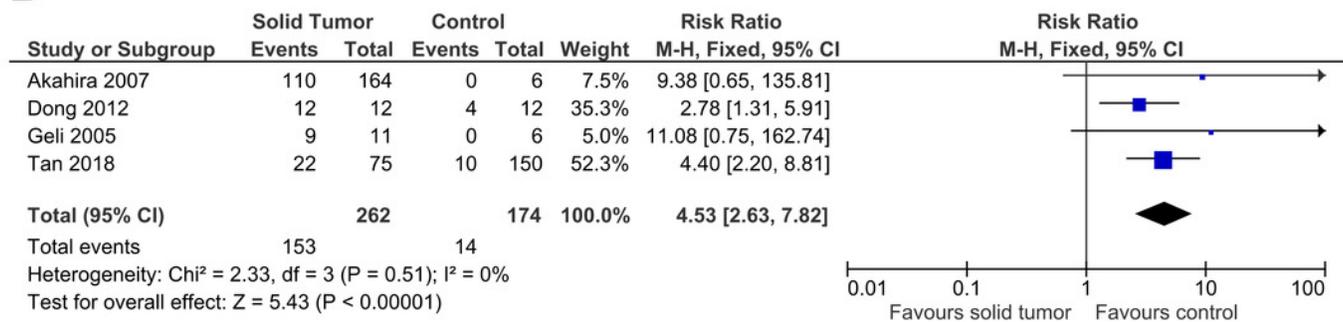
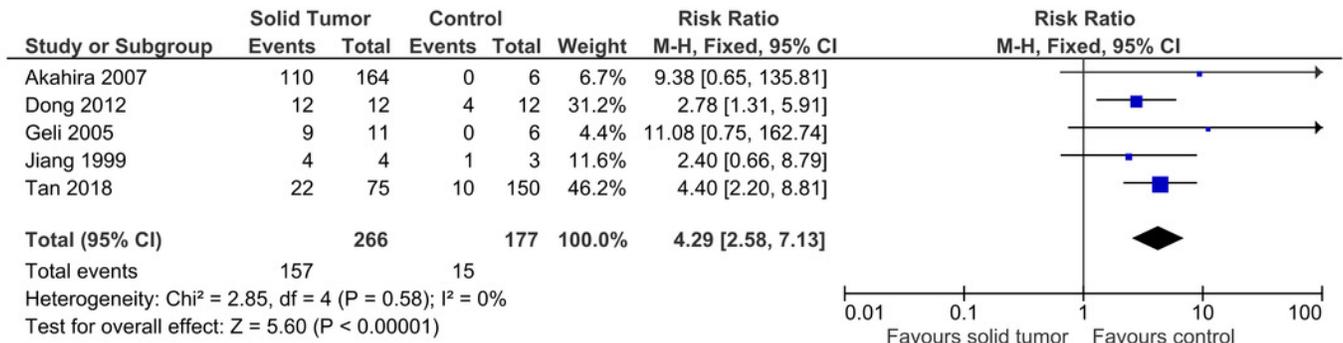


Figure 4

Sensitivity analysis to compare the use of Risk Ratio (RR) with Odds Ratio (OR).

There was a significant result for both analyses: (A) RR: RR 4.29, 95% CI 2.58-7.13, $P < 0.00001$; (B) OR: OR 9.62, 95% CI 4.82-19.19, $P < 0.00001$). The use of OR gave a result two times higher with the 95% CI three times wider when compared to RR. RR had a slightly lower heterogeneity when compared to OR (RR: $X^2 = 2.58$, $P < 0.58$, $I^2 = 0\%$; OR: $X^2 = 4.06$, $P < 0.40$, $I^2 = 1\%$). This sensitivity analysis proved that the results were stable. The horizontal line represents 95% CI. The blue box is the result of each individual study. The black diamond at the bottom of the plot is the pooled analysis of all studies. CI = Confidence interval. df = Degree of freedom. I^2 = Test of heterogeneity. M-H = Mantel-Haenszel.

