

Accuracy of Raman spectroscopy for differentiating skin cancer from normal tissue

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Abstract

Background: Raman spectroscopy could be applied to distinguish tumor from normal tissues. This meta-analysis assessed the accuracy of Raman spectroscopy in differentiating skin cancer from normal tissue.

Methods: PubMed, Embase, Cochrane Library, and CNKI were searched to identify suitable studies before February 4th, 2018. We estimated the pooled sensitivity, specificity, positive, and negative likelihood ratios, diagnostic odds ratio, and constructed summary receiver-operating characteristics curves to identify the accuracy of Raman spectroscopy in differentiating skin cancer from normal tissue.

Results: A total of 12 studies with 2461 spectra were included. For basal cell skin cancer (BCC) ex vivo detection, the pooled sensitivity and specificity were 0.99 (95% confidence interval [CI] 0.97–0.99) and 0.96 (95% CI 0.95–0.97), respectively. The area under the curve (AUC) was 0.9837. For BCC in vivo detection, the pooled sensitivity and specificity were 0.69 (95% CI 0.61–0.76) and 0.85 (95% CI 0.82–0.87), respectively. The AUC was 0.9213. For melanoma (MM) ex vivo detection, the pooled sensitivity and specificity were 1.00 (95% CI 0.91–1.00) and 0.98 (95% CI 0.95–1.00), respectively. The AUC was 0.9914. For MM in vivo detection, the sensitivity (0.93) and the specificity (0.96) balanced relatively well. For squamous cell skin cancer (SCC) ex vivo detection, the pooled sensitivity and specificity were 0.96 (95% CI 0.81–1.00) and 1.00 (95% CI 0.92–1.00), respectively. For SCC in vivo detection, the sensitivity was 0.81 (95% CI 0.70–0.90) and the specificity was 0.89 (95% CI 0.86–0.91).

Conclusion: This meta-analysis suggested that Raman spectroscopy could be an effective and accurate tool for differentiating BCC, MM, SCC from normal tissue, which would assist us in the diagnosis and treatment of skin cancer.

Abbreviations: AUC = area under curve, BCC = basal cell cancer, CI = confidence interval, DA = diagnostic algorithm, DOR = diagnostic odds ratio, MM = melanoma, MRDF-SMLR = maximum representation and discrimination and discriminant algorithms using sparse multinomial logistic regression, NLR = negative likelihood ratio, NNA = Neural network analysis, PCA = Principal components analysis, PLR = positive likelihood ratio, PLS = partial least squares, SCC = squamous cell cancer, TA = texture analysis.

Keywords: basal cell cancer, melanoma, Raman spectroscopy, skin cancer, squamous cell cancer

1. Introduction

Skin cancer is the most common form of cancer, globally accounting for at least 40% of cases.^[1,2] There are 3 main types of skin cancers: basal cell skin cancer (BCC), squamous cell skin cancer (SCC), and melanoma (MM). The first 2, along with a number of less common skin cancers, are known as non-melanoma skin cancer (NMSC).^[3] The World Health Organization (WHO) estimated 2 to 3 million NMSC and 132,000 melanoma skin cancers occur globally each year.^[4]

Early diagnosis and treatment are recommended for the management of skin cancer. The current “criterion standard” for diagnosis is based on clinical examination followed by biopsy and histopathology, which is invasive, costly, and time-consuming.^[5] For the treatment of skin cancer, surgical removal is the optimal choice. However, the challenge for complete removal is to differentiate between normal skin and the cancer. Histopathology can hardly provide surgeons a precise margin of the tumor even with an increased number of biopsies. Meanwhile, more biopsies means increased financial burden and associated discomfort from the additional biopsy procedures. Dermoscope is a noninvasive in situ diagnostic tool, which is based on visual inspection and recognition of morphologic characteristics.^[6] The use of it can improve the accuracy of MM diagnosis but requires well-trained skills and rich experience.^[7] Thus, we need an accurate and objective technique with high efficiency to assist us in diagnosis and treatment of skin cancer.

Recent studies reported Raman spectroscopy (RS) has the potential to diagnose and study the evolution of human malignancies both in vitro and in vivo in esophagus,^[8] stomach,^[9] lung,^[10] breast,^[11] prostate^[12] arteries,^[13] and others. Raman spectroscopy is an optical technique, which uses the inelastic scattering of monochromatic light to analyze vibrational modes of molecules.^[14] Tumor tissue and normal tissue have different compositions because of the changes in the molecular structures of proteins, lipids, and pigments. Raman

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spectroscopy is able to detect the differences, and therefore, it has been considered as a promising tool for cancer diagnosis. Among other noninvasive optical techniques such as optical coherence tomography (OCT),^[15,16] confocal laser scanning microscopy (CLSM)^[17] or multiphoton tomography, Raman spectroscopy is molecular-specific and objective. Besides, its rapidity in examination and analysis allows real-time diagnosis. All these characteristics make RS a complementary tool in incipient lesion differentiation and intraoperative tumor margin assessment where histopathology is relatively impractical.

