

Using IPDmada to perform statistical analyses of diagnostic accuracy in primary studies: explanation and elaboration with a case study

Xueli Zhang¹, Junfeng Wang²

¹College of Science, Wuhan University of Science and Technology, Wuhan, China; ²Division of Pharmacoepidemiology and Clinical Pharmacology, Utrecht Institute for Pharmaceutical Sciences, Utrecht University, Utrecht, The Netherlands

Correspondence to: Junfeng Wang, PhD. Assistant Professor, Division of Pharmacoepidemiology and Clinical Pharmacology, Utrecht Institute for Pharmaceutical Sciences, Utrecht University, Universiteitsweg 99, 3584CG Utrecht, The Netherlands. Email: wangjunfeng7@gmail.com.

Abstract: IPDmada is web-based R Shiny application recently developed for individual patient data meta-analysis (IPD-MA) of diagnostic test accuracy (DTA). IPDmada provides a wide range of analyses in IPD-MA, which are also applicable in primary studies of diagnostic accuracy. These analyses include summary table of patient characteristics (usually referred to as “Table 1”); calculation of all the test performance measures; forest plots of sensitivity and specificity and summary receiver operating characteristic (ROC) plots for dichotomized tests; distribution of test results and ROC curves for continuous test; forest plots of areas under the curve (AUCs) for continuous tests; distribution of covariates and covariate adjusted ROC curves for covariate analysis, which provide more than enough results for a complete DTA study. In this explanation and elaboration document, we use IPDmada to perform both basic and advanced statistical analyses in evaluating a diagnostic test, to show the feasibility of applying IPDmada in primary studies of DTA. The data source used in this case study is from a previous publication by Norton et al., which investigated the diagnostic accuracy of transient evoked otoacoustic emissions (TEOAEs), distortion product otoacoustic emissions (DPOAEs), and auditory brain stem responses (ABRs) for neonatal hearing impairment. In conclusion, IPDmada is an easy-to-use tool, which can provide an one-stop solution to researchers without strong statistical background for all the analyses needed in a DTA study.

Keywords: IPDmada; diagnostic test; sensitivity; specificity; receiver operating characteristic curve (ROC curve)

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Introduction

In a study of diagnostic test accuracy (DTA), one or more tests (referred to as the index tests) are evaluated for their ability in detecting or predicting a target condition or health status (1). The index test results are compared with a reference standard, which is the best available method for determining the target condition. Based on this comparison between the index test and the reference standard, diagnostic accuracy can be quantified with many performance measures, including but not limited to sensitivity, specificity, positive predictive value (PPV), negative predictive value (NPV), positive likelihood ratio

(LR+), negative likelihood ratio (LR-), diagnostic odds ratio (DOR), and areas under receiver operating characteristic (ROC) curve (AUC) (2,3).

The statistical analyses producing the above results can be done in many software, such as SPSS, Stata, SAS, and R. However, researchers need to calculate all those performance measures step-by-step. It is preferable to have an easy-to-use tool that requests minimum work by the users (e.g., only uploading the data) and provides all the results needed in a DTA study.

IPDmada (4) is web-based R Shiny application recently developed for individual patient data meta-analysis (IPD-

Table 1 Description of variables in the dataset

Variables	Description	Value
id	Patient ID	
ear	Ear	1= left, 2= right
sitenum	Center ID	
currage	Corrected age	
gender	Gender	1= female, 2= male
d	Hearing impaired	0= no, 1= yes
y1	DPOAE 65 at 2 kHz	
y2	TEOAE 80 at 2 kHz	
y3	ABR	

TEOAE, transient evoked otoacoustic emission; DPOAE, distortion product otoacoustic emission; ABR, auditory brain stem response.

MA) of DTA. IPDmada provides a wide range of analyses in IPD-MA, which are also applicable in primary studies of diagnostic accuracy. In this explanation and elaboration document, we use IPDmada to perform both basic and advanced statistical analyses in evaluating a diagnostic test to show the feasibility of applying IPDmada in primary studies of DTA.

Data

The data source used in this case study is from a previous publication by Norton *et al.*, which investigated the diagnostic accuracy of transient evoked otoacoustic emissions (TEOAEs), distortion product otoacoustic emissions (DPOAEs), and auditory brain stem responses (ABRs) for neonatal hearing impairment (5).

The dataset is publicly available and can be downloaded from the website of Fred Hutchinson Cancer Research Center, Diagnostic and Biomarkers Statistical (DABS) Center (6), via this link: <https://research.fredhutch.org/content/dam/stripe/diagnostic-biomarkers-statistical-center/files/nnhs2.csv>.

The public dataset contains 5,058 records from 2,742 individuals in 6 centers, and the variables and their description are given in *Table 1*.

Statistical analysis

Since IPDmada requests data in a specific format (i.e., CSV format), and with pre-defined variable names (i.e., Study, disease, test.results), data preparation is needed before

uploading raw data to IPDmada.

In this illustration case study, data preparation is performed in spreadsheets, which can be done with any spreadsheet software, e.g., Microsoft Excel, Apple Numbers, Google Sheets, and WPS Spreadsheet, etc.