A



B

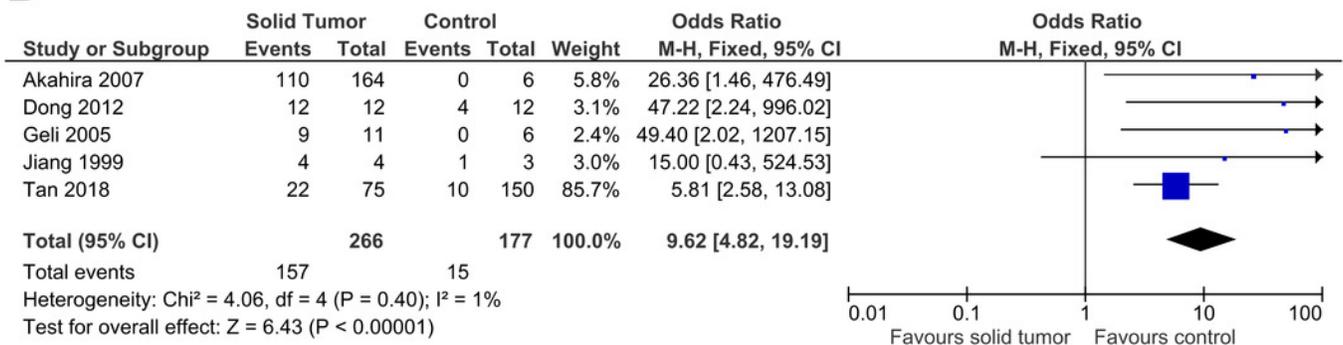
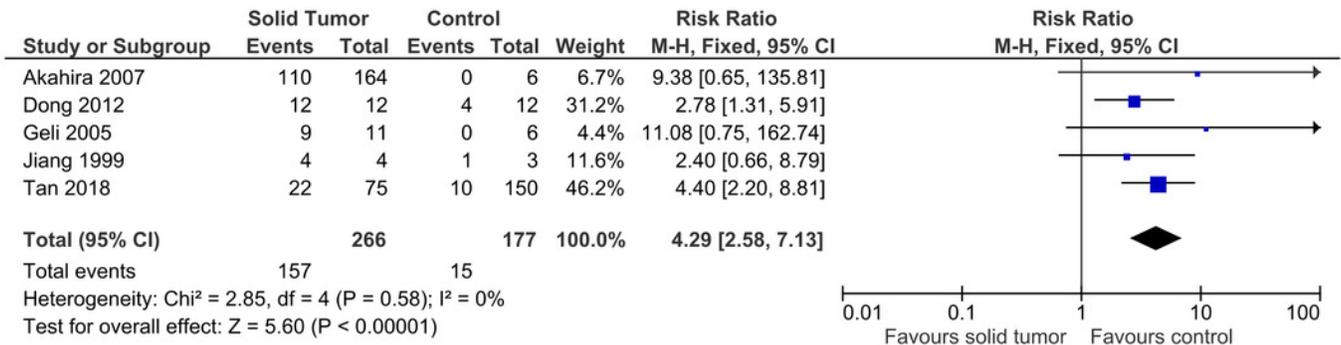


Figure 5

Sensitivity analysis to compare the use of Fixed Effects Model (FEM) with Random Effects Model (REM).

There was a significant result for both analyses: (A) FEM: RR 4.29, 95% CI 2.58-7.13, $P < 0.00001$; (B) REM: RR 3.61, 95% CI 2.28-5.72, $P < 0.00001$. FEM increased RR by 1.2 times higher with 95% CI 1.3 times wider. This sensitivity analysis proved that the results were stable. The horizontal line represents 95% CI. The blue box is the result of each individual study. The black diamond at the bottom of the plot is the pooled analysis of all studies. CI = Confidence interval. df = Degree of freedom. I^2 = Test of heterogeneity. M-H = Mantel-Haenszel.