Raman spectroscopy has gained some clinical acceptance in the diagnosis of skin cancer.^[18–29] However, these studies were inconclusive because of insufficient sample and different diagnostic algorithms. The aim of this meta-analysis was to systematically evaluate the accuracy of RS for discriminating normal tissue and skin cancer tissue.

2. Methods

2.1. Search strategy

As this is a meta-analysis, ethical approval was not necessary. We followed the guidelines for the systematic review and the meta-analysis of diagnostic studies.^[30] Then we searched four databases, including PubMed, Embase, Cochrane Library, and CNKI, for the studies on February 4th, 2018, and no start date limit was applied. The search terms were “Raman” and “skin cancer.” No language restriction is exposed. Reference lists of relevant articles were also searched. Two reviewers independently reviewed the articles. Disagreements were resolved by consensus.

2.2. Study selection criteria

The studies were selected on the basis of the following criteria: only human tissue used in the experiments; Raman spectroscopy was used as a diagnosis tool to distinguish tumor and normal tissues; used histopathology as criterion standard; provided with detailed data to construct a 2×2 contingency table for true-positive (TP), false-positive (FP), true-negative (TN), and false-negative (FN). If the 4 values were not reported, we calculated backwards using indexes including sensitivity, specificity, positive predictive value (PPV), and negative predictive value (NPV). Corresponding authors were contacted for the detailed data if no enough data was available.

Excluded criteria: unrelated articles, abstracts presented at academic conferences; included <10 spectra samples; without sufficient calculable data; duplicated reports, or studies based on the same study.

2.3. Data extraction

Two investigators extracted the data independently and disagreements were resolved by consensus. First author, year of publication, country, sample size, tumor type, methodological and technical data, numbers of TP, FP, TN, and FN were extracted from each study.

2.4. Quality assessment

The quality of each study was assessed by using a checklist based on the Quality Assessment of Diagnostic Accuracy Studies (QUADAS) guidelines, which is an established, evidence-based tool for systematic reviews of diagnostic studies designed for diagnostic accuracy.

2.5. Statistical method

Using the extracted data of TP, TN, FP, and FN, the pooled sensitivity, specificity, positive and negative likelihood ratios (LRs), and diagnostic odds ratio (DOR), with 95% confidence intervals (CI), were calculated based on bivariate generalized linear mixed modeling. Meta-Disc version 1.4 statistical software was used.

Furthermore, summary receiver operator characteristics (SROC) curves were constructed to examine the relationship between sensitivity and specificity. And the area under the curve (AUC) was calculated to assess the overall performance of Raman spectroscopy. In general, a diagnostic tool is regarded excellent when AUC values were between 0.9 and 1, good when AUC values were between 0.8 and 0.9, fair when AUC values were between 0.7 and 0.8, poor when AUC values were between 0.6 and 0.7, and failed when AUC values were between 0.5 and 0.6.^[31] The SROC curves were also performed by Meta-Disc version 1.4.

2.6. Publication bias

Publication bias was assessed using Deeks funnel plot asymmetry test ($P < .05$ was considered that potential publication bias exists). The Deeks funnel plot asymmetry test was performed by Stata 11.0.

3. Results

3.1. Literature research

The initial literature search yielded 140 articles, which were then reviewed in title and abstract. Of these, 31 articles were further reviewed in full text. Then, 19 articles failed to satisfy the inclusion criterion: 5 were not relative, 8 were repeated reports, 3 did not use normal tissue as control group, and 3 had insufficient details to reconstruct the 2×2 table. So, 12 studies were included in this meta-analysis.^[18–29] The study selection process is shown in Figure 1.

3.2. Study characteristics

The detailed characteristics of the 12 studies were given in Table 1. These studies were operated in 6 different countries. All the articles were published between 2003 and 2015, and >75% were after 2008. The number of tissues involved in each study varied from 8 to 223. The number of the spectra retrieved varies from 13 to 505. The total number of spectra was 2461, with an average of 273.

