



ORIGINAL ARTICLE

High sensitivity and specificity of a 5-analyte protein and microRNA biosignature for identification of active tuberculosis

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Abstract

Objectives. Non-sputum-based tests to accurately identify active tuberculosis (TB) disease and monitor response to therapy are urgently needed. This study examined the biomarker capacity of a panel of plasma proteins alone, and in conjunction with a previously identified miRNA signature, to identify active TB disease. **Methods.** The expression of nine proteins (IP-10, MCP-1, sTNFR1, RANTES, VEGF, IL-6, IL-10, TNF and Eotaxin) was measured in the plasma of 100 control subjects and 100 TB patients, at diagnosis (treatment naïve) and over the course of treatment (1-, 2- and 6-month intervals). The diagnostic performance of the nine proteins alone, and with the miRNA, was assessed. **Results.** Six proteins were significantly up-regulated in the plasma of TB patients at diagnosis compared to controls. Receiver operator characteristic curve analysis demonstrated that IP-10 with an AUC = 0.874, sensitivity of 75% and specificity of 87% was the best single biomarker candidate to distinguish TB patients from controls. IP-10 and IL-6 levels fell significantly within one month of commencing treatment and may have potential as indicators of a positive response to therapy. The combined protein and miRNA panel gave an AUC of 1.00. A smaller panel of only five analytes (IP-10, miR-29a, miR-146a, miR-99b and miR-221) showed an AUC = 0.995, sensitivity of 96% and specificity of 97%. **Conclusions.** A novel combination of miRNA and proteins significantly improves the sensitivity and specificity as a biosignature over single biomarker candidates and may be useful for the development of a non-sputum test to aid the diagnosis of active TB disease.

Keywords: diagnosis, microRNA, plasma biomarkers, proteins, tuberculosis

INTRODUCTION

Tuberculosis (TB) is the leading cause of death from an infectious agent globally, with 10 million new cases in 2019, of which an estimated three million were undiagnosed and therefore did not receive treatment.¹ Detecting new cases of TB disease is usually a passive process, when individuals who have symptoms seek medical care. A recent cluster randomised control trial by Marks *et al.*² demonstrated that active case finding was more effective at reducing the prevalence of TB disease than current passive finding measures. Reducing TB burden may rely on actively screening endemic communities to identify infectious individuals and initiate early treatment, especially as evidence suggests that individuals with active TB disease may remain asymptomatic and infectious for a considerable period of time.^{2,3}

To actively screen communities for TB, the diagnostic test should be rapid, relatively cheap and ideally non-sputum-based. Existing tools for the identification of TB rely on the analysis of sputum for *Mycobacterium tuberculosis* by either microscopy, culture or PCR. Expecting sputum increases the risk of transmission, and good quality sputum can be challenging to obtain from children and the elderly. The recent study by Marks *et al.*² reported that only half the study participants could produce a quality sputum sample for Xpert analysis. Nor is sputum useful for identifying extrapulmonary TB, which accounts for 15–30% of TB cases.^{4,5} Alternative biological samples for TB diagnosis, such as urine, saliva and blood, are of increasing research interest for the development of new tests for TB diagnosis.

Efforts to identify *M. tuberculosis* or mycobacterial components from non-sputum biological samples have been largely unsuccessful to date, owing to a lack of sensitivity and specificity.^{6,7} Other measurable changes in the host response may aid in developing a new non-sputum test for screening for active TB. Host biomarkers are an attractive option for new TB diagnostics. Many candidates are currently being explored, with whole-blood or plasma mRNA,^{8,9} proteins¹⁰ and microRNA (miRNA)¹¹ being considered. This study builds on our previous study¹² that demonstrated the biomarker potential of a five miRNA signature to identify active TB disease. The utility of host protein biomarkers in plasma for the identification of active TB and to monitor response to therapy in a

cohort of TB patients, alone or in combination with the previously identified miRNA biosignature was examined. Host proteins are readily measured in the circulation and are relatively stable.¹³ Cytokines, chemokines and growth factors are known to change systemically during TB infection and provide promising candidates.^{10,14,15} IFN γ is crucial for TB disease control, yet biomarker studies have indicated that this cytokine alone does not have the diagnostic accuracy for a point-of-care test for active TB.^{16,17} By contrast, the CXCL10 chemokine, interferon-inducible protein-10 (IP-10), has emerged as a leading candidate as a biomarker for active TB disease, although additional studies with larger sample sizes in different regions are required to confirm these findings.^{16,18–22} To investigate this, we selected the following protein panel based on the diagnostic potential of the components in previous studies: IP-10,^{16,18–22} CCL2 (monocyte chemoattractant protein-1, MCP-1),²³ CCL5 (RANTES),²⁴ CCL11 (Eotaxin),^{25,26} TNF,^{27,28} soluble TNF receptor 1 (sTNFR1),²⁹ IL-6,^{17,24,27} IL-10^{17,27} and vascular endothelial growth factor (VEGF).^{22,28}

The potential of these nine host proteins as a biosignature to distinguish active TB patients from controls was investigated in a cohort of 100 pulmonary TB patients. The biomarker potential of these proteins was then compared to that of the miRNA signature previously shown to have good sensitivity and specificity at distinguishing active TB from healthy controls. All individual samples were from the Ningxia Hui Autonomous Region (NHAR), China, where the TB prevalence was estimated to be 61 cases per 100 000 between 2005 and 2008, although rates are predicted to be higher.³⁰ The prognostic potential of these proteins as indicators of the treatment response was also examined and the biomarker potential of a novel combined miRNA and protein biosignature for active TB described.