A



B

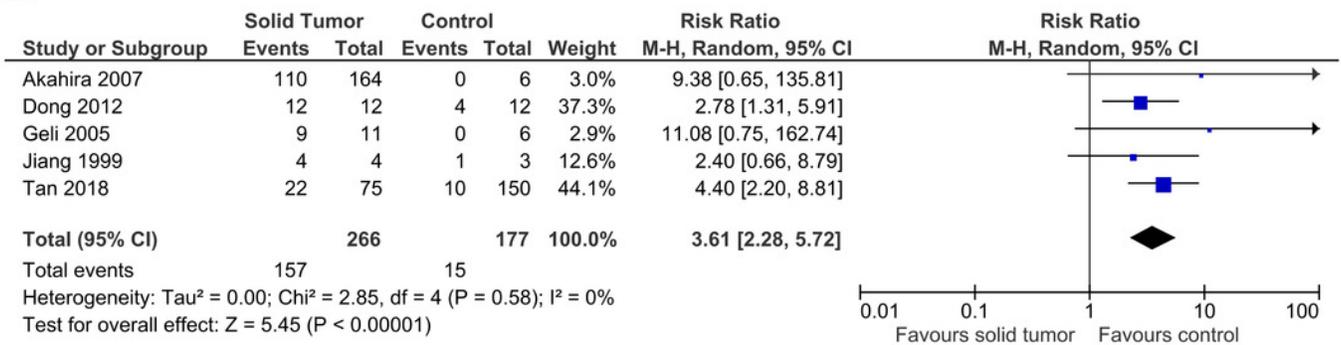


Figure 6

Forest plot for sensitivity and specificity of decreased *PRDM2* gene expression in solid tumor.

Studies that have high sensitivities include Dong et al. (2012) (Sensitivity 1.00, 95% CI 0.74-1.00) and Jiang et al. (1999) (Sensitivity 1.00, 95% CI 0.40-1.00). Studies that have high specificities are Akahira et al. (2007) (Specificity 1.00, 95% CI 0.54-1.00) and Geli et al. (2005) (Specificity 1.00, 95% CI 0.54-1.00). The horizontal line represents 95% CI. The blue box is the result of each individual study. CI = Confidence interval. FN = False negative. FP = False positive. TN = True negative. TP = True positive.

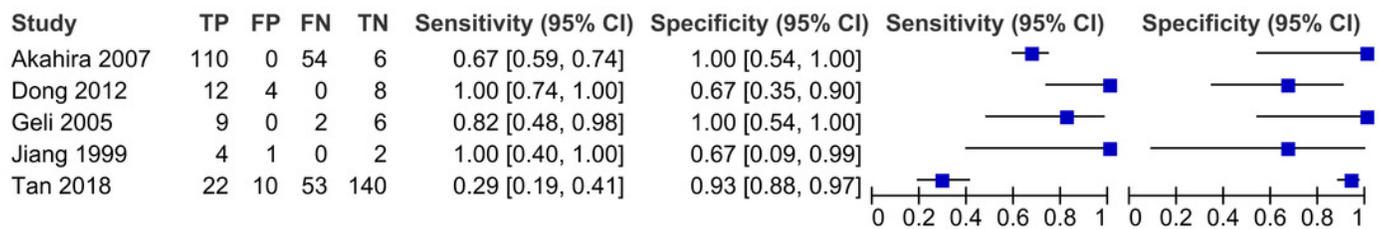


Figure 7

Summary receiver operating characteristic (SROC) curve of decreased *PRDM2* gene expression in solid tumor.

The overall sensitivity and specificity is 84% (95% CI 39-98%) and 86% (95% CI 71-94%), respectively. The calculation of these results can be viewed at Fig. S1. The black circle (summary estimate) represents the summary estimate of sensitivity and specificity. The dotted lines around the summary point represents the 95% confidence region. The dashed lines represent the 95% prediction region (the region within which we are 95% certain that the results of a new study will lie).