In these 12 eligible studies, BCC, SCC and MM were addressed in 10 studies, 4 studies, and 4 studies, respectively. Raman spectroscopy was applied both in vivo (4 studies) and ex vivo (8 studies). The analysis of the Raman spectra was performed with various diagnostic algorithms, including PCA (Principal components analysis), PLS (Partial Least Squares), MRDF-SMLR (maximum representation and discrimination and discriminant algorithms using sparse multinomial logistic regression), TA (Texture analysis), and NNA (Neural network analysis).

3.3. Pooled results

3.3.1. BCC/ex vivo group. Seven studies^[18,20–23,25,26] examined BCC samples ex vivo. The pooled sensitivity and specificity of Raman spectroscopy for discriminating BCC samples and normal tissues ex vivo were 0.99 (95% CI 0.97–0.99) and 0.96 (95% CI

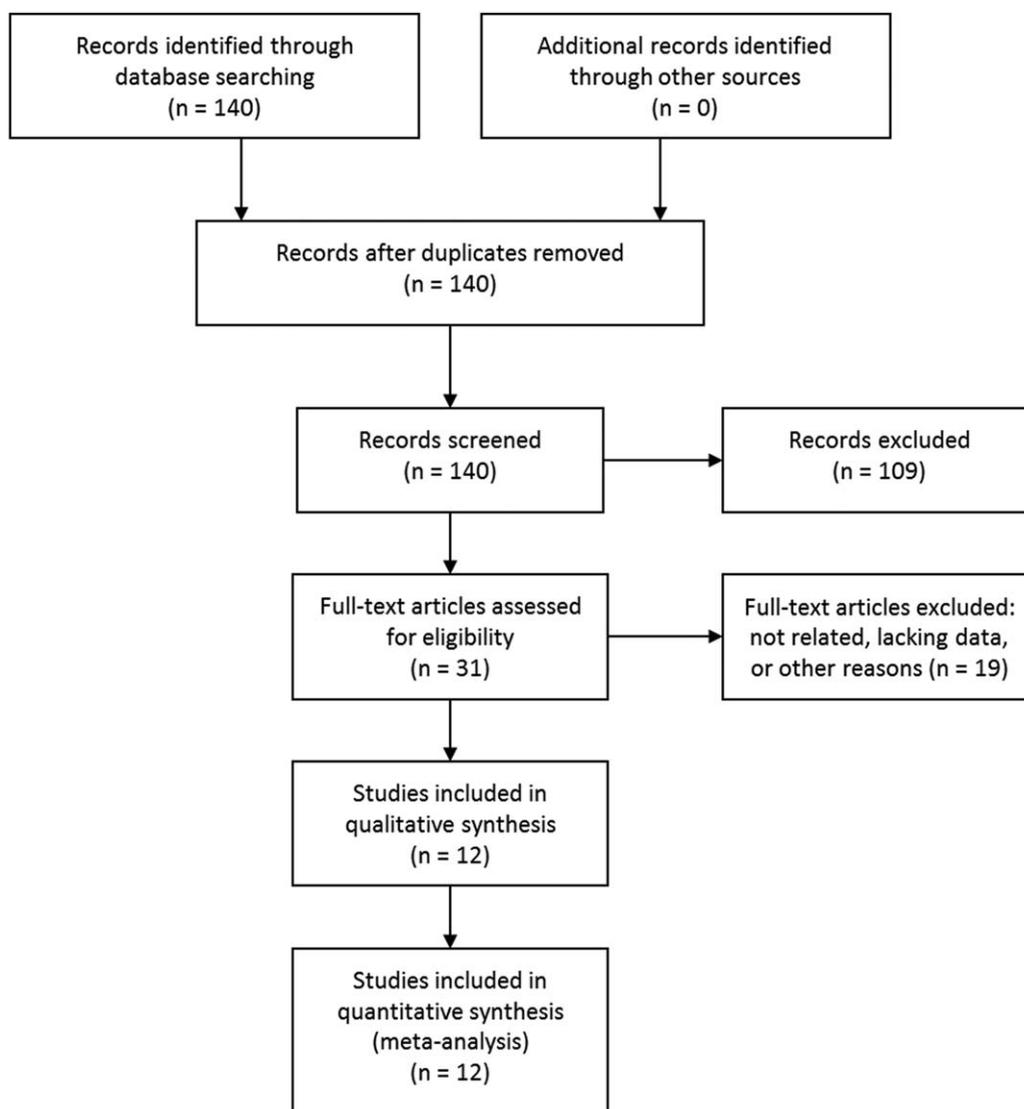


Figure 1. Literature search and selection.

