

Identification of *Mycobacterium abscessus* subspecies by MALDI-TOF Mass Spectrometry and Machine Learning

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25 **ABSTRACT**

26 *Mycobacterium abscessus* complex is one of the most common and pathogenic
27 nontuberculous mycobacteria (NTM) isolated in clinical laboratories. It consists of three
28 subspecies: *M. abscessus* subsp. *abscessus*, *M. abscessus* subsp. *bolletii* and *M.*
29 *abscessus* subsp. *massiliense*. Due to their different antibiotic susceptibility pattern, a
30 rapid and accurate identification method is necessary for their differentiation. Although
31 matrix assisted laser desorption/ionization-time of flight mass spectrometry (MALDI-
32 TOF MS) has proven useful for NTM identification, the differentiation of *M. abscessus*
33 subspecies is challenging. In this study, a collection of 244 clinical isolates of *M.*
34 *abscessus* complex was used for MALDI-TOF MS analysis and for the development of
35 machine learning predictive models. Overall, using a Random Forest model with
36 several confidence criteria (samples by triplicate and similarity values >60%), a total of
37 95.8% of isolates were correctly identified at subspecies level. In addition, differences
38 in culture media, colony morphology and geographic origin of the strains were
39 evaluated, showing that the latter most affected the mass spectra of isolates. Finally,
40 after studying all protein peaks previously reported for this complex, two novel peaks
41 with potential for subspecies differentiation were found. Therefore, machine learning
42 methodology has proven to be a promising approach for rapid and accurate
43 identification of subspecies of the *M. abscessus* complex using MALDI-TOF MS.

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52 INTRODUCTION

53 Nontuberculous mycobacteria (NTM) are a group of mycobacteria present in the
54 environment that, in some cases, can cause different types of infections in humans,
55 such as pulmonary infections, skin and soft tissue infections and disseminated
56 infections (1). *Mycobacterium abscessus* complex is one of the most common and
57 pathogenic NTM isolated in clinical laboratories, causing respiratory infections, even in
58 patients with cystic fibrosis (2). *M. abscessus* complex contains three subspecies: *M.*
59 *abscessus* subsp. *abscessus*, *M. abscessus* subsp. *massiliense* and *M. abscessus*
60 subsp. *bolletii* (3). Hereafter, they will be referred as *M. abscessus*, *M. massiliense* and
61 *M. bolletii*, respectively.

62 The three subspecies show different susceptibility to clarithromycin, a decisive
63 antibiotic for the treatment of these infections. Thus, *M. bolletii* and most strains of *M.*
64 *abscessus* shows resistance to clarithromycin, whereas *M. massiliense* is susceptible
65 (4). The different antibiotic susceptibility pattern, in addition to the recommendation of
66 the American Thoracic Society and Infectious Diseases Society of America
67 (ATS/IDSA) to identify NTMs at species level, makes it necessary to implement novel
68 approaches for rapid and accurate discrimination of these three subspecies (5).

69 Currently, *M. abscessus* complex subspecies can only be identified by
70 molecular methods, such as a commercial kit based on PCR-reverse hybridization (6)
71 or by multiple gene sequencing (*hsp65*, *rpoB*, *erm(41)*, etc.) (7, 8). On the other hand,
72 the use of Matrix Assisted Laser Desorption/Ionization-Time of Flight Mass
73 Spectrometry (MALDI-TOF MS) allows the reliable identification of most NTMs and has
74 become the main identification method in several clinical laboratories (9, 10). However,
75 differentiation of closely related species (like *M. abscessus* complex subspecies)
76 remains a challenge. Although some studies have attempted subspecies identification
77 by protein peak analysis (11-16), there is no consensus on the best strategy to follow.
78 In last years, new approaches for data analysis from MALDI-TOF mass spectra have

79 been applied, such as machine learning methods, which has the potential to get
80 additional information than simple species identification (17).

81 The aim of this study was to evaluate MALDI-TOF MS and Machine Learning
82 algorithms for the differentiation of *M. abscessus* complex subspecies. This study
83 represents the first proof of concept for the identification of these species by applying
84 MALDI-TOF MS and Machine Learning.

85

86 **MATERIALS & METHODS**

87 **Mycobacterial isolates**

88 A total of 244 clinical isolates of *M. abscessus* complex obtained from 152
89 different patients were included in this study. They encompassed 119 *M. abscessus*,
90 84 *M. massiliense* and 41 *M. bolletii* isolates. All the isolates were obtained from
91 Hospital General Universitario Gregorio Marañón (HGM; Madrid, Spain), Hospital
92 Universitario La Princesa (HLP; Madrid, Spain), Instituto de Salud Carlos III-Centro
93 Nacional de Microbiología (ISCIII; Majadahonda, Spain) and Hospital Universitari de
94 Bellvitge (HUB; Hospitalet de Llobregat, Spain). All isolates are described in Table S1.

95 **Bacterial cultures and protein extraction procedure**

96 All isolates were previously identified by PCR-reverse hybridization (GenoType
97 NTM-DR, Hain Lifescience, Nehren, Germany) or whole genome sequencing. All HGM,
98 HLP and ISCIII isolates were cultured from frozen stocks on 7H11 agar plates until
99 growth was observed. Among HUB isolates, 38 were cultured on 7H11 agar plates and
100 48 on Löwenstein-Jensen (BioMérieux, Marcy l'Etoile, France) media. In all cases, the
101 isolates were incubated at 37°C until growth was observed (4-7 days). The protein
102 extraction procedure for MALDI-TOF MS analysis was performed as previously
103 described (10). First, a 1 µl loopful of biomass was suspended in 300 µl of High-

104 Pressure Liquid Chromatography (HPLC) quality water, and then heat inactivated in a
105 dry bath at 95°C during 30 min. After this, 900 µl of ethanol were added, the tubes
106 were centrifuged at 13,000 rpm for 2 min and the supernatant was discarded. After
107 centrifuge and discard the supernatant again, the pellet was dried at room temperature.
108 Then, 0.5 mm silica/zirconia beads were added together with 10 µl of acetonitrile. The
109 tubes were vortexed briefly and sonicated for 15 min. After sonication, 10 µl of formic
110 acid were added, the tubes were vortexed for 10 s and centrifuged at 13,000 rpm for 2
111 min. One microliter of the supernatant was deposited onto the MALDI target plate
112 (Bruker Daltonics, Bremen, Germany) in triplicates, allowed to dry and covered with 1
113 µl of α-cyano-4-hydroxycinnamic acid (HCCA).