RESULTS

Protein expression in newly diagnosed TB patients

Samples were obtained from consenting TB patients and control subjects recruited from NHAR. All individuals were over the age of 18 and had no history of previous TB disease diagnosis or treatment, prior to this presentation. Patients and

controls were of either Han or Hui ethnicity and included both males and females (Supplementary table 1). The expression of nine proteins in the plasma of 100 newly diagnosed TB patients (treatment naïve) and 100 controls, matched for age and sex, was assessed. Six of these proteins (IP-10, MCP-1, sTNFR1, RANTES, VEGF and IL-6) were significantly increased in the plasma of TB patients than controls, while TNF levels were significantly decreased (Figure 1). Data from the TB patients and control subjects were then analysed stratified on the basis of gender and ethnicity (Han or Hui). No significant differences in protein expression were observed between males and females or TB patients from different ethnic backgrounds (Supplementary figures 1 and 2).

To determine the sensitivity and specificity of these proteins to distinguish TB patients from

healthy controls at diagnosis, receiver operator characteristic (ROC) curves were generated for single proteins (Figure 2). IP-10 was the protein with the highest area under the curve (AUC) of 0.874, in distinguishing TB patients from controls with a sensitivity and specificity of 75% and 87% respectively. To examine whether the accuracy of individual proteins could be improved, ROC curve analysis on combinations of several proteins was assessed. Together, the nine proteins could distinguish TB patients from controls with an AUC = 0.908 and sensitivity of 80% and specificity of 88%. The effect of varying the number of proteins in the panel was examined by binary logistic regression. This indicated that a panel of three proteins (IP-10, TNF and Eotaxin) was sufficient to maintain the same level of sensitivity and specificity while only marginally decreasing the AUC to 0.903 (Figure 3 and Table 1).

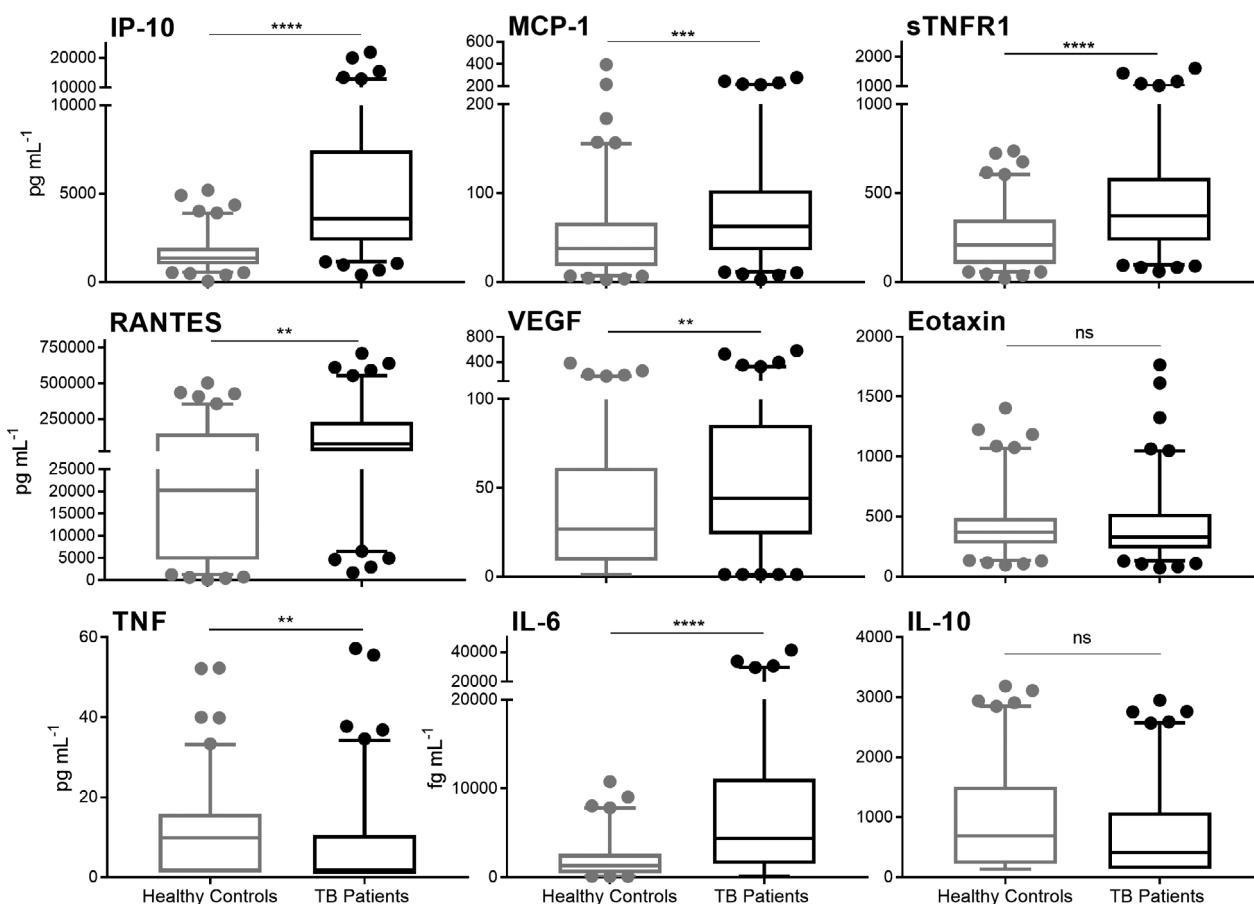


Figure 1. Expression of nine plasma proteins in 100 TB patients at diagnosis (treatment naïve) compared to age- and sex-matched healthy controls. Data are represented by boxplots with median and interquartile range with bars from the 5–95 percentile. Samples were statistically analysed by Welch’s *t*-test (ns = no significance, ***P*-value < 0.01, ****P*-value < 0.001, *****P*-value < 0.0001).