0.95–0.97), respectively. The plots were shown in Figure 2. The pooled PLR and NLR were 26.40 (95% CI 11.22–62.13) and 0.03 (95% CI 0.01–0.09), respectively. The DOR was 938.93 (95% CI 187.85–4693.05), demonstrating high accuracy. The SROC curve analysis was used to summarize overall diagnostic accuracy. The AUC was 0.9837.

3.3.2. BCC/in vivo group. Four studies^[24,27–29] examined BCC samples in vivo. The pooled sensitivity and specificity of Raman spectroscopy for discriminating BCC samples and normal tissues in vivo were 0.69 (95% CI 0.61–0.76) and 0.85 (95% CI 0.82–0.87), respectively. The plots were also shown in Figure 2. The pooled PLR and NLR were 6.20 (95% CI 3.05–12.57) and 0.32 (95% CI 0.19–0.55), respectively. The DOR was 44.85 (95% CI 7.63–263.43), also demonstrating very high accuracy. The SROC curve was also performed to summarize overall diagnostic accuracy. The AUC was 0.9213.

3.3.3. MM/ex vivo group. Three studies^[18,20,23] examined MM samples ex vivo. The pooled sensitivity and specificity of Raman spectroscopy for discriminating MM samples and normal tissues

ex vivo were 1.00 (95% CI 0.91–1.00) and 0.98 (95% CI 0.95–1.00), respectively. The plots were shown in Figure 3. The pooled PLR and NLR were 29.21 (95% CI 5.40–157.90) and 0.04 (95% CI 0.01–0.19), respectively. The DOR was 837.77 (95% CI 104.1–6742.41), demonstrating high accuracy. The SROC curve analysis was used to summarize overall diagnostic accuracy. The AUC was 0.9914.

3.3.4. MM/in vivo group. Only one study^[27] examined MM samples in vivo for the discriminating role of the Raman spectroscopy. The sensitivity (0.93) and the specificity (0.96) balanced relatively well. The DOR was 371.

3.3.5. SCC/ex vivo group. Two studies^[19,23] examined SCC samples ex vivo. The pooled sensitivity and specificity of Raman spectroscopy for discriminating SCC samples and normal tissues in vivo were 0.96 (95% CI 0.81–1.00) and 1.00 (95% CI 0.92–1.00), respectively. The plots were shown in Figure 4. The pooled PLR and NLR were 42.59 (95% CI 6.12–296.61) and 0.07 (95% CI 0.02–0.27), respectively. The DOR was 646.37 (95% CI 51.68–8084.81).

Table 1
Baseline characteristics of included studies.

Author	Year	N1*	N2†	N3‡	Country	Disease	vs	Age	Sample	DA
Nunes et al ^[26]	2003	NR	8	13	Brazil	BCC	Normal	NR	Ex vivo	PCA
Bodanese et al ^[18]	2012	NR	47	145	Brazil	BCC	Normal	NR	Ex vivo	PCA and ED
						MM	Normal		Ex vivo	PCA and ED
Silveira et al ^[29]	2015	25	49	484	Brazil	BCC	Normal	mean 61	In vivo	PLS
Nijssen et al ^[25]	2007	17	28	504	Netherlands	BCC	Normal	NR	Ex vivo	PCA and LDA
Liber	2008	39	39	NR	USA	BCC	Normal	NR	Ex vivo	MRDF/SMLR
						SCC	Normal		Ex vivo	MRDF/SMLR
						MM	Normal		Ex vivo	MRDF/SMLR
Lieber-2	2008	19	21	42	USA	BCC	Normal	NR	In vivo	MRDF/SMLR
						SCC	Normal		In vivo	MRDF/SMLR
Fox et al ^[19]	2014	11	25	150	USA	SCC	Normal	NR	Ex vivo	PCA
Gniadecka et al ^[20]	2004	NR	223	250 scans per sample	Denmark	MM	Normal	NR	Ex vivo	NNA
						BCC	Normal		Ex vivo	NNA
Philipsen et al ^[27]	2013	127	136	246	Denmark	BCC	Normal	69	In vivo	PCA
						MM	Normal	58	In vivo	PCA
Schleusener et al ^[28]	2015	104	137	385	Germany	BCC	Normal	NR	In vivo	PLS
						SCC	Normal		In vivo	PLS
Legesse et al ^[22]	2015	NR	/	/	Germany	BCC	Normal	NR	Ex vivo	TA
Kong et al ^[21]	2013	55	/	492	UK	BCC	Normal	NR	Ex vivo	

BCC=basal cell cancer, DA=diagnostic algorithm, MM=melanoma, MRDF-SMLR=maximum representation and discrimination and discriminant algorithms using sparse multinomial logistic regression, NNA=neural network analysis, NR=not reported, PCA=principal components analysis, PLS=partial least squares, SCC=squamous cell cancer, TA=texture analysis.

