

1 Full length paper

2 **Electronic nose technology for detection of invasive pulmonary aspergillosis in prolonged**
3 **chemotherapy-induced neutropenia – a proof of principle study**

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25 **Abstract**

26

27 Although the high mortality of invasive pulmonary aspergillosis (IA) in patients with prolonged
28 chemotherapy-induced neutropenia (PCIN) can be reduced by timely diagnosis, a diagnostic test that
29 reliably detects IA at an early stage is lacking. We hypothesized that an electronic nose (eNose) can
30 fulfill this need. An eNose can discriminate various lung diseases through analysis of exhaled volatile
31 organic compounds (VOC's). An eNose is cheap, non-invasive and yields results within minutes.
32 **Methods:** in a single-center prospective cohort study we included patients that were treated with
33 chemotherapy expected to result in PCIN. Based on standardized indications a full diagnostic workup
34 was performed to confirm invasive aspergillosis or to rule it out. Patients with “*no aspergillosis*” were
35 considered controls and patients with “*probable/proven aspergillosis*” index cases. Exhaled breath was
36 examined with a Cyranose 320[®]. The resulting data were analyzed using principal component
37 reduction. Primary endpoint was the cross-validated diagnostic accuracy, defined as the percentage of
38 correctly classified patients using the leave-one out method. Accuracy was validated by 100,000
39 random classifications. Results: we included 46 subjects that underwent 16 diagnostic workups
40 resulting in 6 cases and 5 controls. The cross-validated accuracy of the eNose in diagnosing IA was
41 90.9% (p = 0.022; sensitivity 100%, specificity 83.3%). ROC analysis showed an AUC of 0.93.
42 **Conclusion:** these preliminary data indicate that in PCIN patients with IA have a distinct exhaled VOC
43 profile that can be detected with eNose technology. The diagnostic accuracy of eNose in invasive
44 aspergillosis warrants validation.

45 **Introduction**

46

47 The diagnosis of invasive pulmonary aspergillosis poses a significant challenge in clinical practice, due
48 to the fact that symptoms and signs of invasive pulmonary aspergillosis are neither sensitive nor
49 specific.(1-2) This also holds for conventional chest X-ray and cultures of sputum and/or broncho-
50 alveolar lavage. Furthermore, a CT scan of the lungs is a sensitive but non-specific test.(3) The
51 diagnosis is considered proven if a culture (from a normally sterile site that is clinically or
52 radiologically abnormal) yields *Aspergillus* spp.(3) Unfortunately, this requires invasive procedures,
53 such as percutaneous or transbronchial lung biopsy, which are rarely possible in the majority of patients
54 with IA, i.e. hematological patients experiencing prolonged chemotherapy-induced neutropenia. This is
55 due to concurrent thrombocytopenia and the risk of pneumothorax that is usually considered too large
56 in these patients.(4)

57

58 Over the last 10 years a number of new tests have been introduced, most notably the Platelia Assay: a
59 double-sandwich ELISA on galactomannan, a cell wall component of various molds including
60 *Aspergillus* spp. When performed on serum the assay has a sensitivity and specificity of about 80%
61 and, more importantly, a positive Platelia test can precede clinical manifestation by fever and other
62 symptoms.(5) Recently, it was shown that when performed on broncho-alveolar lavage, the sensitivity
63 and specificity of galactomannan is even higher.(6) However, broncho-alveolar lavage is not without
64 burden or even risks, and often not feasible. In addition, galactomannan is not detectable in serum until
65 accumulation of a considerable fungal burden.

66

67 As the mortality of IA is high (>50%) and can be reduced by a timely diagnosis, a diagnostic test that
68 can reliably detect IA at an early stage remains one of the major goals in mycology and supportive care

69 in hematology.(7-9)

70

71 Exhaled air is known to contain thousands of volatile organic compounds (VOC's), derived from
72 various metabolic pathways.(10) These VOC's can be used as biomarkers of lung disease, as has been
73 demonstrated for bronchial carcinoma, infectious diseases, COPD and asthma.(11-16) Recent evidence
74 indicates that a specific VOC, 2-pentyl-furan, could be a potential biomarker of IA.(17-18) However,
75 the need for gas chromatography and mass spectrometry (GC-MS) in the assessment of individual
76 volatile compounds precludes widespread on-site application in clinical practice.