When data is ready for use, it will be uploaded to IPDmada (<https://jwang7.shinyapps.io/ipdmada/>), and all further analyses will be done automatically by IPDmada. Results can be downloaded either in tables (as CSV files) or figures (as PDF files).

The statistical analyses provided in IPDmada include:

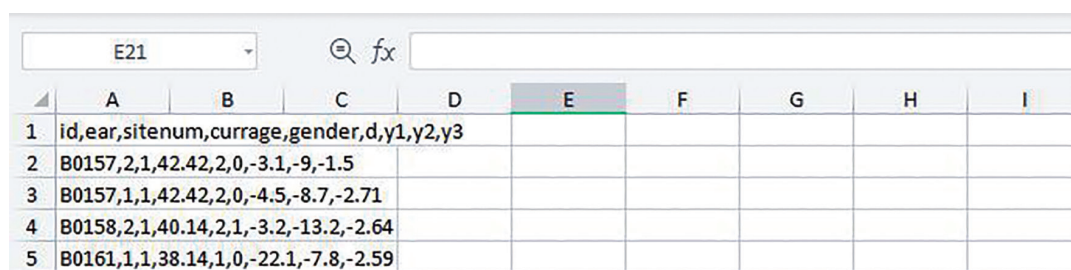
- (I) Test performance measures calculated from confusion matrix:
 - ❖ True positive (TP), false negative (FN), false positive (FP), and true negative (TN);
 - ❖ Sensitivity and specificity;
 - ❖ PPV and NPV;
 - ❖ LR+ and LR-;
 - ❖ DOR.
- Three options are provided to determine the positivity threshold, which is used to generate the confusion matrix:
 - ❖ Different thresholds are used across centers;
 - ❖ Optimal thresholds are determined for each center (based on Youden's Index);
 - ❖ One specific threshold is used in all centers.
- (II) Test performance based on the distribution of test results:
 - ❖ Distribution of test results in disease and non-disease groups;
 - ❖ ROC curve and AUC;
 - ❖ Distribution of covariates in disease and non-disease groups;
 - ❖ Covariate adjusted ROC curve and covariate-adjusted AUC.

Without loss of generality, in this case study we only show the analyses of TEOAEs, and the same analyses can be applied to DPOAEs and ABRs as well. Please note that, all the analyses presented in this report are only for illustration purposes, and sometimes variables are analyzed as a hypothetical example without practical meaning, thus no clinical conclusions should be drawn from these results.

Case study

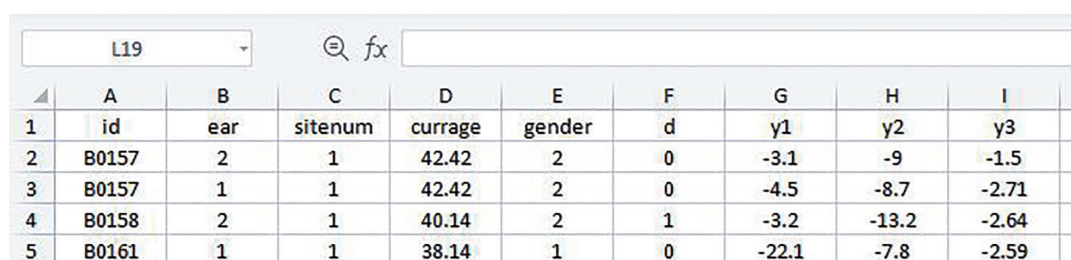
Analyses of data from multiple centers for one test

In a multiple centers study, diagnostic data are collected from N centers. In this case, the data structure is similar



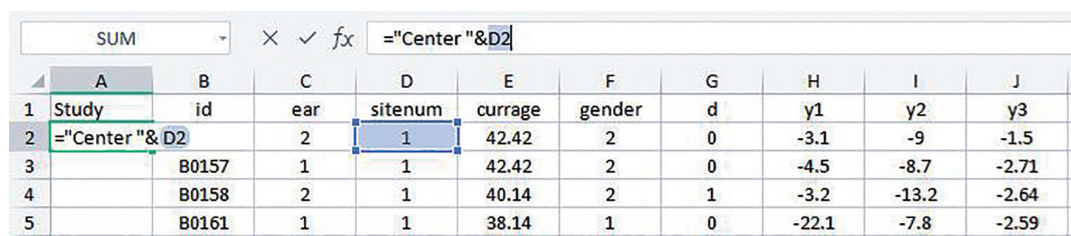
	A	B	C	D	E	F	G	H	I
1	id,ear,sitenum,currage,gender,d,y1,y2,y3								
2	B0157,2,1,42.42,2,0,-3.1,-9,-1.5								
3	B0157,1,1,42.42,2,0,-4.5,-8.7,-2.71								
4	B0158,2,1,40.14,2,1,-3.2,-13.2,-2.64								
5	B0161,1,1,38.14,1,0,-22.1,-7.8,-2.59								

Figure 1 Data read in as comma-separated values.



	A	B	C	D	E	F	G	H	I
1	id	ear	sitenum	currage	gender	d	y1	y2	y3
2	B0157	2	1	42.42	2	0	-3.1	-9	-1.5
3	B0157	1	1	42.42	2	0	-4.5	-8.7	-2.71
4	B0158	2	1	40.14	2	1	-3.2	-13.2	-2.64
5	B0161	1	1	38.14	1	0	-22.1	-7.8	-2.59

Figure 2 Data read in as columns.