114 **Spectra acquisition by MALDI-TOF MS and data processing**

115 Acquisition of protein spectra was performed using the MBT Smart MALDI
116 Biotyper (Bruker Daltonics) in the range of 2,000-20,000 Da. All spots were read three
117 times, resulting in 9 protein spectra per isolate. The spectra were exported and
118 processed with Clover MS Data Analysis software (Clover Biosoft, Granada, Spain).
119 The processing pipeline consisted on: 1) Smoothing by Savitzky-Golay filter (window
120 length=11, polynomial order=3); 2) Baseline subtraction by Top-Hat filter
121 (factor=0.02); 3) Alignment of spectra with 2 Da of constant tolerance and 300 ppm of
122 linear mass tolerance; and 4) Normalization by Total Ion Current (TIC).

123 **Predictive models and external validation**

124 Once the spectra were processed, unsupervised –Principal Component
125 Analysis (PCA), Hierarchical Cluster Analysis (HCA)- and supervised –Partial Least
126 Squares Discriminant Analysis (PLS-DA), Support Vector Machine (SVM), Random
127 Forest (RF) and K-Nearest Neighbors (KNN)- algorithms were applied for the creation
128 of predictive models. A total of 43 isolates (20 *M. abscessus*, 15 *M. massiliense* and 8
129 *M. bolletii*) collected in HGM, ISCIII and HUB were included in the test set for the

130 creation of the predictive models, they represented a total of 539 mass spectra. These
131 isolates were randomly selected in order to represent all the variability observed
132 previously (subspecies, morphology, culture media and geographical origin). *M.*
133 *massiliense* and *M. bolletii* spectra were balanced by oversampling in order to obtain
134 the same number of spectra for each category. Internal validation was performed by
135 10-fold cross validation. For external validation, 201 isolates collected in all centres
136 were used (99 *M. abscessus*, 69 *M. massiliense* and 33 *M. bolletii*), and the
137 identification obtained in each of the three spots used was considered.

138 **Ethics statement**

139 The Ethics Committee of the Gregorio Marañón Hospital (CEIm) evaluated this
140 project and considered that all the conditions for waiving informed consent were met,
141 since the study was conducted with microbiological samples and not with human
142 products.

143

144 **RESULTS**

145 **Analysis of isolates by unsupervised algorithms**

146 After the analysis of all isolates by PCA, two main clusters were observed
147 (Figure 1). Different variables included in the study that may influence on the mass
148 spectra were examined: the *M. abscessus* complex subspecies, the morphology of the
149 colonies, the type of culture media and the geographical origin of the isolates. As can
150 be observed in Figure 1, the two main clusters corresponded to the geographical origin
151 of the isolates, separating those collected in Madrid hospitals (HGM, HLP and ISCII)
152 from those collected in Barcelona (HUB). On the other hand, isolates from different
153 subspecies and different morphology overlapped in both clusters, as well as isolates
154 from HUB, which were analyzed in two different culture media.

155 **Analysis of isolates by supervised algorithms**

156 Internal validation of predictive models

157 The results for each algorithm after applying a 10-fold cross validation are
158 showed in Table 1. The algorithms PLS-DA (Figure S1), SVM and RF showed the
159 same accuracy (99.8%), with only one spectrum of the 539 misclassified as other
160 category (Table S2), while KNN was the algorithm with lower accuracy (Figure S1).

161 External validation of predictive models

162 Blind analysis of the 201 isolates used for external validation showed that PLS-
163 DA and KNN produced low identification rates, while RF algorithm yielded 89.9%
164 correct classification (Table 1). With this algorithm, *M. bolletii* obtained the lower
165 identification rate, with 21 spectra misclassified (Table S3). Due to RF obtained the
166 highest identification rate (Figure 2A), its results were further analyzed. Among the
167 three identifications obtained in each spot for each isolate, the subspecies obtained in
168 at least 2 spots was considered as the final identification. A total of 184 (91.5%) of the
169 isolates obtained uniform identification results (Table 2), and only one isolate obtained
170 different subspecies identification for each spot. Among the isolates with identical
171 identification, the accuracy rate of subspecies level identification was higher than in
172 those with only two matching identifications. Moreover, the probability of correct
173 identification provided by RF was evaluated in order to establish a confidence cut-off.
174 For all subspecies, 172 (85.6%) isolates obtained a probability higher than 60% (Figure
175 2B), so this cut-off was proposed for confident result. Considering the categorical
176 result, 89.6% of isolates were correctly identified at subspecies level, while establishing
177 the confidence cut-off at 60% of probability, the accuracy rate increased to 95.3%
178 (Figure 2C). Moreover, when both parameters were considered (same identification in
179 three spots and confidence higher than 60%), out of the 170 isolates that met these
180 criteria, 163 (95.8%) were correctly identified (Table 2). The three subspecies

181 performed similar, with similar Area Under the Curve (AUC) between them (Figure 2D).
182 Finally, Positive Predictive Values (PPV) were evaluated for each subspecies.
183 Considering all identification results, the PPV obtained was 89.3% for *M. abscessus*,
184 89.3% for *M. bolletii* and 90.0% for *M. massiliense*. When we considered only those
185 isolates with a probability result higher than 60%, the PPV increased to 96.4%, 91.7%
186 and 95.3% for each subspecies, respectively (Figure 2E).