* Number of patients.
† Number of samples.
‡ Number of spectra.

3.3.6. SCC/in vivo group. SCC samples were examined in vivo in 2 studies,^[24,28] as shown in Table 2. The sensitivity of Raman spectroscopy for discriminating SCC samples and normal tissues in vivo was 0.81 (95% CI 0.70–0.90) and the specificity was 0.89 (95% CI 0.86–0.91), whereas the PLR and NLR were 7.29 (95% CI 5.69–9.33) and 0.22 (95% CI 0.13–0.35), respectively. The DOR was 33.81 (95% CI 17.78–64.31). The plots were shown in Figure 4.

3.4. Assessment of study quality

Two reviewers evaluated methodological quality for each study according to the QUADAS guidelines independently. Of the 12 studies, one study had a total quality score of 14 (100% rate of Y), 10 studies had a score of 13 (92.9% rate of Y), and the last 1 study had a score of 12 (85.7% rate of Y). Table 3 shows the results of the evaluation of each study.

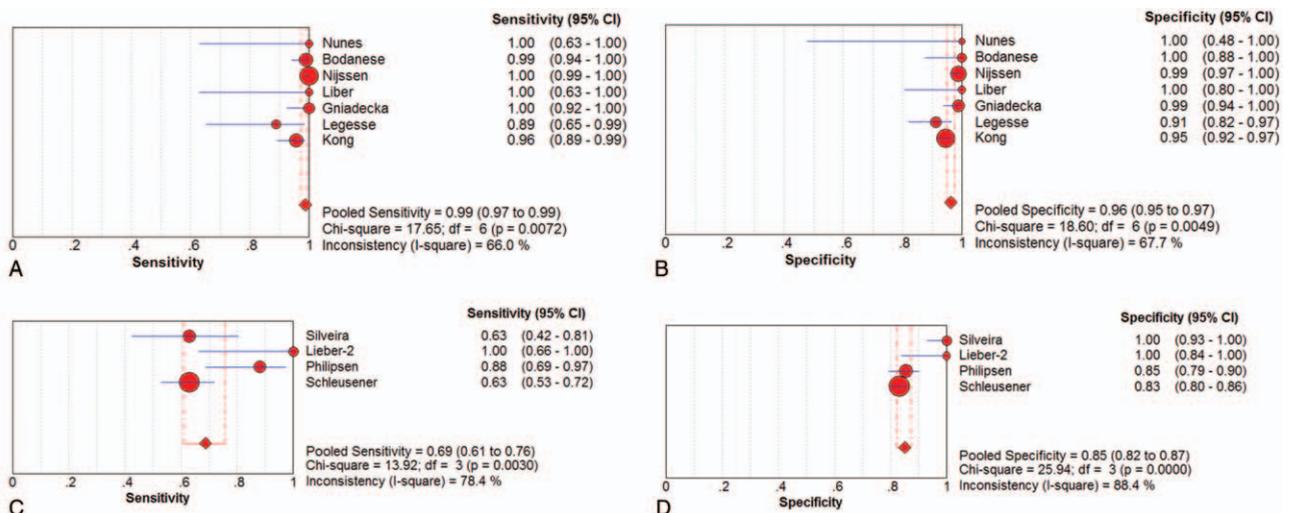


Figure 2. Individual study and pooled estimates of sensitivity and specificity and their 95% confidence intervals (CIs) of Raman spectroscopy to differentiate basal cell cancer from normal tissues ex vivo (A, B) and in vivo (C, D).