	A	B	C	D	E	F	G	H	I	J
1	Study	id	ear	sitenum	currage	gender	d	y1	y2	y3
2	="Center" & D2		2	1	42.42	2	0	-3.1	-9	-1.5
3		B0157	1	1	42.42	2	0	-4.5	-8.7	-2.71
4		B0158	2	1	40.14	2	1	-3.2	-13.2	-2.64
5		B0161	1	1	38.14	1	0	-22.1	-7.8	-2.59

Figure 3 Insert center names as Study ID.

with an IPD-MA including N primary studies. A single center study, or a study treating data from different centers as if from one single center, can be seen as a special case of a multiple centers study with $N=1$.

Data preparation

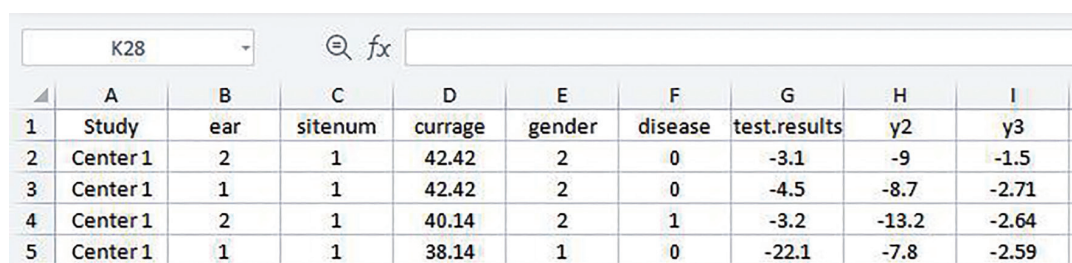
We start the data preparation from the CSV file downloaded from the website given in the Data section. Depending on the software, data can be read in as comma-separated values (Figure 1) or columns (Figure 2). Data in the form of Figure 1 can be transformed into Figure 2 with the *Convert Text to Columns Wizard*.

To facilitate the analyses provided in IPDmada, three columns are required and have to be named “Study”, “test. results”, and “disease”, and the variable names are case

sensitive (e.g., Study must have a capital S). The column called “Study” indicates which primary study the records come from. In the analyses of a multiple centers study, we can use this column to reflect data are from which center. We insert one column to the left and set the header of this column as “Study”, and add a formula [=“Center”\&D2] in cell A2 (Figure 3). If all the data are from one single center, just add the center name in cell A2.

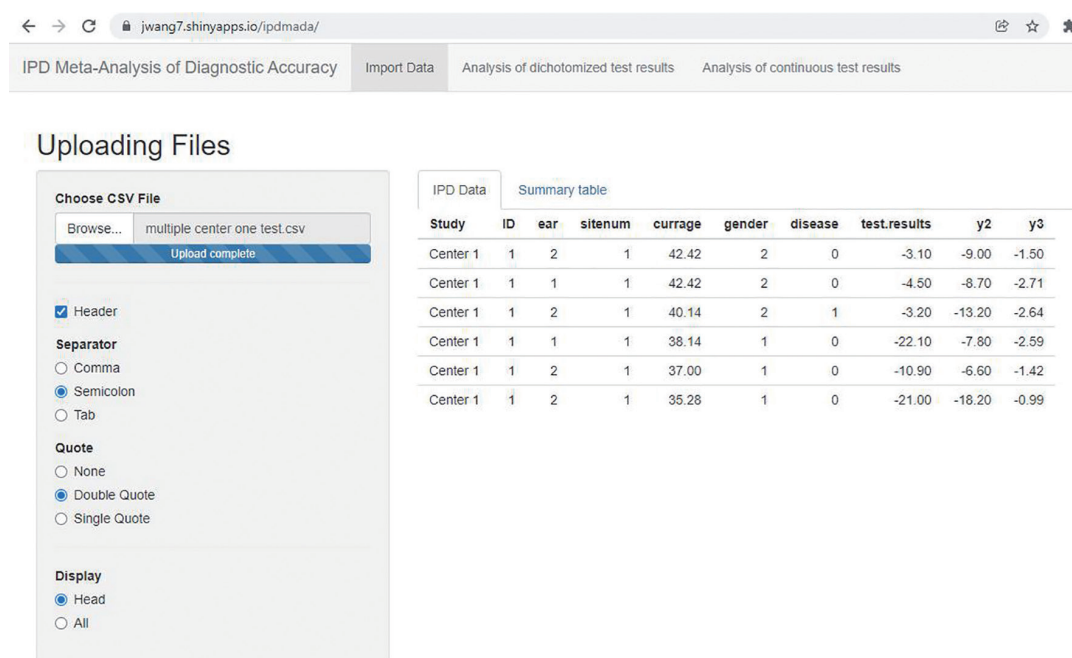
Double click on the *plus sign* (+) in the bottom right corner of cell A2, then the formula (or value in case of single center study) is copied to all the rest cells in Column A.