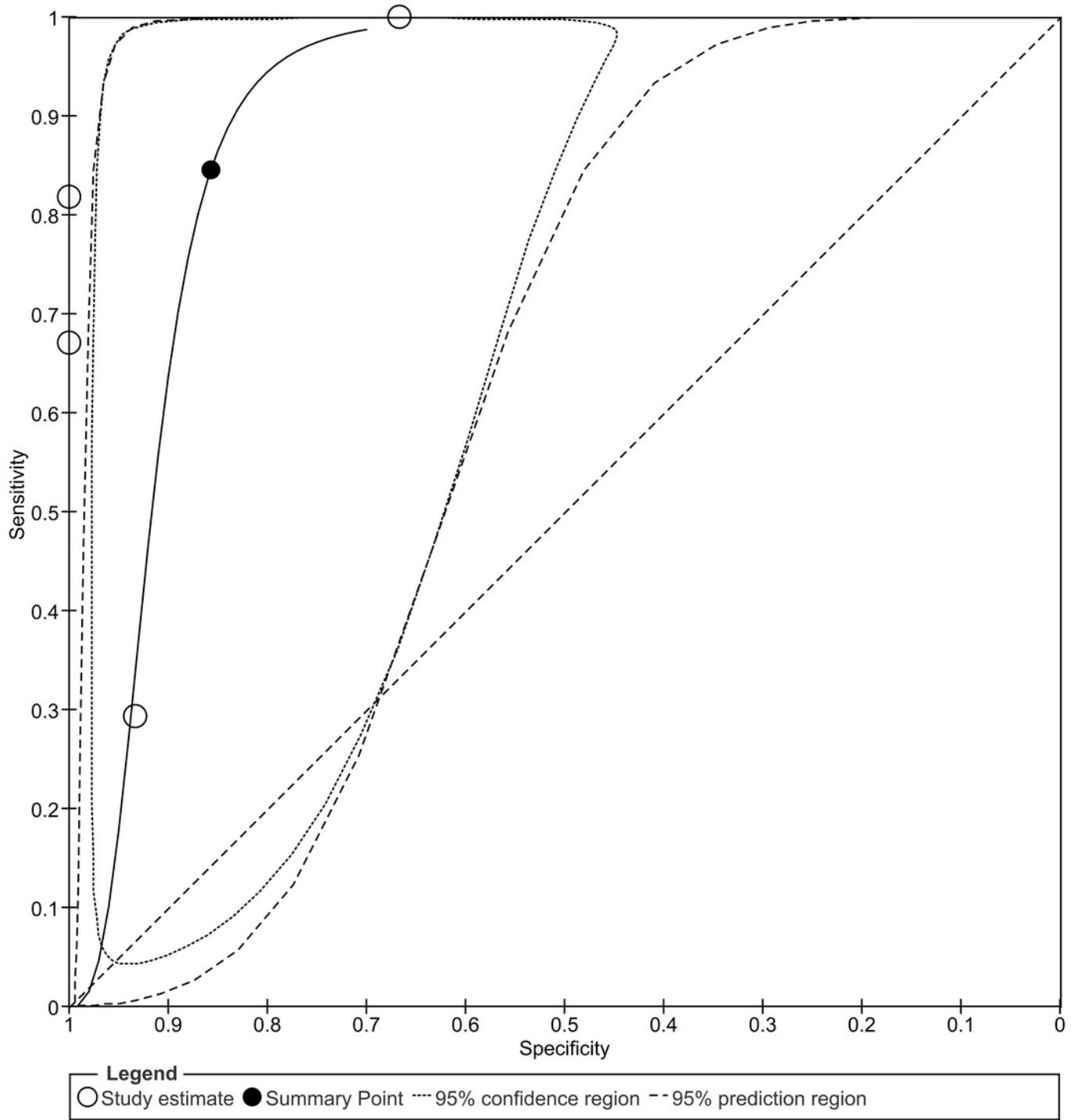


Figure 8

Trial Sequential Analysis (TSA) results of the meta-analysis.