77

78 An alternative way of assessing VOC-mixtures is using electronic noses (eNoses). An electronic nose is
79 an artificial olfactory system that discriminates complex odors using an array of sensors. When exposed
80 to exhaled breath the sensors react in a promiscuous way to the different fractions of VOC's.(19-21)
81 Each odor, which represents a unique mixture of VOC's, will result in a pattern of sensor signals unique
82 to that odor. This is called a "breathprint" when it concerns exhaled air. Using pattern recognition
83 algorithms, complex mixtures of VOC's can thus be discriminated at high-throughput without
84 identifying the individual molecular components as such. eNoses are relatively cheap, mostly hand-
85 held, easy to operate and yield a result within minutes. From a patient's perspective exhaled breath
86 analysis is appealing, because it is non-invasive, safe, rapid, simple to perform and effort independent.
87 Therefore, biomedical validation of eNoses is emerging.(19, 22)

88

89 We hypothesized that exhaled breath analysis using an electronic nose (eNose) can be used to diagnose
90 IA. To test that hypothesis, we performed a prospective proof of principle study.

91 **Methods**

92

93 **Subjects** – Patients were included if they: 1) were 18 years of age or older, 2) had given written
94 informed consent, and 3) were treated for a hematological malignancy with chemotherapy expected to
95 result in severe neutropenia ($<0.5 \times 10^9$ neutrophils/L) of more than 7 days, e.g. hematopoietic stem
96 cell transplantation or induction/consolidation treatment for acute myeloid leukemia. Patients were
97 excluded if they were previously diagnosed with an invasive mycosis, or if they were unable to perform
98 the breathing manoeuvre needed for eNose-analysis of exhaled air. The Medical Ethics Committee of
99 the Academic Medical Centre approved the protocol of the study. The study was registered at
100 ClinicalTrials.gov as study NCT01395446. All patients gave informed consent.

101

102 **Design** – The study was a single-center prospective cohort study. Based on standardized indications a
103 full diagnostic workup was performed to confirm invasive aspergillosis or to rule it out. The results
104 were classified according to the EORTC criteria, revised in 2008.(3) In the event of no possible,
105 probable or proven aspergillosis and no sero-positivity, the patient qualified as neutropenic control. In
106 the event of probable or proven aspergillosis a patient qualified as case. The breathprints of cases and
107 controls were compared. Exhaled breath analysis was performed only once in each patient.

108

109 According to this design, patients with possible aspergillosis were excluded. We chose for this design
110 because of the proof of principle nature of our study. Patients with possible aspergillosis can truly have
111 invasive aspergillosis, but more often do not have invasive aspergillosis. Including patients with
112 possible aspergillosis would make it harder to detect a breathprint associated with invasive
113 aspergillosis.

114

115 **Antifungal prophylaxis** - All patients were managed identically and according to a standardized
116 protocol based on recent guidelines with respect to the prevention, diagnosis and treatment of
117 mycoses.(23) Except for analysis of exhaled air using the eNose every prophylactic, diagnostic and
118 treatment-related procedure was according to standard care. Prophylactic antifungal treatment was
119 started the same day as the chemotherapy. Patients received 500 mg amphotericin B every 6 hours
120 orally until the peripheral neutrophil count exceeded $0.5 \times 10^9/L$. Patients undergoing myeloablative
121 allogeneic stem cell transplantation received 200 mg of fluconazole daily. If oral amphotericin B was
122 not tolerated, no substitute was started. In principle, no antimycotics with activity against *Aspergillus*
123 *spp.*, such as voriconazole or posaconazole, were administered prophylactically. If the treating
124 physician judged that the administration of prophylactic anti-mold treatment was necessary, the patient
125 was excluded from the study as eNose results could be influenced by anti-mold therapy.

126

127 **Diagnostic strategy** - From the start of chemotherapy onwards cultures of throat, nose, rectum and, if
128 possible, sputum were performed weekly. From the start of neutropenia ($<0.5 \times 10^9$ neutrophils/L) a
129 serum galactomannan assay was performed twice weekly. Both were continued until the peripheral
130 neutrophil count exceeded $0.5 \times 10^9/L$. A complete diagnostic workup was performed in case of a
131 number of standardized indications for a diagnostic workup, based on international guidelines: 1) a rise
132 of serum galactomannan above 0.5, or 2) five or more days of fever unresponsive to broad empiric
133 antibiotic treatment and without alternative explanation, or 3) a new infiltrate developing under broad
134 antibiotic coverage or highly dosed steroids, or 4) an abnormality on a chest X-ray consistent with
135 invasive pulmonary mycosis, or 5) hyphae or molds found in a respiratory specimen, or 6) symptoms
136 and/or signs considered by the treating physician to be possibly due to an invasive mycosis.(23)