Change the header of Column G from “d” to “disease”, and change the header of Column H from “y1” to “test. results”. Remove Column B “id” which is not needed and may lead confusion to IPDmada (Figure 4).



	A	B	C	D	E	F	G	H	I
1	Study	ear	sitenum	currage	gender	disease	test.results	y2	y3
2	Center 1	2	1	42.42	2	0	-3.1	-9	-1.5
3	Center 1	1	1	42.42	2	0	-4.5	-8.7	-2.71
4	Center 1	2	1	40.14	2	1	-3.2	-13.2	-2.64
5	Center 1	1	1	38.14	1	0	-22.1	-7.8	-2.59

Figure 4 Data ready for uploading to IPDmada. IPD, individual patient data.



IPD Meta-Analysis of Diagnostic Accuracy | Import Data | Analysis of dichotomized test results | Analysis of continuous test results

Uploading Files

Choose CSV File

Browse... multiple center one test.csv

Upload complete

☒ Header

Separator

☐ Comma

☒ Semicolon

☐ Tab

Quote

☐ None

☒ Double Quote

☐ Single Quote

Display

☒ Head

☐ All

Summary table

Study	ID	ear	sitenum	currage	gender	disease	test.results	y2	y3
Center 1	1	2	1	42.42	2	0	-3.10	-9.00	-1.50
Center 1	1	1	1	42.42	2	0	-4.50	-8.70	-2.71
Center 1	1	2	1	40.14	2	1	-3.20	-13.20	-2.64
Center 1	1	1	1	38.14	1	0	-22.10	-7.80	-2.59
Center 1	1	2	1	37.00	1	0	-10.90	-6.60	-1.42
Center 1	1	2	1	35.28	1	0	-21.00	-18.20	-0.99

Figure 5 Uploading data to IPDmada. IPD, individual patient data.

After all these steps, data can be saved as a CSV file, and ready for uploading to IPDmada.

Analyses results

Upload data and check the correctness

Data will be shown in a table right after being uploaded (Figure 5). Please note that, because decimal point (.) and comma (,) are used in different ways in different regions in the world, the separator used in the CSV file may vary based on the system settings (regional settings in the control panel). If an error occurs, please switch the separator from “Semicolon” to “Comma” (or vice versa), usually this will solve the problem.

Calculation of test performance based on confusion matrix

Test performance calculated based on confusion matrix are shown in the “Analysis of dichotomized test results” panel.

The threshold(s) can be determined by (I) entering different threshold values for each center with the slider bar; (II) using the optimal threshold values for each center; (III) entering one threshold value for all centers in the input box. The first option is useful when different thresholds are used in different centers, e.g., index test is from different manufacturers and the recommended thresholds are different; the second option is often desired when we need to determine an optimal threshold; the third option allows the users to tune the threshold for sensitivity analysis and