The cumulative Z-curve (blue line) crossed the conventional meta-analysis significance boundary (horizontal red lines at $Z = + 1,96$ and $Z = - 1,96$), confirming that type I error was avoided. However, the cumulative Z-curve has not crossed the trial sequential significance boundary (diagonal red line at the top and bottom of the plot), suggesting that type II error might have not been avoided. Furthermore, the cumulative Z-score has also failed to reach the vertical red line on the right, indicating that this review has not reached the required sample size which is 7743. It is interesting to note that the cumulative Z-curve did not cross the trial sequential futility boundary (triangular red line on the right), implying that the addition of new samples could potentially improve the TSA results. In conclusion, this TSA analysis proved that this meta-analysis still requires more samples in order ensure that type II error was avoided. This is a magnified version of the TSA. The TSA results on a standard scale can be viewed at Fig. S2.

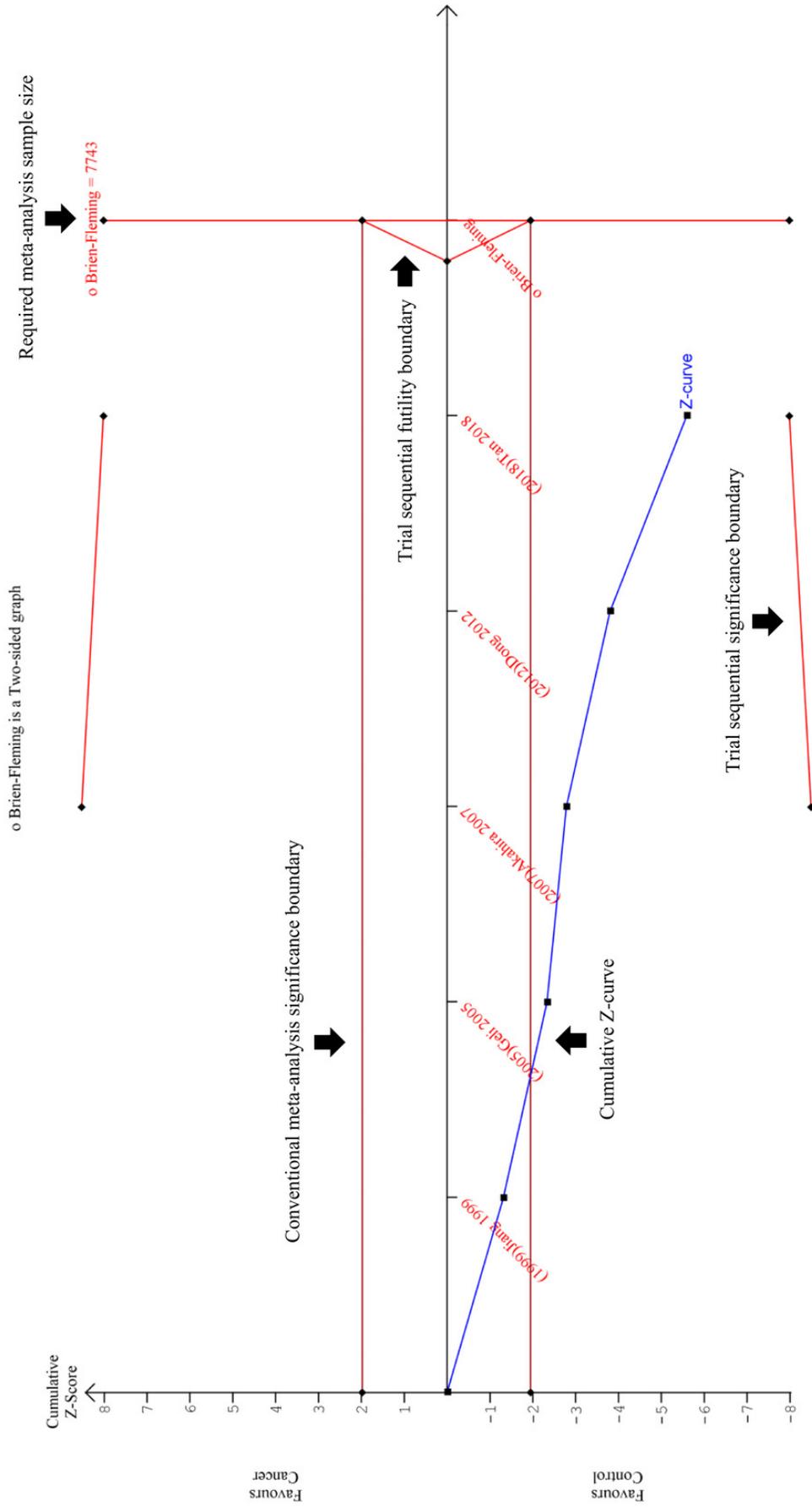


Table 1 (on next page)

Study characteristics of studies included in systematic review.