Changes in plasma protein expression during treatment

The capacity of the expression levels of these nine proteins to identify response to therapy of TB patients during their treatment course was then assessed. Samples taken at 1, 2 and 6 months during standard therapy were measured. Interestingly, only IP-10 and IL-6 levels significantly decreased within the first month of therapy and continued to trend towards levels in healthy control subjects during treatment (Figure 4). All of the proteins that were significantly elevated at diagnosis remained increased compared to the levels in healthy controls at the end of treatment, except for IL-6 (Figure 4).

Plasma protein levels at diagnosis (time 0) were stratified based on whether the patient successfully

completed recommended treatment. Treatment failure was determined by a positive sputum culture at the end of 6-month standard therapy and/or progressive radiological changes. Eleven TB patients out of the 100 were considered to have failed therapy. As shown in Figure 5, these 11 TB patients had significantly elevated IL-6 and RANTES plasma levels at diagnosis when compared to the TB patients that successfully completed therapy.

Protein and miRNA

In addition to the nine proteins measured in this study, our previous study¹² measured the biomarker potential of a panel of 10 miRNA in these same plasma samples by qPCR. The expression levels for all nine proteins and 10 miRNA were collated. Analysis of all 19 analytes

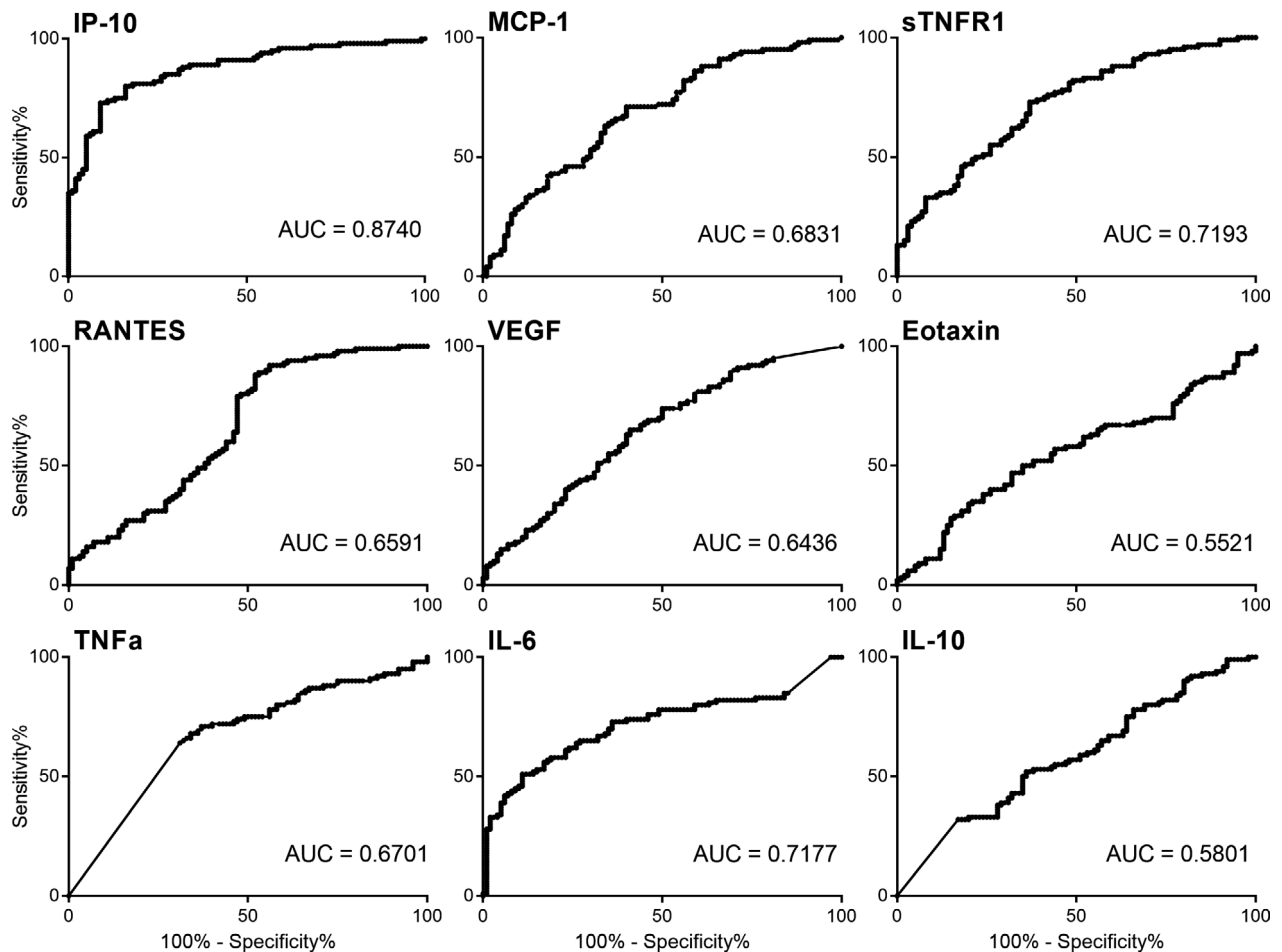


Figure 2. ROC curves and AUC values for the nine individual proteins to distinguish TB patients at time of diagnosis from controls. IP-10 plasma levels generated the best diagnostic accuracy from the nine proteins with an AUC = 0.874. AUC, area under the curve; ROC, receiver operator characteristic.