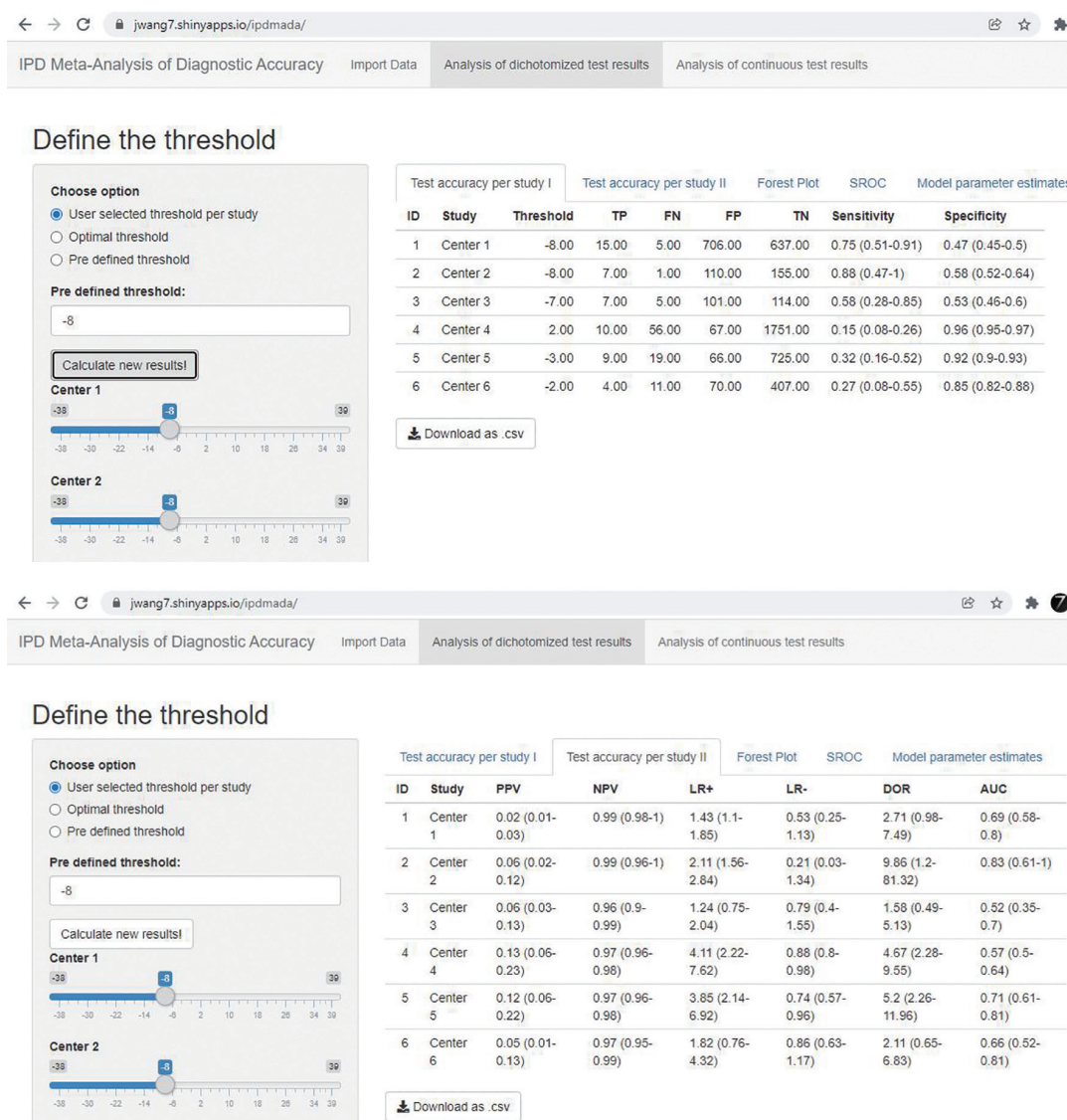


Figure 6 Test performance based on confusion matrix. IPD, individual patient data; TP, true positive; FN, false negative; FP, false positive; TN, true negative; PPV, positive predictive value; NPV, negative predictive value; LR+, positive likelihood ratio; LR-, negative likelihood ratio; DOR, diagnostic odds ratio; AUC, area under the curve.

data visualization purposes, to have a better understand how sensitivity and specificity will change when the threshold changes.

Numbers in the confusion matrix (i.e., TP, FN, FP, TN) and the corresponding test performance will be presented/updated after clicking the “Calculate new results!” button (Figure 6). Forest plots of sensitivity and specificity are also provided in the “Forest Plot” sheet (results not shown).

Calculation of test performance based on distribution of test results

Before doing any calculation, the distribution of test results is visualized with a ridgeline plot (Figure 7). This ridgeline plot shows the distribution of index test in diseased group (Present) and non-diseased group (Absent), and smaller overlap between the two groups indicates better discrimination power of the index test.

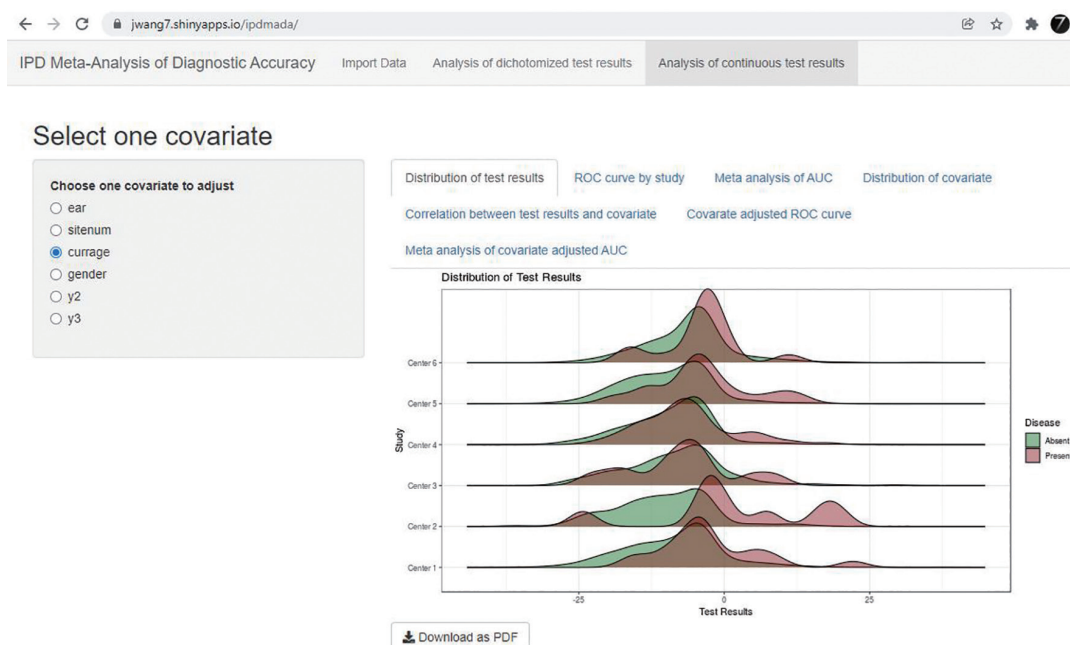


Figure 7 Distribution of test results in each center. IPD, individual patient data; ROC, receiver operating characteristic; AUC, area under the curve.

ROC curve of the test results is plotted for each center (Figure 8A), and the corresponding AUCs are summarized in a forest plot (Figure 8B). ROC curve shows the relation between TP rate/sensitivity (y-axis) and FP rate/(1-specificity) (x-axis) at all possible threshold values.