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188 **Specific peak analysis for subspecies discrimination**

189 All protein peaks reported in previous studies were searched among the
190 analyzed isolates (Table 3). No unique subspecies specific peaks were found, although
191 some of them were present in most strains of certain subspecies. Thus, almost all *M.*
192 *abscessus* isolates showed peaks at 2081, 3378 and 7637 *m/z*; *M. bolletii* showed
193 2081, 3123, 3463 and 7637 *m/z*; and *M. massiliense* showed 3378, 4385 and 6711
194 *m/z*. In addition, two novel potential peaks were found in this study: 2673 *m/z*, which
195 was present in 88.9% of *M. abscessus* isolates, 17.1% of *M. bolletii* and 7.3% of *M.*
196 *massiliense*; and 6960 *m/z*, which was present in 90.5% of *M. abscessus* isolates,
197 9.8% of *M. bolletii* and 26.0% of *M. massiliense* (Table 3).

198

199 **DISCUSSION**

200 Differentiation of *M. abscessus* complex subspecies by MALDI-TOF MS has
201 been attempted in previous studies using conventional peak analysis (11-16).
202 However, many variables can hinder this objective, such as the culture media used, the
203 morphology of the colonies or the geographic origin of the strains (14, 18). Moreover,
204 due to the large number of protein peaks that are usually found in mass spectra,
205 accurate identification based on only a few peaks may not be entirely reliable.

206 Therefore, it is necessary to apply novel strategies capable of analyzing a large amount
207 of data, such as machine learning methodologies (17, 19).

208 In the present study, we applied machine learning using both unsupervised and
209 supervised algorithms. The first approach by unsupervised methodology (PCA) did not
210 provide subspecies differentiation (Figure 1). In the case of morphology, no important
211 spectral differences between smooth and rough variants were observed. The main
212 difference of these two variants is the expression of glycopeptidolipids on the surface
213 (20, 21), and due to MALDI-TOF MS analyzes mainly ribosomal proteins, these
214 differences were not detected. On the other hand, because this is a proof of concept
215 and identification of mycobacteria from liquid media could be more complex (22), only
216 solid culture media were evaluated, and differences between 7H11 and Löwenstein-
217 Jensen were not observed. Interestingly, the two main clusters obtained by PCA
218 corresponded to the geographic origin of the isolates. All strains obtained from the
219 three Madrid hospitals (HGM, HLP and ISCIII) grouped together, while those from
220 Barcelona (HUB) were separated. All isolates from Madrid were analyzed in the same
221 hospital (HGM) and Barcelona isolates in HUB by the same operator, so differences in
222 experience and preparation of the protein extracts were discarded. In addition, the
223 MALDI-TOF MS model in both centers was the same (MBT Smart Biotype) and the
224 acquisition of spectra was performed with the same technical parameters, so the
225 influence of the instruments was minimal. Therefore, these results may suggest that
226 differences in mass spectra of *M. abscessus* complex from different origins is greater
227 than expected, and highlights the importance of include strains from diverse origins in
228 this type of studies.

229 The application of supervised machine learning algorithms was targeted to
230 differentiation of the three subspecies. Among the four algorithms tested, the lower
231 results were obtained by PLS-DA (Figure S2A), SVM (Figure S2B) and KNN (Figure
232 S2C), while Random Forest was able to identify a greater number of isolates at

233 subspecies level (Table 1). As recommended by other studies, the identification of
234 NTM by MALDI-TOF MS should be performed in 2 or 3 replicates (10), so for more
235 accurate identifications we used three spots for each isolate. When the identification of
236 the three spots was considered, the accuracy was higher in those cases where the
237 same subspecies was obtained in all spots (Table 2). The categorical result of RF is
238 accompanied by a probability result, so we aimed to establish a confidence cut-off in
239 order to reach higher accuracy of identification. Without applying probability cut-off, RF
240 correctly identified 89.9% of the isolates. Most of isolates (172; 85.6%) obtained
241 probability results above 60% (Figure 2B), so when the cut-off was established at 60%,
242 the accuracy rate increased to 95.3% (Figure 2C). Moreover, by applying this cut-off,
243 the PPV for all subspecies increased to higher than 90% (Figure 2E). Among the 170
244 isolates that met the criteria of obtaining the same identification in all spots and a
245 confidence higher than 60%, a total of 95.8% were correctly classified, with 7 isolates
246 misidentified: 2 *M. abscessus* identified as *M. massiliense*, 2 *M. massiliense* identified
247 as *M. abscessus*, 2 *M. massiliense* as *M. bolletii* and 1 *M. bolletii* identified as *M.*
248 *abscessus*.

249 In order to analyze the protein peaks found in this study, all previously reported
250 peaks were searched and compared with our isolates. In most cases, the detection rate
251 of the peaks was similar to previous reports (Table 3) and, in addition, the most
252 important peaks were found in the range of 2000-10000 Da (Figure 2F). However,
253 remarkable discrepancies in few cases were found. Some peaks were found with a
254 lower presence than reported previously: this is the case of the 4391 m/z peak (Figure
255 S3A) in *M. abscessus* and *M. bolletii*, with only half of our isolates presenting it; the
256 peak around 8782 m/z (Figure S3B) that was present in 61% of *M. bolletii* isolates in
257 comparison with 100% reported by Suzuki et al. (12) and Kehrmann et al. (15); and the
258 peak around 7667 m/z (Figure S3C) in *M. massiliense* that was found in 47.1% of our
259 isolates. Strikingly, peaks 3354 and 8508 m/z were found only in a few *M. massiliense*