137

138 **Diagnostic workup** - The workup consisted of 1) analysis of sputum (using direct microscopy and

139 culture), 2) a high resolution CT of the thorax, 3) in case of abnormalities on the HR-CT consistent
140 with invasive pulmonary mycosis bronchoscopy and broncho-alveolar lavage. The BAL was examined
141 using direct microscopy; a PCR on *M.tuberculosis* complex, mycobacterial culture, PCR's on
142 respiratory viruses (human boca virus, parainfluenza 1 to 4, parechovirus, coronavirus, rhinovirus,
143 RSV, human metapneumovirus, enterovirus, influenza A and B, adenovirus, HSV, EBV, CMV), routine
144 culture and measurement of galactomannan were performed as well. In case a broncho-alveolar lavage
145 was not performed a throat gargle specimen was examined using PCR's on the abovementioned
146 respiratory viruses. Sinonasal, ophthalmological and neurological symptoms and signs were actively
147 sought. On indication CT of the liver and spleen, CT or MRI of the brain and sinus, consultation of a
148 neurologist, otolaryngologist or ophthalmologist was performed.

149

150 **Exhaled breath analysis** – Every “diagnostic workup” was followed by exhaled breath analysis as
151 previously described.(11, 24) Patients were asked to breath through a mouthpiece for 5 minutes with
152 the nose clipped. Through a three-way non-rebreathing valve this mouthpiece was connected to an
153 inspiratory VOC filter (A2; North Safety, Middelburg, The Netherlands) as well as an expiratory silica
154 reservoir. Then, a deep inspiratory capacity maneuver was followed by the exhalation of a vital
155 capacity volume. The exhaled breath was collected in a 10 liter Tedlar bag which was connected to the
156 silica reservoir. Within 30 minutes the Tedlar bag was sampled using the electronic nose, a Cyranose
157 320[®] (Smith Detections, Pasadena, CA). This is a hand-held chemical vapor analyzer based on a nano-
158 composite sensor array with 32 polymer sensors.(19) The change in electrical resistance in each of the
159 32 sensors was stored as raw data for further analysis. Every sampling procedure was repeated after
160 which the first measurement was disregarded, as previously described because of deviant raw data at
161 first run.(11)

162

163 **Analysis** – As our primary analysis we compared the breathprints of cases and controls. We performed
164 offline analysis of the raw data using R (version 2.11.1). First, data reduction by principal component
165 analysis (PCA) was performed to reduce the original data of the 32 sensors to a non-predefined number
166 of principal components, capturing at least 99.9% of the variance within the dataset. Secondly, t-tests
167 (equal variance assumed depending on the outcome of an F-test) were used to assess which PCA
168 factors discriminated between the two groups; a two-sided p-value of 0.10 was considered significant.
169 Then, based on the differentiating PCA factors a categorical division was made using linear canonical
170 discriminant analysis, assuming an equal chance of being a member of one of the two groups. The
171 discriminant function was chosen to best distinguish between categories. Finally, the accuracy of this
172 model was established. This was defined as the percentage of correctly classified patients, cases and
173 control subjects combined. Cross-validation using the leave-one out method was used to calculate the
174 cross-validated accuracy. The 95% confidence intervals (CI) were calculated using the exact binomial
175 test. To calculate our p-value, we generated 100,000 random classifications of our subjects (“whether or
176 not case or control”) and determined the chance that a random classification would have led to a cross-
177 validated accuracy identical to our primary outcome or better, constructing a new pattern recognition
178 algorithm for each of the random classifications using the statistical method of the primary
179 analysis.(25) Finally, ROC analysis was performed.