F = Female. IHC = Immunohistochemistry. LAC = Lung adenocarcinoma. LSCC = Lung squamous cell carcinoma. M = Male. ND = Not determined. PRDM2 = Positive Regulatory/Su(var)3-9, Enhancer-of-zeste and Trithorax Domain 2. qRT-PCR = Quantitative reverse transcription-polymerase chain reaction. RT-PCR = Reverse transcription-polymerase chain reaction.

Study	Country	Age	Gender	Race	Method	No. of sample	No. of control	Cancer Type	Cancer				PRDM2 expression	P value
									Stage		Differentiation			
									I+II	III+IV	Well/Moderate	Poor		
Akahira et al. (2007)	Japan	<50 (n=42/67) >=50 (n=68/97)	ND	Asian	IHC	164	6 ^a	Ovarian cancer	69	95	107 ^h	36 ^h	Decreased ⁱ	<0.05
Dong et al. (2012)	China	ND	ND	Asian	RT-PCR IHC	40	40 ^b	Esophageal squamous cell carcinoma	ND	ND	ND	ND	Decreased ⁱ	<0.05 ^j
Ge et al. (2015)	China	ND	ND	Asian	qRT-PCR	20	20 ^c	Renal cell carcinoma	ND	ND	ND	ND	Decreased ⁱ	<0.001 ^k
Geli et al. (2005)	Sweden	ND	7 M 4 F	Caucasian	qRT-PCR	11	6 ^d	Pheochromocytoma (n=4) Abdominal paraganglioma (n=7)	ND	ND	ND	ND	Decreased ⁱ	<0.001 ^l
Jiang et al. (1999)	United States of America	ND	ND	Caucasian	RT-PCR	4	3 ^e	Hepatoma	ND	ND	ND	ND	Decreased ⁱ	ND
Tan et al. (2018)	China	<60 (n=30) >=60 (n=45)	56 M 19 F	Asian	RT-PCR IHC	75	150 ^f	LSCC (n=52) LAC (n=23)	63 ^g	12 ^g	46 ^g	29 ^g	Decreased ⁱ	<0.05 ^m

^a Normal ovaries

^b Adjacent non-cancerous tissue

^c Adjacent non-malignant renal tissue

^d Normal adrenal cells

^e Normal liver tissue

^f Tumor adjacent tissue and distant lung tissue

^g Classification based on International Association for the Study of Lung Cancer 2009

^h Classification based on universal grading system for ovarian epithelial cancer

ⁱ PRDM2 expression level is decreased when compared to control

^j Chi-square test; $X^2 = 12.00$

^k Median fold difference = 0.08 (interquartile range 0.03-0.50)

^l Wilcoxon matched pair test

^m Student's t-test or one-way analysis of variance, followed by Newman-Keuls test

Table 1: Study characteristics of studies included in systematic review.

F = Female. IHC = Immunohistochemistry. LAC = Lung adenocarcinoma. LSCC = Lung squamous cell carcinoma. M = Male. ND = Not determined. PRDM2 = Positive Regulatory/Su(var)3-9, Enhancer-of-zeste and Trithorax Domain 2. qRT-PCR = Quantitative reverse transcription-polymerase chain reaction. RT-PCR = Reverse transcription-polymerase chain reaction.

Table 2 (on next page)

Study characteristics of studies included in meta-analysis.