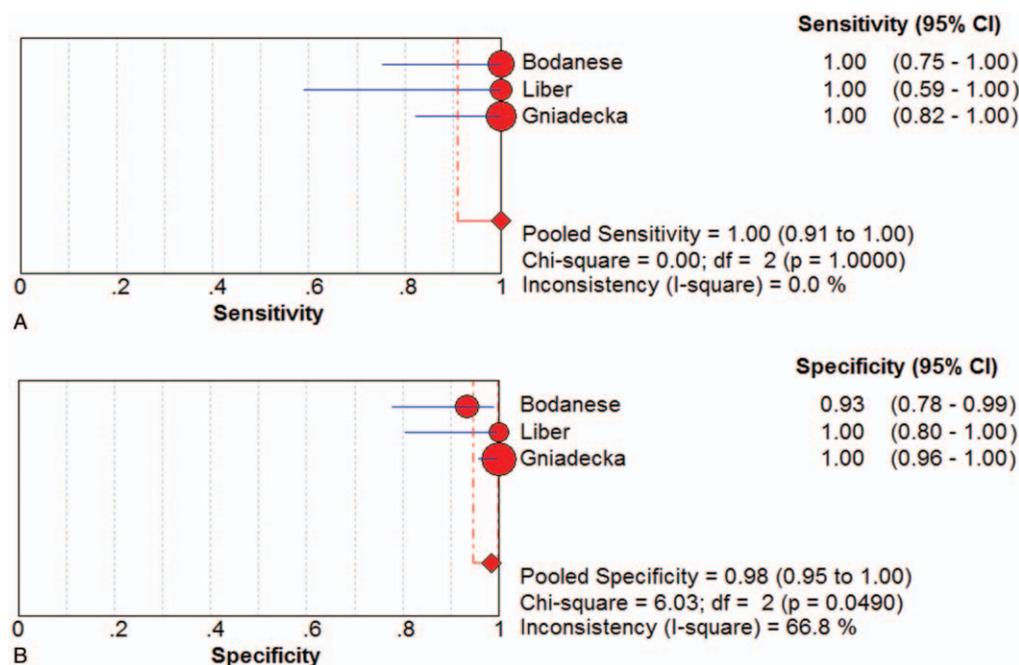


Figure 3. Individual study and pooled estimates of sensitivity (A) and specificity (B) and their 95% confidence intervals (CIs) of Raman spectroscopy to differentiate melanoma from normal tissues ex vivo.

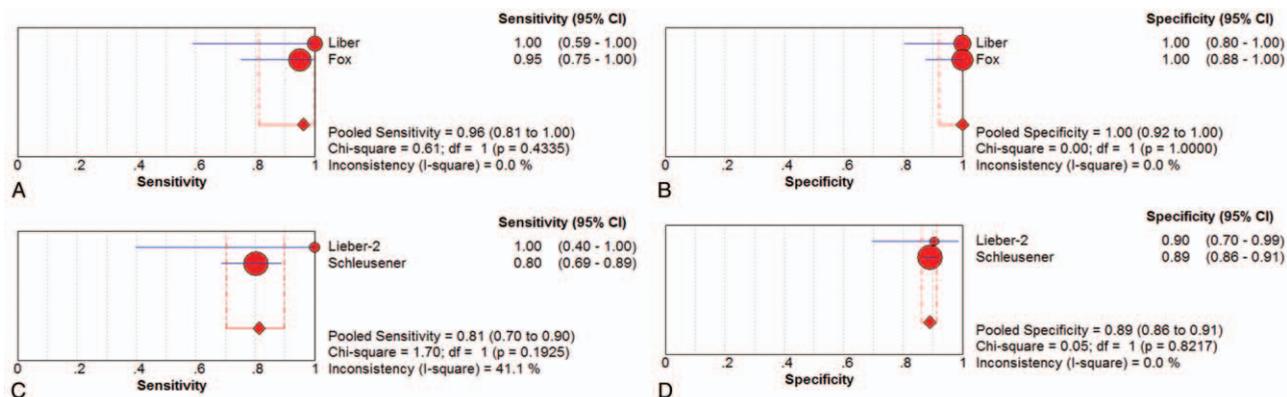


Figure 4. Individual study and pooled estimates of sensitivity and specificity and their 95% confidence intervals (CIs) of Raman spectroscopy to differentiate squamous cell cancer from normal tissues ex vivo (A, B) and in vivo (C, D).

3.5. Publication bias

No publication bias was found in this meta-analysis (*P* values shown in Table 2).

4. Discussion

4.1. Implications

In this meta-analysis, we assessed the accuracy of Raman spectroscopy in differentiating skin cancer from normal tissue. For BCC ex vivo samples, the pooled sensitivity and specificity of RS were 0.99 and 0.96, respectively. The AUC was 0.9837. And those for BCC in vivo samples were 0.69 (sensitivity), 0.85 (specificity), and 0.9213 (AUC). For BCC, RS gave a better

performance ex vivo comparing to in vivo detection. This can be found in MM and SCC as well. The most important reason might be that in vivo detection could provide less information about the lesions owing to the limited time to collect Raman images.^[19] Ex vivo Raman images by scanning across the surface of the lesions could provide the spatial distribution of different tissue structures. However, it usually takes a long period of time. It would not be feasible for in vivo use, as it is hard for the patients to keep highly immobilized to obtain high spatial-resolution images.^[18,19] There is a dearth of studies focused on in vivo detection, whereas many reports have demonstrated various ex vivo detection.^[18] Thus, continued patient recruitment and future development of in vivo RS will be necessary to elucidate this matter.^[18]

Table 2

Pooled estimations of sensitivity, specificity, positive likelihood ratio, negative likelihood ratio, diagnostic odd ratio, and area under curve of Raman spectroscopy to differentiate skin cancer from normal tissues.