Advanced analyses of covariates are also provided (7). The distribution of a continuous covariate is plotted with a ridgeline plot (Figure 9A), while the distribution of a categorical covariate is plotted as a bar chart (Figure 9B). Investigating the distributions of covariates can help to identify differences in patient characteristics in diseased and non-diseased groups, which may influence the true test performance. If substantial differences are observed, we can consider using covariate adjusted ROC curve for further analysis.

In the end, covariate adjusted ROC curve of the test results is plotted for each center (Figure 10A), and the corresponding covariate AUCs are summarized in a forest plot (Figure 10B).

Analyses of data from a single center for multiple tests

Data preparation

If we want to analyze data of multiple tests from a single

center, it can be done by changing the center name in Column “Study” to the test name.

We show this in a hypothetical example, where we treat tests performed on left ears and right ears as two different tests, and want to evaluate these two tests in one study. Please note that, the study design (i.e., paired data, unpaired data or randomized data) cannot be reflected in the analyses, and the numbers of patients underwent each test can be different. This is a limitation of using IPDmada in this situation: diagnostic accuracy performance measures can be calculated separately for each index test, however a formal comparison with statistical test between index tests is not supported.

We can start from the data shown in Figure 4. Add a formula [=“Test”&B2] in cell A2 (Figure 11), and double click on the *plus sign* (+) in the bottom right corner of cell A2 to copy the formula to all the rest cells in Column A (results not shown). After doing this, save the data as a CSV file.

Another example is, if the two (or more) tests being analyzed are saved in wide format (i.e., saved in two separate columns like y2 and y3), we need to transform the wide format data into long format and rename the column of test type as “Study”.

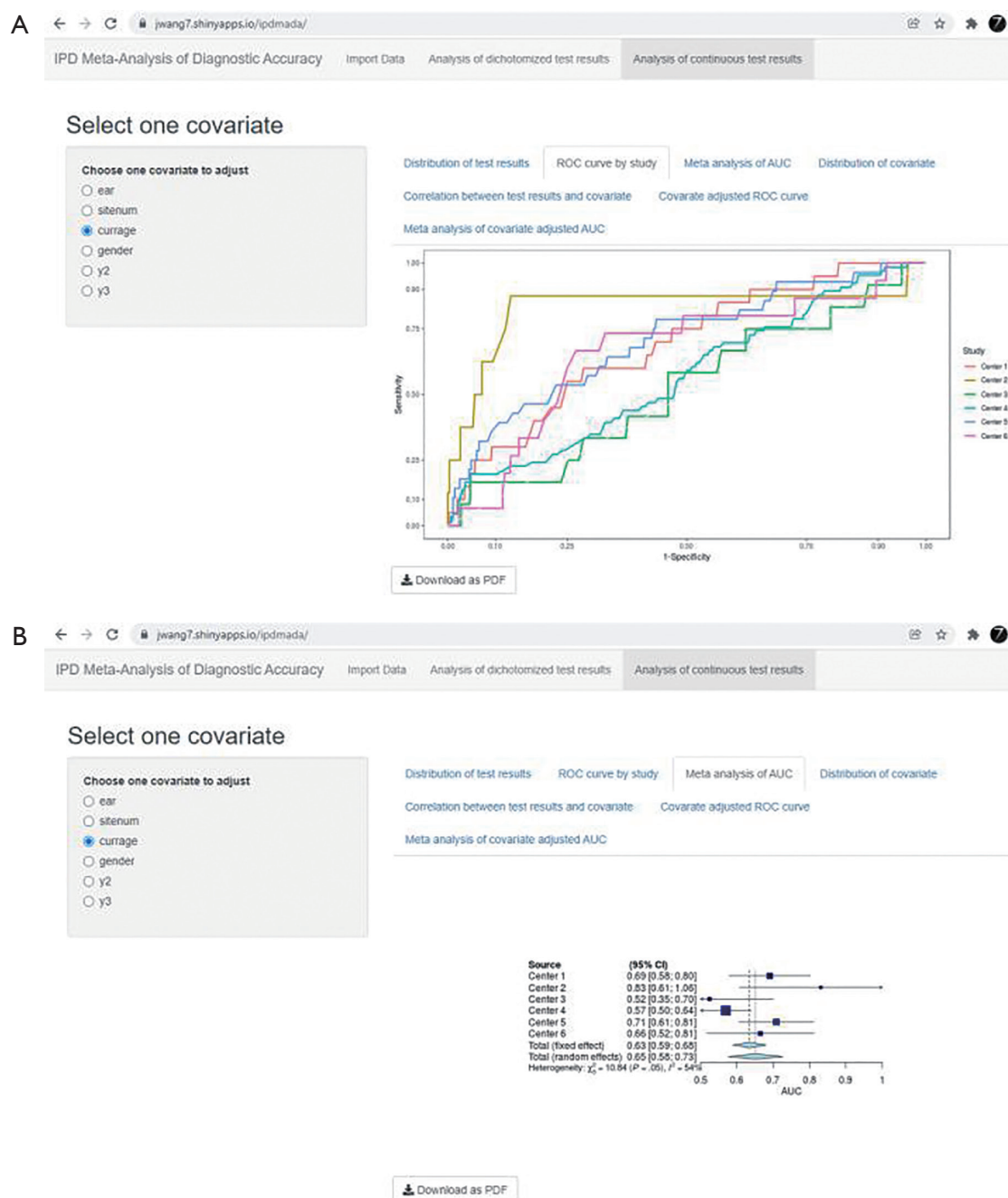


Figure 8 ROC curves (A) and forest plot (B) of AUCs. IPD, individual patient data; ROC, receiver operating characteristic; AUC, area under the curve; CI, confidence interval.