260 isolates, while they were previously reported in most isolates of this subspecies (13,
261 14). On the other hand, peaks that were reported as absent in some subspecies, were
262 found in some of our isolates. That was the case of 3108 (Figure S3D) and 4385 m/z
263 (Figure S3A) in *M. abscessus* and *M. bolletii*; 3123 m/z (Figure S3D) in *M. massiliense*;
264 3378 m/z in *M. bolletii* (Figure 3A); 3463 m/z (Figure S3E) in *M. abscessus* and *M.*
265 *massiliense*; and 6711 m/z (Figure S3F) in *M. abscessus*. The greater differences were
266 in peaks 2081 (Figure 3B) and 7637 m/z, which have never been reported in *M.*
267 *massiliense* (12, 16) and we found them in more than 50% of *M. massiliense* isolates.
268 All these differences could have been influenced by two factors. First, it is important to
269 include a high number of strains, representing the three subspecies in order to confirm
270 that the peaks found are specific to them. The second factor is the geographic origin of
271 the isolates. There have been reported differences in peak patterns according to the
272 origin of the strains (14), so multicentric studies are needed to search common peaks
273 worldwide and create accurate identification algorithms. On the other hand, two novel
274 potential peaks have been found: 2673 (Figure 3C) and 6960 m/z (Figure 3D), both of
275 them present in most *M. abscessus* isolates and in low number of isolates from the
276 others subspecies.

277 Due to the variability in the detection of peaks observed previously, the present
278 study showed that the application of novel methodologies for data analysis, such as
279 machine learning, could be an innovative way for improve the MALDI-TOF MS
280 accuracy on identifying *M. abscessus* subspecies. Recently, other novel strategies
281 have been evaluated for the same purpose. Khor et al. used the MALDI Biotyper Sirius
282 system (Bruker Daltonics) for the detection of subspecies-specific lipids, and was able
283 to differentiate a few *M. abscessus* complex isolates (23). On the other hand, Bajaj et
284 al. evaluated for the first time the Liquid Chromatography-Mass Spectrometry for
285 identification of *M. abscessus* subspecies (24). However, these novel methods need to

286 be validated with larger collections of clinical isolates to confirm their utility in a
287 microbiology laboratory setting.

288 In conclusion, the high correct identification rate of *M. abscessus* complex
289 subspecies obtained in this study, states the utility of machine learning strategy for
290 identification purposes. This method could be further refined in near future by the
291 addition of a greater number and diversity of isolates.

292

293 **Author contributions**

294 DRT: conceptualization, experimentation, formal analysis, data collection, validation,
295 visualization, original draft preparation and review/editing. LH, FA, DD, NV, PM, MJRS:
296 submission of isolates, writing and review/editing. MJA GM, LM: data analysis,
297 validation, writing and review/editing. BRS: conceptualization, project administration,
298 formal analysis, supervision, validation, visualization, original draft preparation and
299 review/editing.

300

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310

311 **Conflicts of interest**

312 The authors declare no conflict of interests. MJA, GM and LM are employees of Clover
313 Bioanalytical Software, S.L.

314

315 **Figure legends**

316 **Figure 1.** Principal Component Analysis of all isolates included in the study, colored
317 according to different characteristics: **A**, comparison of *M. abscessus* complex
318 subspecies; **B**, comparison of colony morphology; **C**, comparison of culture media; **D**,
319 comparison of geographical zone of origin.

320 **Figure 2.** Analysis of mass spectra by Random Forest (RF) algorithm. **A.** RF plot of the
321 model. **B.** Percentages of identification probably obtained by RF on validation isolates..
322 **C.** Number of correctly identified isolates according to probability cut-off obtained by
323 RF. **D.** ROC and Precision Recall curves for validation isolates by RF. **E.** Total Positive
324 Predictive Value (PPV) for RF results and PPV using a 60% probability cut-off. **F.**
325 Feature importances of mass peaks for RF model.

326 **Figure 3.** Novel potential protein peaks reported in the present study and other
327 relevant peaks. **A.** 3,378 m/z. **B.** 2,081 m/z. **C.** 2,673 m/z. **D.** 6,960 m/z.

328

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Algorithm	<i>M. abscessus</i>	<i>M. bolletii</i>	<i>M. massiliense</i>	Total
10-fold cross validation				
PLS-DA	99.4%	100%	100%	99.8%
SVM	99.4%	100%	100%	99.8%
RF	100%	100%	99.4%	99.8%
KNN	79.9%	97.2%	92.2%	89.8%
External validation				
PLS-DA	58.2%	80.8%	73.9%	67.3%
SVM	87.9%	77.8%	90.8%	87.2%
RF	92.3%	78.8%	91.8%	89.9%
KNN	45.1%	23.2%	61.8%	47.3%

433 **Table 1.** Accuracy results for internal 10-fold cross validation and external validation.

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RF identification	N isolates (%)	N correct (%)
Same ID in 3 spots	184 (91.5%)	173 (94.0%)
Same ID in 2 spots	16 (8.0%)	7 (43.8%)
Three different ID	1 (0.5%)	0 (0%)
Isolates with confidence >60%	172 (85.6%)	164 (95.3%)
Same ID (3 spots) and confidence >60%	170 (84.6%)	163 (95.8%)

436 **Table 2.** Random Forest (RF) accuracy according to the identification (ID) obtained in

437 each spot.