together demonstrated an AUC = 1.000 with sensitivity of 100% and specificity of 100% for distinguishing TB patients from control subjects at diagnosis (Supplementary table 2). As shown in Figure 6, reducing the number of miRNA and proteins in this combined panel to just five analytes (IP-10, miR-29a, miR-146a, miR-99b and miR-221) still gave an AUC of 0.995 with a sensitivity of 96% and specificity of 97%. IP-10 was the most accurate predictor of TB disease from all 19 plasma miRNA and proteins.

DISCUSSION

Decreasing the TB burden globally requires new diagnostic tools. A simple, rapid test that does not rely on sputum would greatly benefit this goal. Our study identified a 3-protein biomarker panel that distinguishes individuals with active TB disease from controls with an AUC = 0.903 and a high sensitivity and specificity of 80% and 88%, respectively. Interestingly, most of this diagnostic accuracy was because of the single plasma protein IP-10.

IP-10 was significantly up-regulated in the plasma of Chinese TB patients compared to controls. IP-10 has been identified in previous studies as a biomarker candidate for diagnosing active TB disease.^{21,31–34} Studies in both low and high TB endemic regions including China, Uganda, Norway and Denmark have demonstrated that IP-10 in blood, plasma or serum was significantly up-regulated in TB patients in comparison with control cohorts.^{31–34} Alone, however, IP-10 does not have the sensitivity required for a non-sputum biomarker test for active TB disease, which according to the WHO requirements is $\geq 90\%$.³⁵ ROC curve analysis of IP-10 in previous biomarker studies ranged from 0.801 to 0.950 and sensitivity ranged from 72.4 to 83.3%,^{31,32,34} which is in accordance with our study.

Changes in plasma IP-10 levels during infection are not unique to TB disease. IP-10 is a chemokine associated with inflammation. An increase in IP-10 levels has shown biomarker potential in a number of TB biomarker studies.^{19,21,36} Increased IP-10 has also been associated with a number of other infections in adults with respiratory viruses such as influenza,³⁷ and with severe malaria³⁸ and in cystic fibrosis patients during pulmonary exacerbations.³⁹ Binary logistic regression showed that increased IP-10, together with TNF and

Eotaxin levels, was able to identify TB patients from control subjects with a high sensitivity and specificity, despite Eotaxin levels not differing significantly between the two groups. This may indicate a biological link between the levels of these three proteins during TB disease progression.

Practically, however, increases in IP-10 and IL-6 plasma levels, which were able to identify TB patients with an AUC = 0.875, a sensitivity of 75% and specificity of 87%, represent changes that may be easier to measure as indicators of inflammation and disease. These data suggest a role for IP-10 as a triage tool whereby high levels are indicative of active infection and the requirement for further testing. Utilising a plasma IP-10 cut-off level of 1000 pg mL⁻¹, whereby any individual with higher levels would be deemed positive, identifies 97% of all TB patients in this cohort. This equates to a false-negative rate of only 3%, although a high false-positive rate of 72%. Adding IL-6 to this test, with a cut-off level of 3000 fg mL⁻¹, lowers the false-positive rate to 12%. If applied as a triage tool, high expression of both IP-10 and IL-6 is strongly indicative of TB disease and could be used to trigger further microbiological testing.

One way to improve the sensitivity of biomarkers may be the addition of multiple analytes^{25,26,40} or to expand the types of biomarkers within the panel. Our previous study showed that a panel of five miRNA could identify active TB patients with a sensitivity and specificity of 94% and 88%, respectively.¹² Additional studies have also identified the potential of plasma miRNAs as biomarkers for active TB disease.¹¹ Therefore, the biomarker potential of a combined miRNA and protein panel was assessed. Analysing these data using the combination of 19 miRNA and protein analytes greatly improved diagnostic accuracy with an AUC of 1.00, sensitivity and specificity of 100%. A panel of 19 markers is unrealistic for development as a biomarker test. Reducing this panel to five analytes (IP-10, miR-29a, miR-146a, miR-99b and miR-221) maintained good accuracy with an AUC = 0.995, sensitivity of 96% and specificity of 97%. These data indicate that a novel combination of miRNA and protein biomarkers was able to improve the accuracy of either proteins or miRNA biomarker candidates alone. It is increasingly clear that multiple biomarkers will be required to reach the necessary high sensitivity

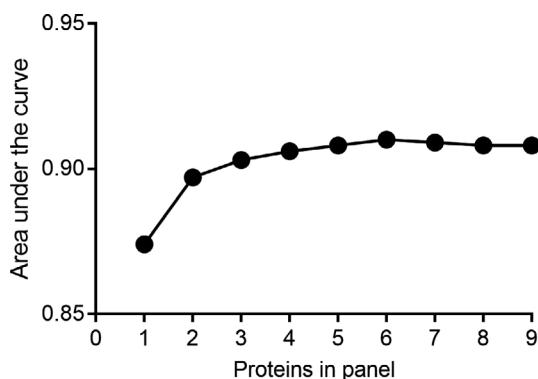


Figure 3. The area under the curve for increasing numbers of plasma proteins in the biosignature for distinguishing pulmonary TB patients at the time of diagnosis from control subjects, as determined binary logistic regression.

and specificity outlined by the WHO,⁴¹ and this novel combination may be a promising avenue for active TB biomarker development.