180 **Results**

181

182 During the study period there were 53 eligible patients. As 5 refused informed consent and 2 had
183 previously been diagnosed with an invasive mycosis, 46 patients were included. In 16 of these subjects
184 one or more triggers for a diagnostic workup occurred. This resulted in 6 controls and 5 cases, see
185 Table 1. Principal component analysis of the raw data resulted in 8 principal components (PC's) that
186 described 99.9% of the variance. Of these 1 discriminated between cases and controls. Subsequent
187 canonical discriminant analysis showed a cross validated accuracy value of 90.9% (95% CI 59% to
188 100%). The sensitivity and specificity were 100 (95% CI 48 to 100%) and 83.3% (95% CI 36% to
189 100%). Figure 1 shows the individual discriminant scores. ROC analysis of the discriminant scores
190 revealed an area-under-the-curve of 0.933. In our simulation 2.2% of the 100,000 random
191 classifications resulted in a cross-validated accuracy identical to or better than 90.9%. In all patients in
192 whom both values were determined we calculated the correlation between the discriminant scores and
193 BAL galactomannan values, currently the most accurate single test to diagnose invasive pulmonary
194 aspergillosis. Although these were correlated, this was not statistically significant (unstandardized
195 regression coefficient -0.23, 95% CI -0.54 to 0.08, R2 0.36), see Figure 2.

196 **Discussion**

197 Our study shows that patients with invasive aspergillosis have an exhaled VOC profile, distinctive from
198 controls, which can be established by eNose technology. The accuracy is high and, as shown by the
199 random classifications, is not a coincident finding. This implies that in the future exhaled breath
200 analysis could become a non-invasive addition to the diagnostic arsenal in invasive aspergillosis that is
201 cheap, fast and simple to perform.

202

203 eNose technology will hopefully enable us to detect invasive aspergillosis at an earlier point in time
204 than currently available diagnostic tools. At the time of the exhaled breath analysis in subject 4, he was
205 thought to have *no possible, proven or proven aspergillosis* based on the diagnostic workup according
206 to protocol. Two weeks later however, probable aspergillosis was diagnosed. In retrospect, very small
207 pulmonary lesions were already seen two weeks before at the locations where later aspergilloma
208 developed. Therefore, he was classified as being a case in our study. Out of interest, we also performed
209 a second exhaled breath analysis two weeks later, when we diagnosed probable aspergillosis. This
210 measurement was not used to derive the pattern recognition algorithm for our primary analysis, off
211 course. We compared the two exhaled breath analyses. Although the first signal (discriminant score of -
212 0.49) was less pronounced than the second (-1.37), it did already indicate IA.

213

214 To our knowledge, this is the first study examining the accuracy of exhaled air in the early diagnosis of
215 invasive aspergillosis. It is however in line with previous *in vitro* research, which already showed that
216 an eNose can reliably differentiate *in vitro* the most frequently encountered pathogens in pneumonia.
217 Moens et al. demonstrated that an eNose could differentiate the headspaces of various micro-organisms
218 after 17 hours of culturing with a diagnostic accuracy of 100%. They examined Gram negative bacteria
219 (*P.aeruginosa*, *E.coli*, *K.pneumoniae*, *E.aerogenes*, *P.vulgaris*), Gram positive bacteria (*S.aureus*,

220 *S.pneumoniae*, *E.faecalis*), a yeast species (*Candida albicans*) and a mold species, *Aspergillus*
221 *fumigatus*.(26) Other groups confirmed that an eNose is able to differentiate the headspaces of various
222 micro-organisms.(27-28) These results were already extended to an in vivo situation, i.e. ventilator-
223 associated pneumonia (VAP). Hockstein et al. calculated pneumonia scores in 44 ventilated patients
224 based on a number of clinical criteria.(29-30) An eNose could reliably differentiate between the 7
225 patients with a high pneumonia score and the 29 patients with a low pneumonia score. Our data thus
226 support and extend the accumulating evidence that eNose technology can be used to diagnose
227 pulmonary infections.

228

229 Our study has a number of strong points. It studied a prospective cohort in which the patients were
230 followed according to a state-of-the-art diagnostic protocol, defining the timing of our exhaled breath
231 analyses and characterizing our population well with respect to whether aspergillosis occurred. This
232 also yielded a well-characterized control group.

233

234 On the other hand, our study is subject to two major limitations. First of all, the sample size was small
235 due to the low incidence of IA. This precluded external validation of our results. As the aim of the
236 study was to detect IA. However, our 100,000 random classifications indicated that the chance of false-
237 positive discovery was only 2.2%. It also Eventually, according to guidelines on stepwise assessment of
238 diagnostic accuracy of novel tests, the confirmation of our results in a separate group of subjects that
239 was not involved in generating the pattern-recognition algorithm will be required to definitively
240 establish the ability of an eNose to detect IA.(31) Such external validation has already been provided
241 for the differential diagnosis by eNose between COPD and asthma.(32)