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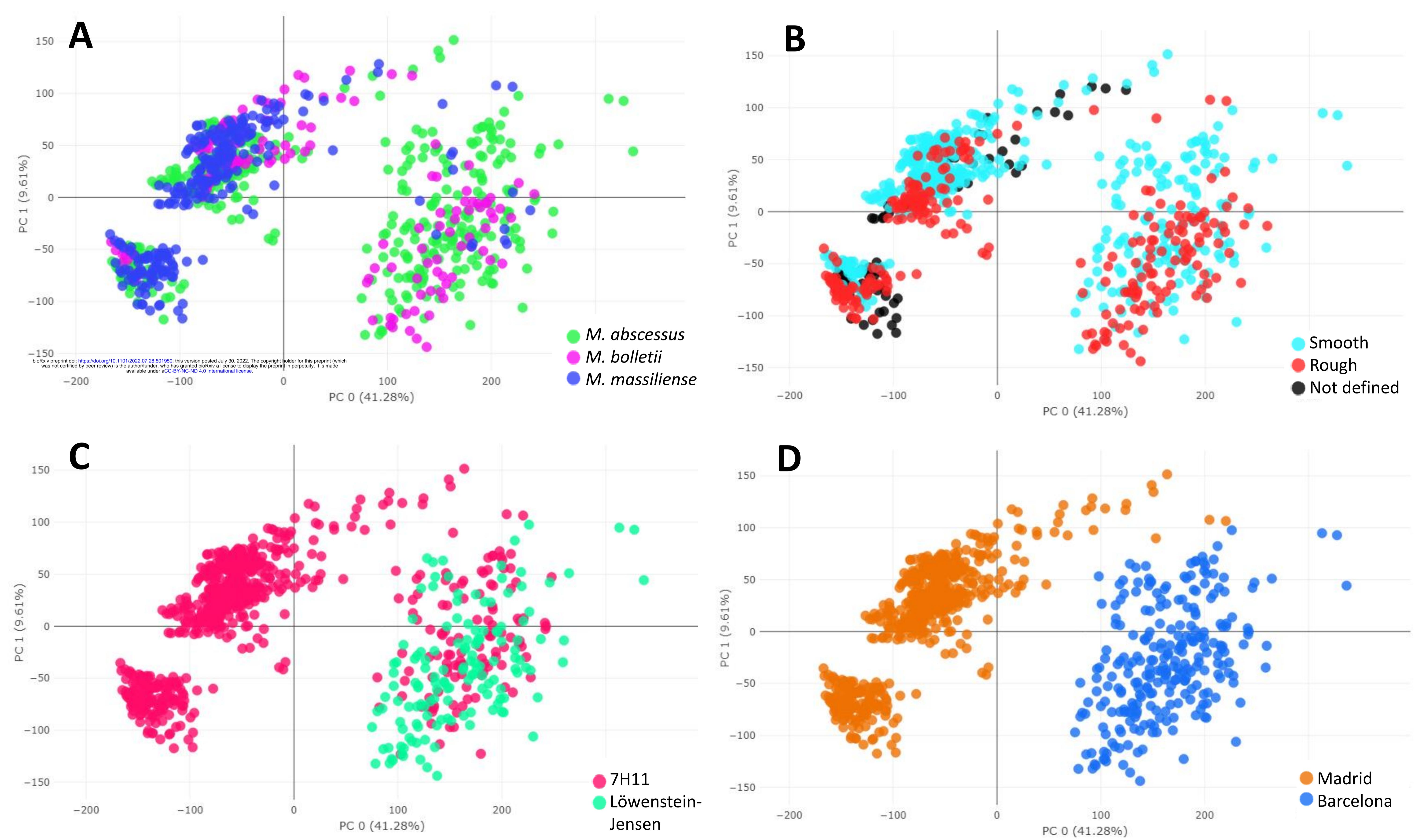
<i>m/z</i>	<i>M. abscessus</i>		<i>M. bolletii</i>		<i>M. massiliense</i>	
	Present study (%)	Previous studies (%)	Present study (%)	Previous studies (%)	Present study (%)	Previous studies (%)
Previously reported peaks						
2081	96.0	31.7-96.5	100	100	53.1	0
3108	10.3	0	17.1	0	79.2	100
3123	84.1	100	95.1	100	9.4	0
3354	4.0	1.4	12.2	NA	3.1	100
3378	99.2	100	36.6	0	97.9	100
3463	23.8	0	90.2	66.7	31.2	0
4385	31.7	0-1.4	29.3	0	89.6	89.5-100
4391	53.2	98.6-100	41.5	100	9.4	0-5.2
6711	32.5	0	73.2	NA	90.6	100
7637	92.6	93.2-100	90.2	100	62.5	0
7667	11.9	3.4	2.4	0	41.7	88.1-100
8508	7.1	4.9	14.6	NA	9.4	84.6
8768	2.4	0-0.7	7.3	0	70.8	38.4-100
8782	72.2	89.2-100	61.0	100	6.2	0-2.8
9475	78.6	17-100	73.2	100	16.7	0-9.5
Novel potential peaks						
2673	88.9	NA	17.1	NA	7.3	NA
6960	90.5	NA	9.8	NA	26.0	NA