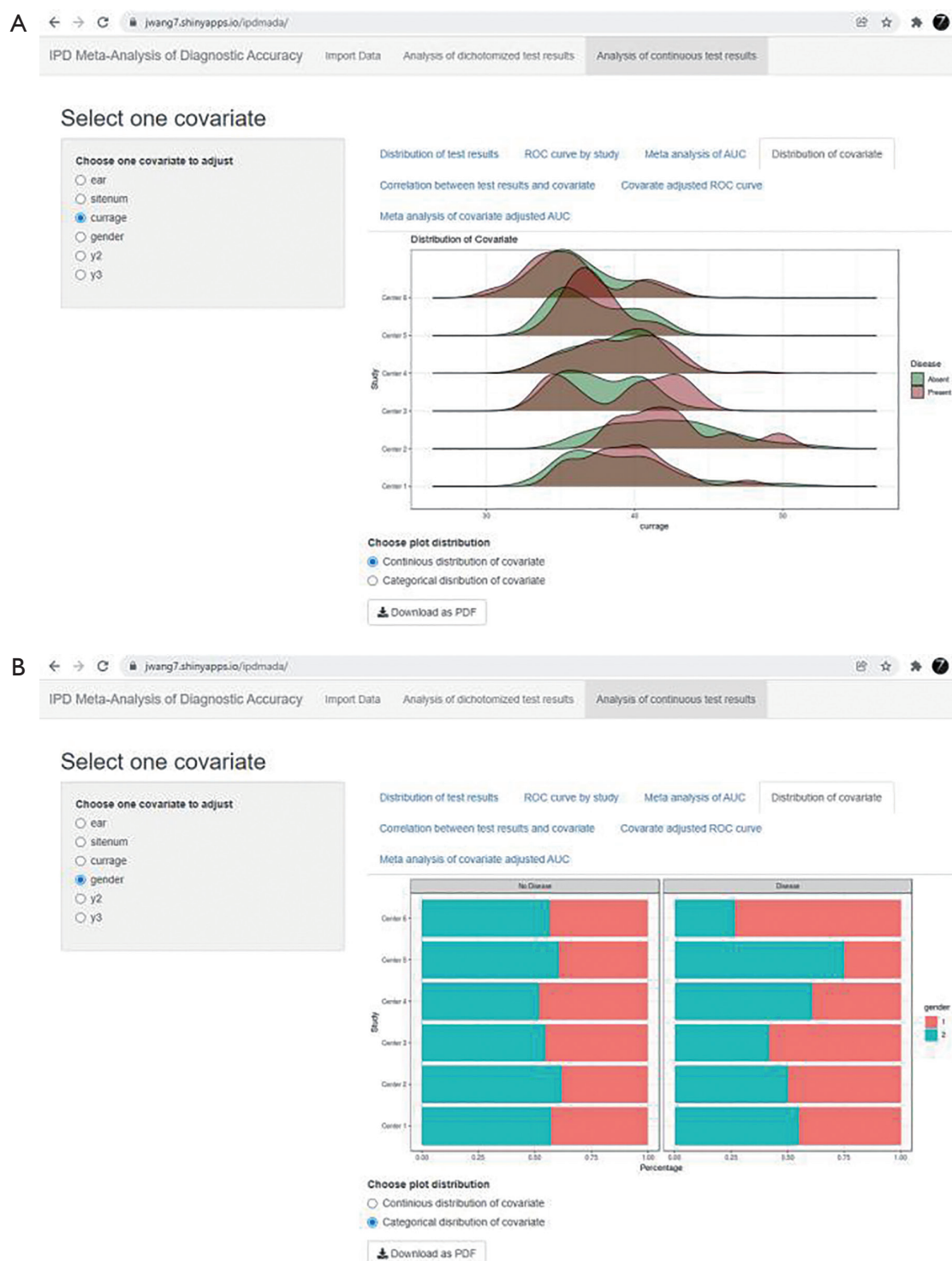


Figure 9 Distributions of corrected age (A) and gender (B) in each center. IPD, individual patient data; ROC, receiver operating characteristic; AUC, area under the curve.