Using proteins as biomarkers for a non-sputum test is attractive as proteins are relatively stable in the circulation and easy to measure with limited tools. While miRNA are also considered relatively stable in the circulation, they are more difficult to measure and require PCR technology. Requiring the simultaneous measurement of both protein and miRNA is a drawback for developing this biomarker test. As technology advances, new techniques capable of measuring miRNA in bead-based assays are being developed⁴² and there is the potential to combine the measurement of both miRNA and protein on a similar platform.

One requirement for advancing a biomarker test for active TB is developing a universal test with diagnostic accuracy across multiple ethnicities and geographical locations. In this study, the biomarker panel was equally effective in identifying TB disease in both males and

females, and in the Han and Hui subjects (Supplementary figures 1 and 2), further studies in larger diverse populations at multiple locations will be required to determine the effectiveness of this panel as a universal diagnostic tool. There is a growing body of literature assessing host proteins as biomarkers of active TB disease.⁴³ The lack of validation of candidates between populations, and differences in sample choice, preparation and processing are all inhibiting progress in diagnostic development. There is a need to assess the potential of these biomarker candidates in different ethnic and geographical populations to identify both baseline and levels associated with disease. A recent review acknowledged that one shortcoming to advancing this field is the availability of data.⁴³ It proposed a TB biomarker database as an efficient and necessary tool to summarise the current findings and match studies with similar biomarker candidates to improve validation and the development of a test for active TB. Developing systems to share biomarker data more efficiently and confirm results in multiple populations is essential to progress biomarker development.

Another consideration for TB biomarker development is the expression profile of individuals with LTBI. It is estimated that almost a quarter of the global population have LTBI.⁴⁴ Most will not develop active TB disease and do not require treatment.⁴⁵ However, studies have shown that in the 6 months before reactivation there are measurable changes in the whole-blood mRNA profiles of reactivating LTBI-infected individuals.⁴⁶ Reactivation of LTBI infection may also modify the miRNA and protein expression profiles and this may aid early diagnosis. Future biomarker studies could include analysis of reactivation within a LTBI cohort.

Table 1. Sensitivity, specificity, negative predictive value and positive predictive value for increasing numbers of proteins in a biosignature panel, to differentiate TB patients from controls

Number of proteins in panel	Proteins added to panel	Sensitivity (%)	Specificity (%)	Positive predictive value (%)	Negative predictive value (%)
1	IP-10	75.00	87.00	85.23	77.68
2	TNF	79.00	85.00	84.04	80.19
3	Eotaxin	80.00	88.00	86.96	81.48
4	RANTES	79.00	87.00	85.87	80.56
5	IL-6	79.00	86.00	84.95	80.37
6	sTNFR1	82.00	87.00	86.32	82.86
7	MCP-1	80.00	87.00	86.02	81.31
8	IL-10	80.00	88.00	86.96	81.48
9	VEGF	80.00	88.00	86.96	81.48