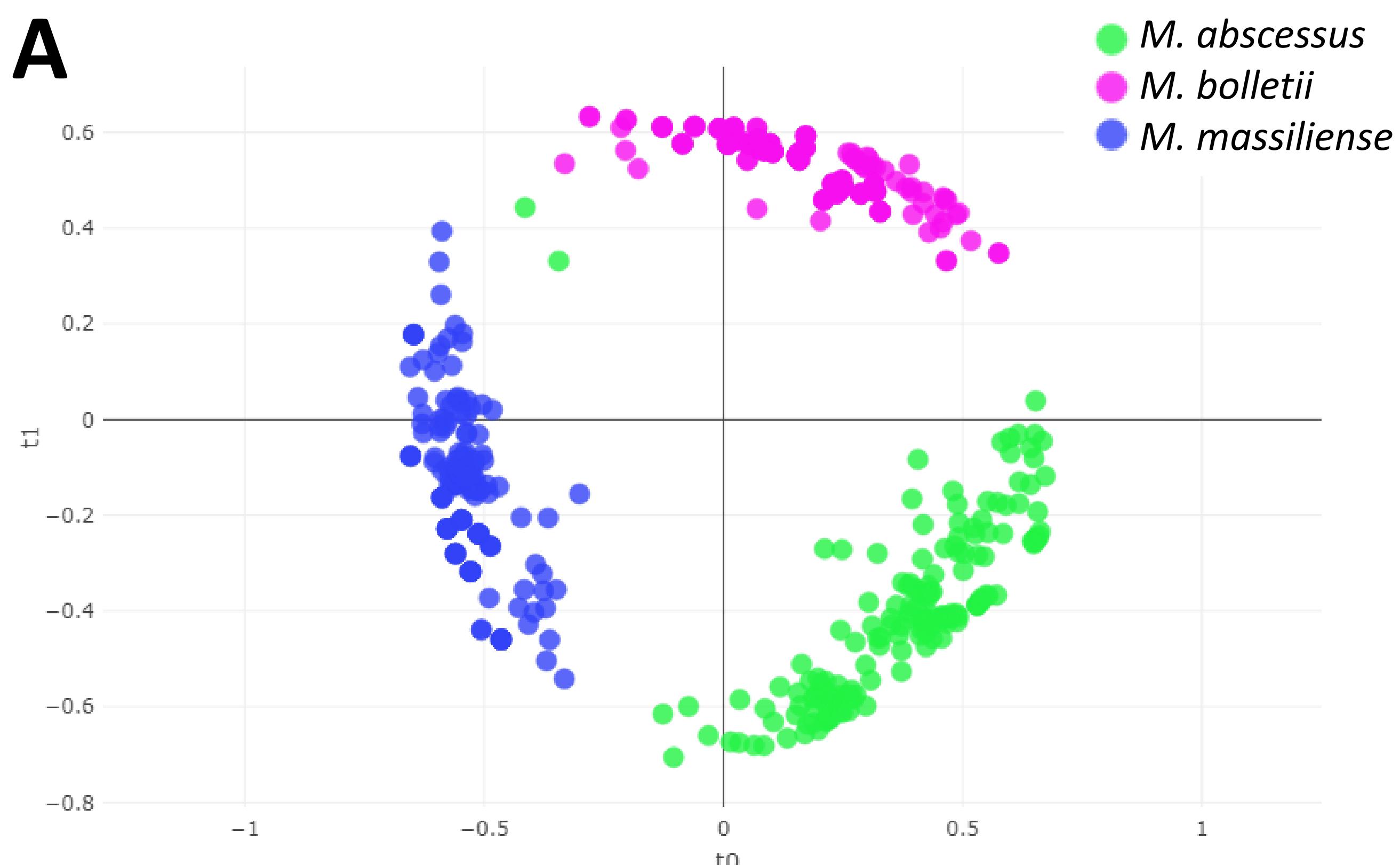
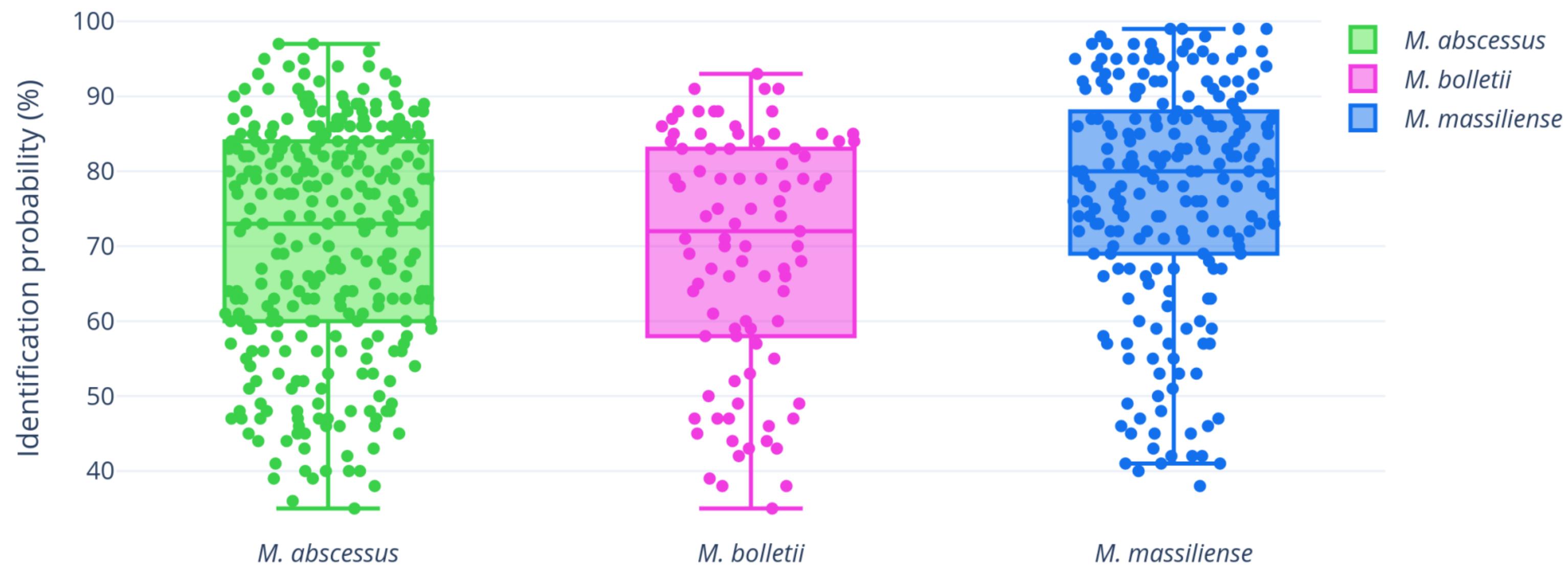
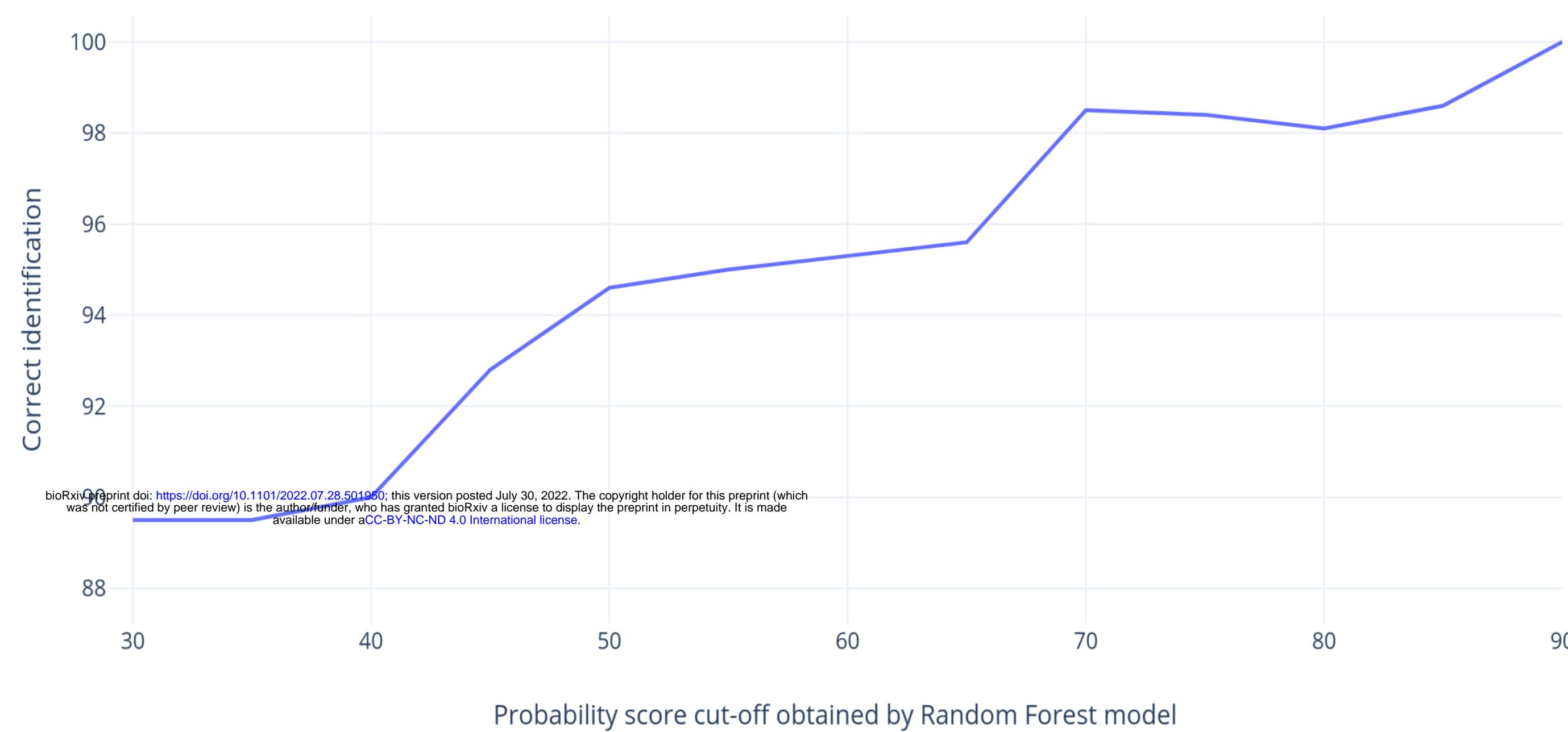
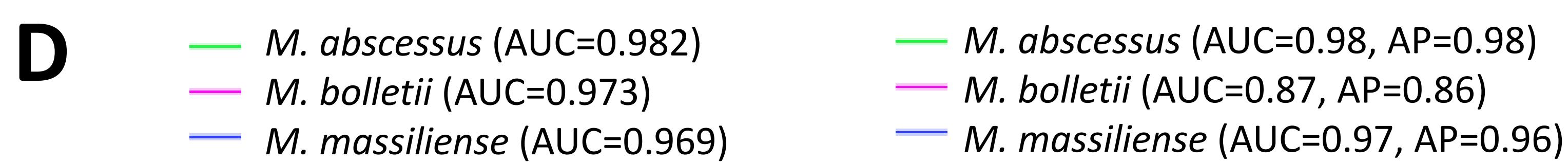