Disease	Sample	N1*	N2†	SEN (95% CI)	SPE (95% CI)	PLR (95% CI)	NLR (95% CI)	DOR (95% CI)	AUC	P‡
BCC	Ex vivo	7	1376	0.99 (0.97–0.99)	0.96 (0.95–0.97)	26.40 (11.22–62.13)	00.03 (0.01–0.09)	938.93 (187.85–4693.05)	0.9837	1.000
BCC	In vivo	4	1029	0.69 (0.61–0.76)	0.85 (0.82–0.87)	6.20 (3.05–12.57)	0.32 (0.19–0.55)	44.85 (7.63–263.43)	0.9213	.734
MM	Ex vivo	3	171	1.00 (0.91–1.00)	0.98 (0.95–1.00)	29.21 (5.40–157.90)	0.04 (0.01–0.19)	837.77 (104.1–6742.41)	0.9914	1.000
MM	In vivo	1	180	0.93	0.96	25.67	0.07	371	—	—
SCC	Ex vivo	2	72	0.96 (0.81–1.00)	1.00 (0.92–1.00)	42.59 (6.12–296.61)	0.07 (0.02–0.27)	646.37 (51.68–8084.81)	—	1.000
SCC	In vivo	2	715	0.81 (0.70–0.90)	0.89 (0.86–0.91)	7.29 (5.69–9.33)	0.22 (0.13–0.35)	33.81 (17.78–64.31)	—	1.000

AUC=area under curve, BCC=basal cell cancer, CI=confidence interval, DOR=diagnostic odds ratio, MM=melanoma, NLP=negative likelihood ratio, PLR=positive likelihood ratio, SCC=squamous cell cancer, SEN=sensitivity, SPE=specificity.

* Number of study.

† Number of samples.

‡ P value of publication bias.

Table 3

Quality assessment of included studies using QUADAS questionnaire.

Author	Year	Q1	Q2	Q3	Q4	Q5	Q6	Q7	Q8	Q9	Q10	Q11	Q12	Q13	Q14	Rate of Y (%)
Bodanese et al ^[18]	2012	Y	Y	Y	Y	Y	Y	Y	Y	Y	NR	Y	Y	Y	Y	92.86%
Fox et al ^[19]	2014	Y	Y	Y	Y	Y	Y	Y	Y	Y	NR	Y	Y	Y	Y	92.86%
Gniadecka et al ^[20]	2004	Y	N	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	Y	85.71%
Kong et al ^[21]	2013	Y	Y	Y	Y	Y	Y	Y	Y	Y	NR	Y	Y	Y	Y	92.86%
Legesse et al ^[22]	2015	Y	Y	Y	Y	Y	Y	Y	Y	Y	NR	Y	Y	Y	Y	92.86%
Liber	2008	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	Y	92.86%
Lieber-2	2008	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	92.86%
Nijssen et al ^[25]	2007	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	92.86%
Nunes et al ^[26]	2003	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	100.00%
Phillipsen et al ^[27]	2013	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	92.86%
Schleusener et al ^[28]	2015	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	Y	92.86%
Silveira et al ^[29]	2015	Y	Y	Y	Y	Y	Y	Y	Y	Y	N	Y	Y	Y	Y	92.86%

N=no, NR=not reported, QUADAS=Quality assessment of diagnostic accuracy studied, Y=yes.

Q1. Was the spectrum of patients representative of the patients who will receive the test in practice? Q2. Were selection criteria clearly described? Q3. Is the reference standard likely to correctly classify the target condition? Q4. Is the time period between reference standard and index test short enough to be reasonable? Q5. Did the whole sample, or a random selection of the sample, receive verification using a reference standard of diagnosis? Q6. Did patients receive the same reference standard regardless of the index test result? Q7. Was the reference standard independent of the index test (i.e., the index test did not form part of the reference standard)? Q8. Was the execution of the index test described in sufficient detail to permit replication of the test? Q9. Was the execution of the reference standard described in sufficient detail to permit its replication? Q10. Were the index test results interpreted without knowledge of the results of the reference test? Q11. Were the reference standard results interpreted without knowledge of the results of the index test? Q12. Were the same clinical data available when test results were interpreted as would be available when the test is used in practice? Q13. Were interpretable/intermediate test results reported? Q14. Were withdrawals from the study explained?