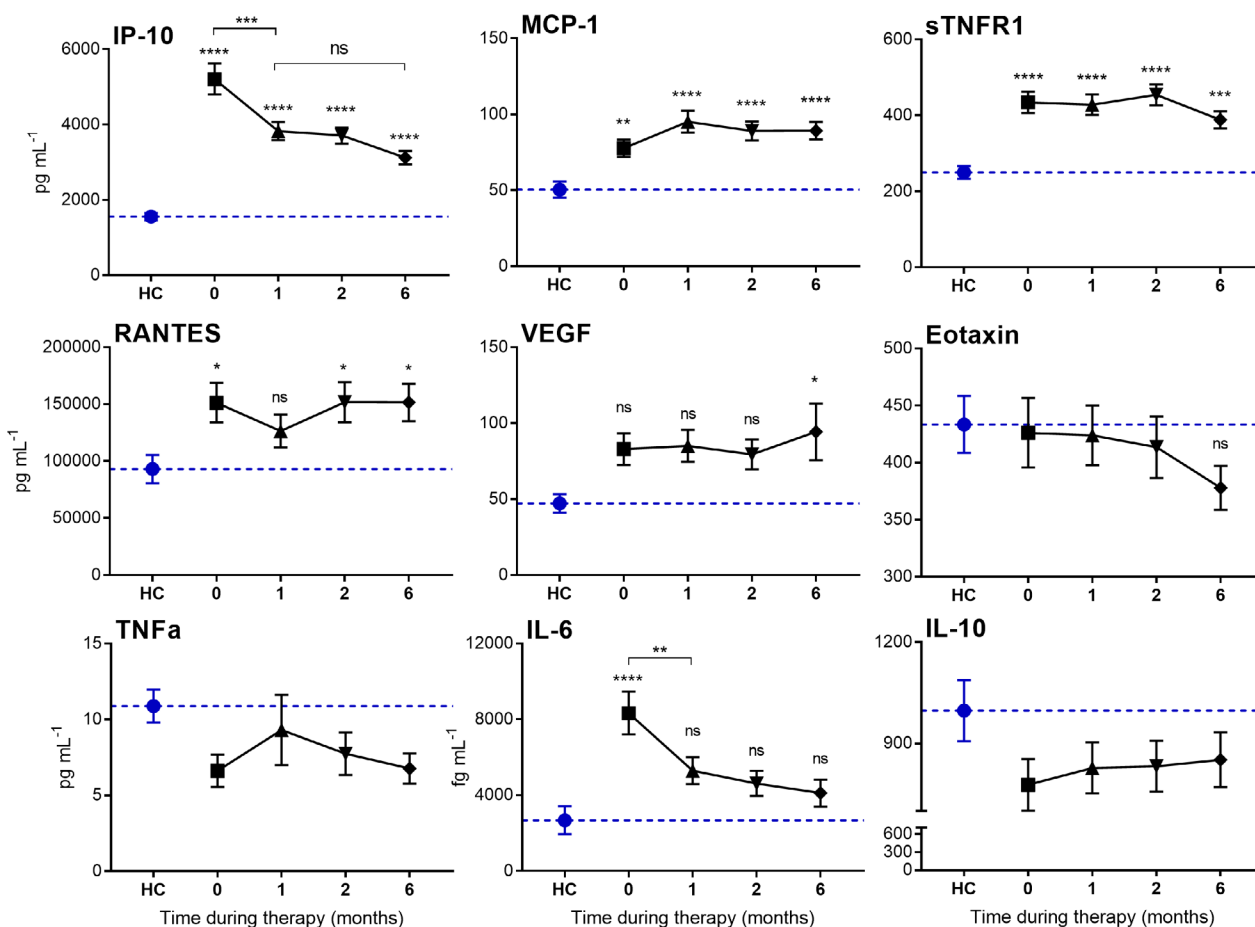


Figure 4. Expression of nine plasma proteins in TB patients during 6 months of standard TB treatment and in healthy control subjects (HC). Data are mean ± SEM, *n* = 100 per group. The dotted line is representative of the mean protein level for the HC cohort. Statistical differences in levels compared to healthy controls were assessed by one-way ANOVA with Tukey's multiple comparisons test (ns = no significance, **P*-value < 0.05, ***P*-value < 0.01, ****P*-value < 0.001, *****P*-value < 0.0001).

One strength of this study was that TB patients were monitored throughout their treatment, providing samples during therapy. This allowed us to identify how the levels of these inflammatory proteins change with treatment and recovery. There is growing interest in biomarkers to not only identify TB disease but also to monitor the response to therapy and so rapidly identify patients who are not responding to treatment. This failure to respond may be caused by drug resistance, poor compliance or poor drug absorption. Currently, TB cure is defined as being sputum culture negative at the end of intensive phase (2 months) and at treatment completion. Interestingly, of the 6 proteins significantly elevated at diagnosis in TB patients, only two, IP-10 and IL-6, declined significantly over the 6-month therapy, and only IL-6 returned to baseline

levels. In another study, IP-10 was similarly shown to be reduced by therapy.²² This study also found that VEGF fell significantly during treatment, but this finding was not reproduced by our study. Further initial levels of IL-6 and RANTES were significantly higher in the plasma of TB patients who went on to fail therapy than in patients who successfully completed treatment. These proteins may have potential as early indicators of treatment outcome. However, the group of TB patients that failed therapy was small with only 11 individuals, and further research with larger cohorts is required to validate these findings. Four proteins (MCP-1, RANTES, sTNFR1 and VEGF) that were elevated pre-treatment remained significantly elevated upon the conclusion of treatment, indicating that the extensive inflammation and lung remodelling that occurs