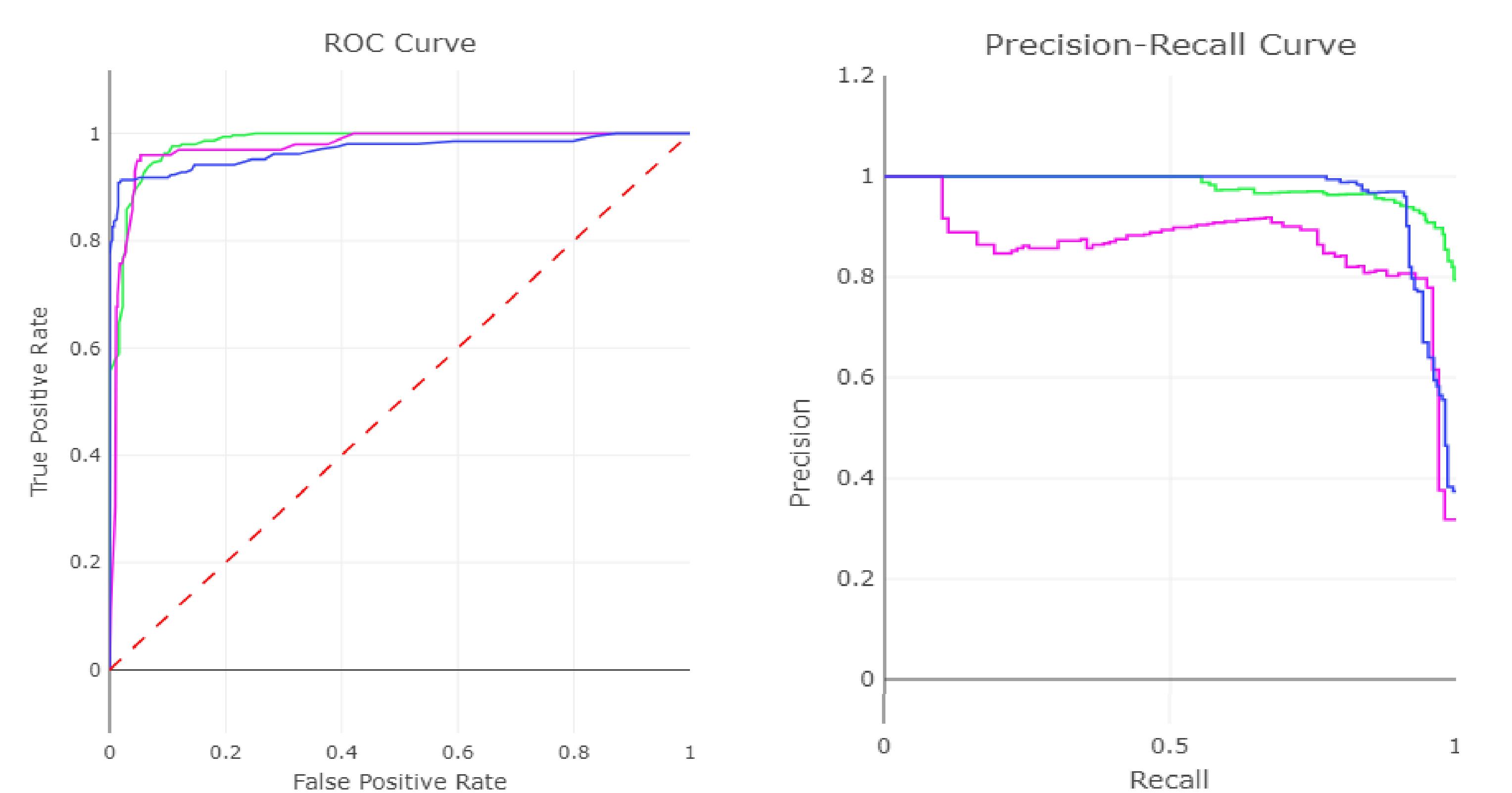
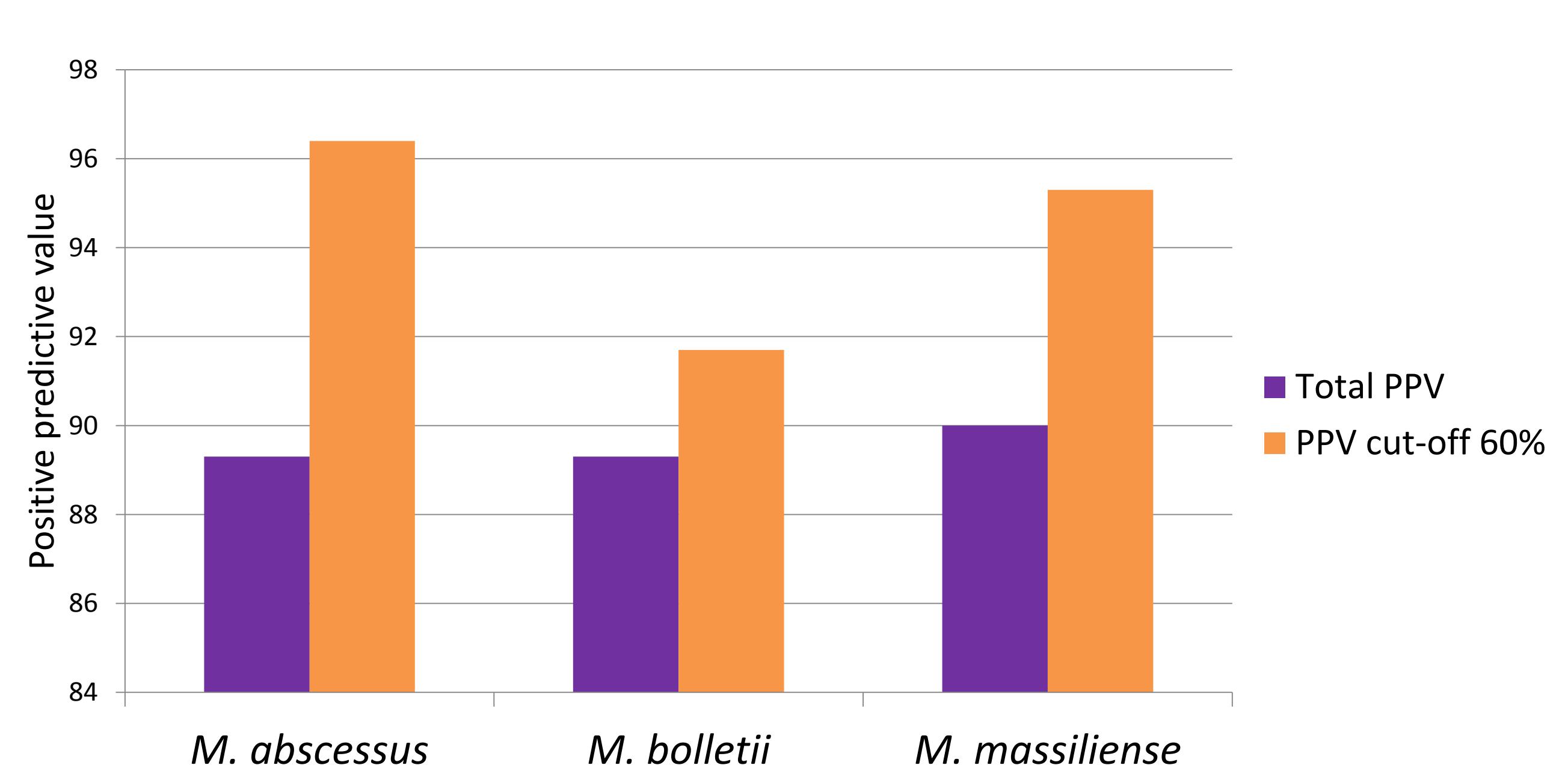
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447 **Table 3.** Presence of all protein peaks reported in previous studies and in the present
448 one among isolates of each subspecies. NA: not analyzed.

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A**B****C****D****D****E****F**