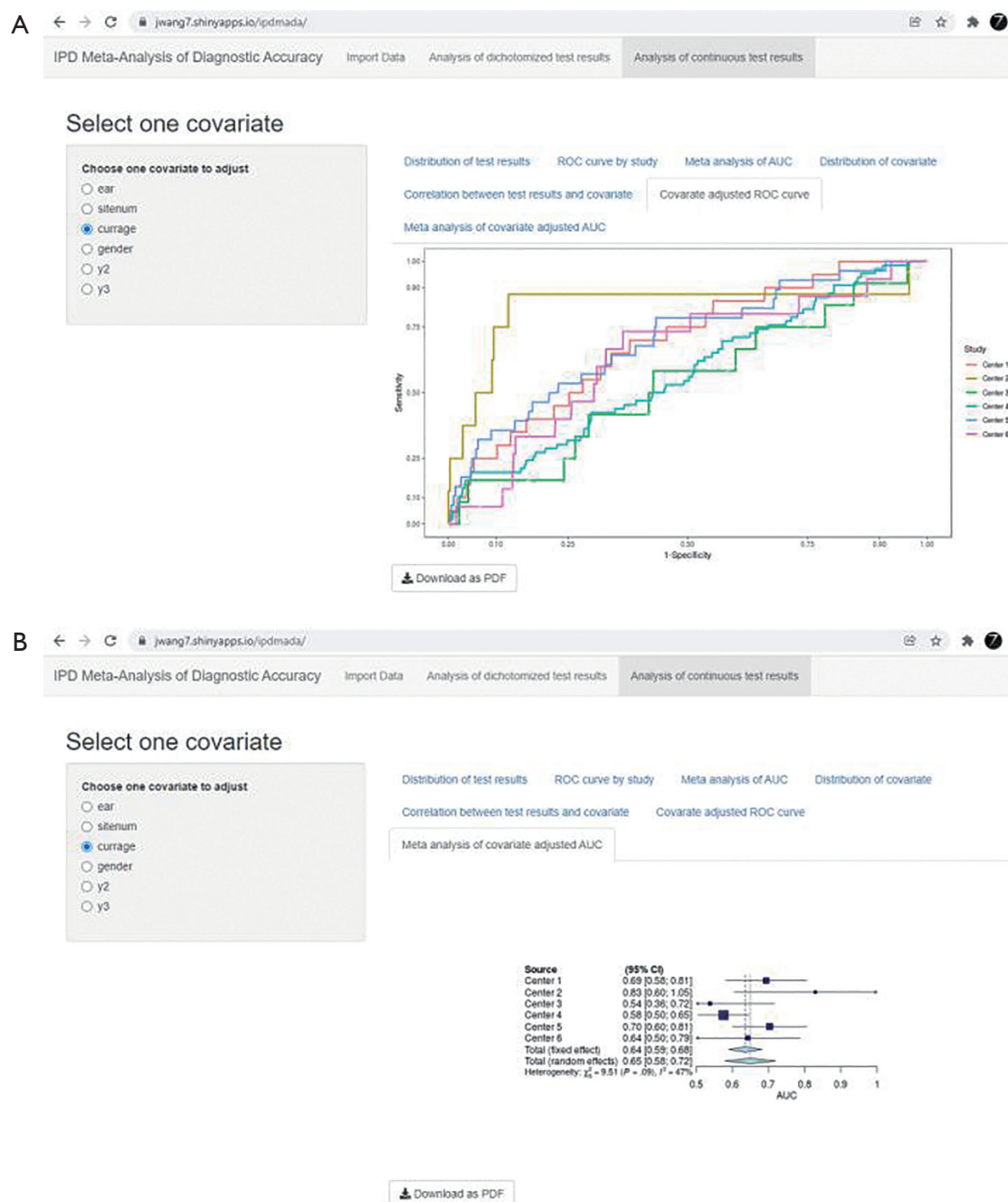


Figure 10 Covariate adjusted ROC curves (A) and forest plot (B) of covariate adjusted AUCs. IPD, individual patient data; ROC, receiver operating characteristic; AUC, area under the curve; CI, confidence interval.

	A	B	C	D	E	F	G	H	I
1	Study	ear	sitenum	currage	gender	disease	test.results	y2	y3
2	=Test "&	2	1	42.42	2	0	-3.1	-9	-1.5
3	B2	1	1	42.42	2	0	-4.5	-8.7	-2.71
4		2	1	40.14	2	1	-3.2	-13.2	-2.64
5		1	1	38.14	1	0	-22.1	-7.8	-2.59

Figure 11 Insert test names as Study ID.

Analyses results

Analyses results are similar to those presented in multiple centers for one test, thus not shown again.

Discussion

IPDmada is a powerful tool, which can produce all analyses in a DTA study in a very efficient way. The time needed for data preparation and analyses is significantly reduced, and more importantly, less statistical knowledge is required from the researchers. All the analyses are facilitated by R, but the installation of R or running R locally is not needed. Compared to other statistical software, IPDmada is code-free and cost-free.

There are some limitations of IPDmada. First, the analyses provided by IPDmada are originally designed for IPD-MA, thus some of them may not be suitable for a primary DTA study. Second, when evaluating multiple index tests, the study design is not reflected and comparison between tests is not available. Third, IPDmada only provides standard analyses and has less flexibility. Last, all data preparation is done in a spreadsheet, which cannot be as convenient as statistical software.

In conclusion, IPDmada is an easy-to-use tool, which can provide a one-stop solution to researchers without strong statistical background for all the analyses needed in a DTA study.

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Footnote

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Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved.

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