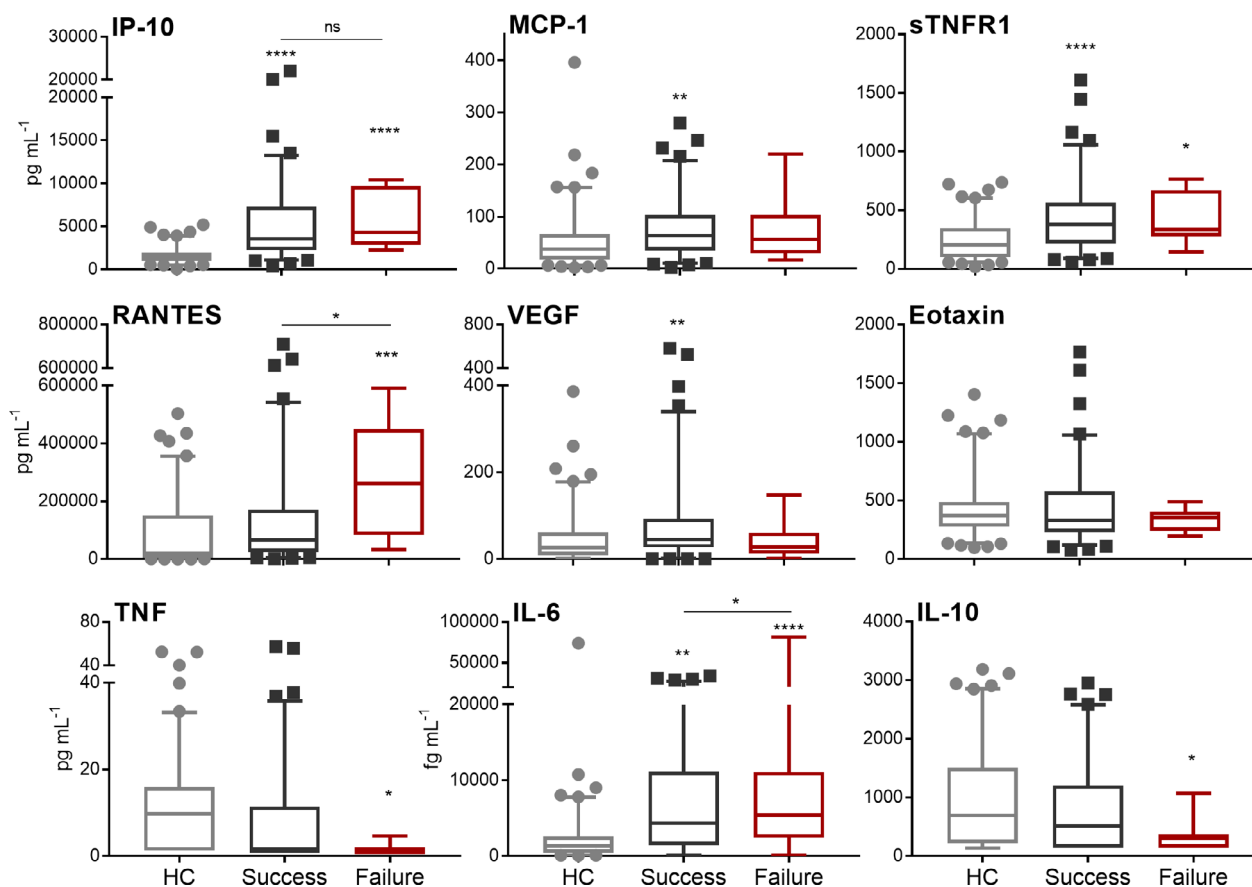


Figure 5. Plasma protein levels in TB patients at time of diagnosis (treatment naïve) who successfully completed standard therapy ($n = 89$, black) or who failed therapy ($n = 11$, red) compared to levels in controls ($n = 100$, grey). Statistical differences were assessed by one-way ANOVA with Tukey's multiple comparisons test. Significant differences displayed are relative to the HC group unless otherwise indicated (ns = no significance, * P -value < 0.05, ** P -value < 0.01, *** P -value < 0.001, **** P -value < 0.0001).

during TB disease is continuing even at the completion of successful treatment.⁴⁷ This is consistent with the continuing inflammation identified by PET/CT scanning of pulmonary TB patients at the completion of therapy.⁴⁸

It is a strength of this study to have sampled plasma from the TB patients four times throughout treatment, and a limitation that control subjects were only sampled once, so we are unable to assess variation in the levels of these plasma proteins in controls over time. Further research is needed to determine the individual variation in baseline plasma protein levels and how this may impact the utility of these proteins as indicators of treatment response. There is also the need to determine the biomarker signature of those with other respiratory diseases to determine whether these proteins, either as a triage tool or as a point-of-care test, could

distinguish active TB from other diseases with similar symptoms. The method used to diagnose individuals with active TB disease followed national guidelines. 33% of these individuals had microbiological confirmed sputum smear of *M. tuberculosis*. At the time of sample collection, GeneXpert analysis was not available in the NHAR. Future work should address these weaknesses to strengthen the research in this area.

Conclusion

Plasma proteins are a viable option as potential biomarkers of active TB disease with IP-10 the leading candidate with an AUC = 0.874, although a biomarker panel or signature of multiple analytes is more likely required to achieve a better diagnostic accuracy. Plasma IP-10 may have

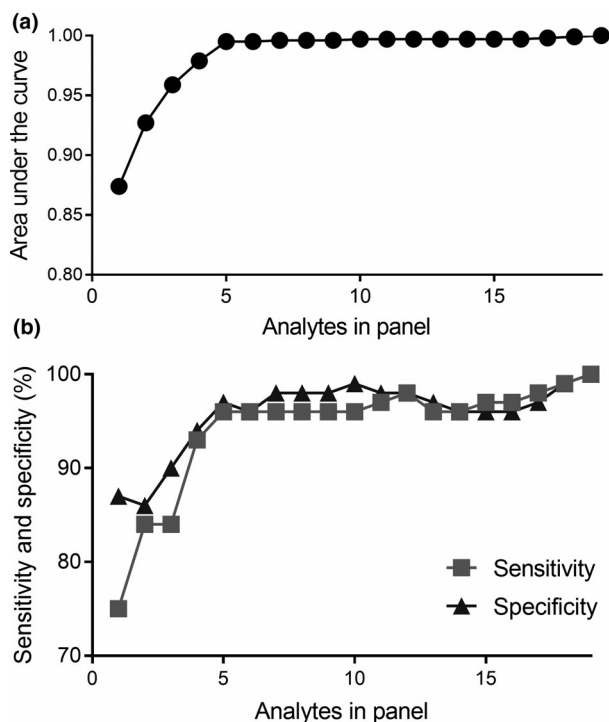


Figure 6. The area under the curve (a), sensitivity and specificity (b) for varying the number of plasma analytes in a biomarker signature to distinguish active TB patients from control subjects. The 19 analytes consisted of nine proteins measured in this study and 10 miRNA reported in a previous study.¹² The best single biomarker candidate was IP-10.

an important role as a triage tool, identifying patients that require further investigation to exclude TB. This study demonstrates that a novel combination of miRNA and protein biomarkers performed better than either proteins or miRNA biomarker panels alone. Further validation of these biomarker candidates in other geographical or ethnic populations is required to advance TB biomarker development.