As we can see in Table 2, the accuracy of RS to differentiate between MM and normal tissue was higher than that of BCC and SCC. It is like because of the differences in the cellular origins of the cancers, as both BCC and SCC involve malignancy of keratinized epidermal cells and melanomas result from malignancy of melanocytes.^[32–34] Because of the differences in composition, the spectral differences in the melanoma spectra are seen to be much more significant and at different wave number ranges than the BCC and SCC spectra, whereas the BCC and SCC spectra show significant differences in similar wave number ranges.^[35] That may be the reason why the accuracy was different between MM and other 2 types of skin cancer.

According to the results, we can conclude that RS is a viable candidate for differentiating skin cancer from normal skin tissue. This research for the first time summarized the evidence on the accuracy of Raman spectroscopy in the detection of BCC, MM, and SCC both in vivo and ex vivo.

Raman spectroscopy is a promising diagnostic tool with several advantages. First, RS is easy to perform and requires no special staining or preparation.^[19] This makes real-time diagnosis possible and be able to avoid surgical workflow disruption. Second, it only

takes a few minutes to obtain an accurate diagnostic result with RS, whereas traditional analytic technique requires hours or days.^[36] Third, RS is a noninvasive technique and does no harm to the patients. Besides, RS is molecular-specific, and therefore objective.^[37] That is why RS is able to differentiate incipient lesions. Also, its high accuracy helps decrease the number of expensive tests needed to guarantee the correct diagnosis.^[38]

With these characteristics, RS can be used as intraoperative guidance in skin cancer excision. By providing clear tumor margin, RS contributes to minimize the volume of residual tumor and avoid excessive removal of normal tissue. Besides, Raman-guided biopsy allows more accurate biopsy and reduces the incidence of repeated stereotactic biopsy procedure when no representative tumor tissue is found for the first time.^[39] With its ability in differentiating incipient lesions, RS can be used in early screening test for skin cancer.^[40] In radiation therapy, Raman technique can also find its value.^[41]

To achieve extensive application of Raman spectroscopy, several factors need to be taken into consideration, including cost, maintenance, personnel training, analysis of the data, and time of investigation. RS requires periodic calibration and routine

maintenance, which may increase its cost.^[42] Although using RS to get data from the sample is easy to perform, the analysis of these data requires exquisite skills. That is why the training for qualified algorithm designers is so important.^[43]

4.2. Limitations

This study also had several limitations. First, this meta-analysis is based on a limited number of studies. Although the number of spectra involved in is large (2461 spectra), more studies are needed. Second, the patient size in each study was small and the numbers of spectra differed sharply among the included studies and this variability might have affected the outcome. Third, two-thirds of the studies used *ex vivo* tissue. To prove whether Raman spectroscopy is an optimal diagnostic tool or not, more studies involving *in vivo* technique are needed. Furthermore, different techniques of Raman spectroscopy, and multiple algorithms were used in the included studies. Finally, the publication bias was a major concern for all meta-analysis. In our meta-analysis, although no publication bias was found ($P > .05$), it should be noted that any meta-analysis could not completely exclude biases. Therefore, more studies with more patients examined *in vivo* are needed.

4.3. Future research

In conclusion, our study suggested that Raman spectroscopy could be an effective and accurate tool for differentiating BCC, MM, SCC from normal tissue. The application of this promising novel method would improve the accuracy of skin cancer diagnosis and surgical removal in the future, by both avoiding removal of normal tissue and minimizing the volume of residual tumor. However, more studies are warranted to verify that and more efforts are still needed to improve this equipment and better serve clinical work.

Author contributions

Conceptualization: Jing Zhang, Jianguo Xu.

Data curation: Jing Zhang, Yimeng Fan, Yanlin Song.

Formal analysis: Jing Zhang, Yimeng Fan, Jianguo Xu.

Investigation: Jing Zhang.

Methodology: Jing Zhang.

Resources: Jing Zhang.

Software: Jing Zhang, Yimeng Fan, Yanlin Song.

Supervision: Jianguo Xu.

Writing – original draft: Jing Zhang, Yimeng Fan, Yanlin Song.

Writing – review & editing: Jing Zhang, Yimeng Fan, Yanlin Song, Jianguo Xu.

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