FN = False negative. FP = False positive. IHC = Immunohistochemistry. LAC = Lung adenocarcinoma. LSCC = Lung squamous cell carcinoma. qRT-PCR = Quantitative reverse transcription-polymerase chain reaction. TN = True negative. TP = True positive.

Study	Method	No. of sample	No. of control	Cancer type	TP	FP	FN	TN
Akahira et al. (2007)	IHC	164	6	Ovarian cancer	110	0	54	6
Dong et al. (2012)	IHC	12	12	Esophageal squamous cell carcinoma	12	4	0	8
Geli et al. (2005)	qRT-PCR	11	6	Pheochromocytoma (n=4) Abdominal paraganglioma (n=7)	9	0	2	6
Jiang et al. (1999)	qRT-PCR	4	3	Hepatoma	4	1	0	2
Tan et al. (2018)	IHC	75	150	LSCC (n=52) LAC (n=23)	22	10	53	140

Table 2: Study characteristics of studies included in meta-analysis.

FN = False negative. FP = False positive. IHC = Immunohistochemistry. LAC = Lung adenocarcinoma. LSCC = Lung squamous cell carcinoma. qRT-PCR = Quantitative reverse transcription-polymerase chain reaction. TN = True negative. TP = True positive.

Table 3 (on next page)

Quality Assessment of Diagnostic Accuracy Studies – 2 (QUADAS-2) risk of bias assessment.

Table 3: Quality Assessment of Diagnostic Accuracy Studies – 2 (QUADAS-2) risk of bias assessment.

Quality Assessment of Diagnostic Accuracy Studies - 2 (QUADAS-2)							
Study	Risk of bias				Applicability concerns		
	Patient selection	Index test	Reference standard	Flow and timing	Patient selection	Index test	Reference standard
Akahira et al. (2007)	Low	Low	Low	Low	Low	Low	Low
Dong et al. (2012)	Low	Low	Low	Low	Low	Low	Low
Ge et al. (2015)	Low	Unclear	Low	Low	Low	Low	Low
Geli et al. (2005)	Low	Unclear	Low	Low	Low	Low	Low
Jiang et al. (1999)	Unclear	Unclear	Low	Low	Low	Low	Low
Tan et al. (2018)	Low	Unclear	Low	Low	Low	Low	Low

Table 4(on next page)

Grading of Recommendations, Assessment, Development, and Evaluations (GRADE) evidence profile for the studies included in the meta-analysis.

ND = Not determined. QUADAS-2 = Quality Assessment of Diagnostic Accuracy Studies - 2.

TSA = Trial sequential analysis.

Table 4: Grading of Recommendations, Assessment, Development, and Evaluations (GRADE) evidence profile for the studies included in the meta-analysis.

Grading of Recommendations, Assessment, Development, and Evaluations (GRADE)								
Outcome	No. of studies	Design	Risk of bias	Indirectness	Inconsistency	Imprecision	Publication bias	Quality of evidence
True positives (patients/samples with solid tumor)	5 studies (443 patients/samples)	Cross-sectional studies	Not serious	Serious ¹	Not serious	Serious ²	ND ³	⊕○○○ Very low
False negatives (patients/samples incorrectly classified as not having solid tumor)	5 studies (443 patients/samples)	Cross-sectional studies	Not serious	Serious ¹	Not serious	Serious ²	ND ³	⊕○○○ Very low
True negatives (patients/samples without solid tumor)	5 studies (443 patients/samples)	Cross-sectional studies	Not serious	Serious ¹	Not serious	Serious ²	ND ³	⊕○○○ Very low
False positives (patients/samples incorrectly classified as having solid tumor)	5 studies (443 patients/samples)	Cross-sectional studies	Not serious	Serious ¹	Not serious	Serious ²	ND ³	⊕○○○ Very low

¹ All samples undergo both index test and reference standard, introducing indirectness into the studies.

² Most of the individual studies have a wide confidence interval and inconclusive TSA results.

³ Publication bias could not be evaluated as the number of studies is <10.

ND = Not determined. QUADAS-2 = Quality Assessment of Diagnostic Accuracy Studies – 2. TSA = Trial sequential analysis.