METHODS

Study population

All participants were recruited through the TB Control Program in the NHAR, China,¹² and TB patients were diagnosed based on national guidelines.³⁰ National guidelines for TB detection involve passive case finding, whereby individuals with symptoms seek care. A chest X-ray was performed for every patient and control subject recruited. These were read by two independent medical specialists, and all patients had radiological findings in keeping with active pulmonary TB. Sputum analysis including smear microscopy and culture was performed

where samples were available. Thirty-three patients had culture-positive sputum, in line with data from NHAR, where culture-positive TB was identified in approximately one-third of new patients. The diagnosis of TB was made on the basis of clinical suspicion, radiological findings and sputum smear and culture. In sputum culture-negative cases, clinical response to therapy coupled with radiological improvement was used to confirm the initial diagnosis of pulmonary TB. Ethical approval for this study was granted from the University of Sydney Ethics Review Committee (protocol no. 2012/1076) and the Ningxia Medical University Human Ethics Committee (approval date 6/6/2013). All subjects gave written informed consent before their enrolment in the study.

Plasma samples were collected from 100 TB patients at diagnosis (treatment naive) and 100 healthy, age- and gender-matched controls. The characteristics of participants are summarised in Supplementary table 1. The TB patients were initiated on standard TB treatment according to national guidelines that follow WHO DOTS guidelines for 6 months. Additional blood samples were collected from patients at 1, 2 and 6 months during treatment.

Sample collection and preparation

Ten millilitres of venous blood was collected into an EDTA tube.¹² Blood was processed within 2 h of collection. Blood was separated by centrifugation at 800 g for 10 min. Plasma was removed and stored at -80°C . Samples were examined for haemolysis using a haemolysis chart and samples with $> 100 \text{ mg dL}^{-1}$ of haemoglobin were discarded.

Protein expression

Protein expression was detected in plasma samples by cytometric bead array (CBA) (BD Biosciences, Australia). IL-6 and IL-10 were assessed with an enhanced sensitivity CBA, as per the manufacturer's protocol, with a lower detection limit of 274 fg mL^{-1} . IP-10, MCP-1, RANTES, TNF, sTNFR1, Eotaxin and VEGF levels were measured with the CBA array kit, with a detection limit of 2.5 pg mL^{-1} , as per the manufacturer's instructions. Plasma samples were diluted 1:4 as recommended for protein quantification. RANTES expression was higher than the array's upper limit of detection of 2500 pg mL^{-1} , and therefore, the plasma samples were diluted 1:200 for RANTES quantification. Any samples that fell below the array's lower limit of detection were allocated a value equal to the half limit of detection. Samples were acquired by flow cytometer, either a BD Fortessa X20 (University of Technology Sydney) or BD Canto-II (Centenary Institute).

Statistical analysis

Flow cytometry files were analysed by FCAP Array software (BD Biosciences, Australia). Statistical analysis was in GraphPad Prism v7 by Welch's *t*-test or one-way ANOVA with Tukey's multiple comparison test.

Receiver operating characteristic (ROC) curves for single proteins were analysed in GraphPad Prism v7. ROC curves

show the relationship between clinical sensitivity and specificity. They are generated by plotting the sensitivity and specificity of data based on a series of cut-off values. Sensitivity refers to the rate of true positives (TB patients) correctly identified. Specificity is the rate of true negatives (control subjects) correctly identified. For the resulting ROC curve generated, an AUC value is calculated. An AUC = 1.00 demonstrates that the test is 100% accurate and every positive or negative sample is correctly identified. The positive predictive value (PPV) and negative predictive value (NPV) of each biomarker were examined. PPV refers to the probability that a positive test result is a true TB patient whereas NPV refers to the probability that a negative test result does not have TB.

Receiver operator characteristic curves for both protein panels and combined miRNA and protein panels were generated in IBM SPSS Statistics 25 software following binary logistic regression calculations.

The protein data set generated in this study was then combined with our miRNA data set (10 miRNA – confirmatory study) previously published by Barry *et al.* 2018 which describes the sample collection and a miRNA profiling biomarker study on these samples.¹² miRNA biomarker candidates were miR-21-5p; –99b-5p; –29a-5p; –223-5p; –221-3p; –146a-5p; –26a-5p; –28-5p; –133a; and –652-3p.

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JP, NB, BS proposed the experimental design. JP executed the experiments. BS, SB, YY, GG, XW and WB undertook the collection of all plasma samples. SB, BS, ME and WB designed the miRNA profiling experiments,¹² of which the miRNA confirmatory dataset was used in this study. AF helped with data interpretation and analysis. JP and BS composed the manuscript. All authors reviewed and edited the manuscript. This work was supported by the National Health and Medical Research Council – Centres for Research Excellence in Tuberculosis Control [APP1043225, APP1153493]. JP was supported by an Australian Government Research Training Program Scholarship.

CONFLICT OF INTEREST

The authors report no conflicts of interest with regard to this publication.

AUTHOR CONTRIBUTION

Jessica L Pedersen: Conceptualization; Data curation; Formal analysis; Investigation; Methodology; Writing-original draft. **Simone E Barry:** Data curation; Investigation; Writing-review & editing. **Nilesh J Bokil:** Conceptualization; Project administration; Supervision; Writing-review & editing. **Magda Ellis:** Data curation; Writing-review & editing. **YuRong Yang:** Investigation; Resources; Writing-review & editing. **Guangyu Guan:** Investigation; Resources; Writing-review & editing. **Xiaolin Wang:** Investigation; Resources; Writing-review & editing. **Alen Faiz:** Software; Writing-review & editing. **Warwick Britton:**

Conceptualization; Project administration; Writing-review & editing. **Bernadette M Saunders:** Conceptualization; Funding acquisition; Project administration; Supervision; Writing-review & editing.

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Supporting Information

Additional supporting information may be found online in the Supporting Information section at the end of the article.



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