242

243 Secondly, eNose technology, albeit applicable for medical applications, does not allow identification of

244 the individual VOC's that drive the signal. It is unknown which VOC's enable the detection of IA by
245 eNose technology. First, these could be VOC's produced by *A.fumigatus* itself. In the literature a
246 number of potential candidates have been suggested. One such compound is 2-pentylfuran, which was
247 reported by a research group from New Zealand to be *A.fumigatus*-specific, being exhaled by subjects
248 with colonization as well as invasive disease caused by *A.fumigatus*.(17-18) However, differences in
249 the composite molecular signatures as captured by breathprints may arise from other sources rather
250 than *A.fumigatus*, such as the host response. The presence of *Aspergillus spp.* in the airways triggers an
251 immune response. In a number of patients this even leads to the clinical entity called allergic broncho-
252 pulmonary aspergillosis (ABPA).(33) Notably, inflammatory airway diseases, such as asthma and
253 COPD, can be discriminated at a high level of accuracy through eNose technology, in which the signals
254 by eNose as well as GC-MS are significantly associated with cellular and molecular markers of airways
255 inflammation.(34-35) Such inflammatory host responses could have played a major role in our study,
256 augmenting the difference in exhaled VOC profiles and aiding in early detection. Invasive aspergillosis
257 induces a major immune response, despite neutropenia.(36)

258

259 The implications of our results are potentially wide. Exhaled breath analysis could increase the
260 accuracy of the diagnostic workup of a patient suspected of having invasive aspergillosis. It could also
261 decrease the mortality of invasive aspergillosis, for example through a reduction of the diagnostic delay
262 by monitoring patients with prolonged chemotherapy-induced neutropenia twice per week. And lastly,
263 if it were to improve the diagnostic accuracy enough, it could obviate the need for bronchoscopy,
264 thereby making the workup less invasive. Furthermore, if further translational research would unravel
265 the molecules involved in the generation of the specific breathprint, eNoses could be "tailor-made" to
266 detect these VOC's to improve the diagnostic accuracy even further.(16)

267

268 In conclusion, this study shows the potential of eNose technology in the detection of IA in patients
269 experiencing prolonged chemotherapy-induced neutropenia through analysis of exhaled breath. This
270 warrants the next step in testing diagnostic accuracy by performing a large-scale validation study in
271 order to determine how much diagnostic delay can be prevented by adding twice weekly exhaled breath
272 analysis using eNose to a state-of-the-art diagnostic strategy in invasive aspergillosis.(31)

273

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275

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383 **Legenda of figures and table**

384

385 Table 1. Subject characteristics.

386

387 AML: acute myeloid leukemia; ALL: acute lymphoblastic leukemia; R-VIM: ritxumab/
388 etoposide/iphosphamide/methotrexate; n.p. not performed; P: positive as a clinical EORTC
389 criterium; N: negative as a clinical EORTC criterium; induction: induction chemotherapy

390

391 **Figure 1.** Individual discriminant scores derived from exhaled breath profiles of patients with and
392 without invasive pulmonary aspergillosis.

393

394 **Figure 2.** Correlation of galactomannan on BAL and discriminant scores based on exhaled breath
395 profiles in subjects in whom bronchoscopy was performed. The fit line based on linear regression and
396 the 95% mean prediction intervals are shown.

Table 1. Subject characteristics

	subject	age	sex	diagnosis	therapy	HR-CT of the lungs	serum galactomannan	BAL galactomannan	cultures positive for <i>Aspergillus spp.</i>	EORTC classification	remarks
probable/proven aspergillosis											
	1	62	F	AML	induction	P	0.1	2.3	no	probable	concurrent influenza (H1N1)
	3	35	F	AML	induction	P	0.2	7.4	no	probable	
	4	47	M	AML	induction	P	1.9	7.6	no	probable	
	8	70	M	ALL	induction	P	0.3	8.1	no	probable	
	9	56	M	Waldenström	R-VIM	P	1.2	n.p.	no	probable	
	means	54.0	40%	female			0.7	6.4			
no aspergillosis											
	2	53	F	AML	induction	N	0.1	n.p.	no	no	
	5	51	F	AML	induction	N	0.2	n.p.	no	no	
	6	65	F	AML	induction	P	0.1	0.3	no	no	bacterial pneumonia
	7	46	F	AML	induction	N	0.1	n.p.	no	no	left-sided pleural fluid
	10	63	M	AML	induction	N	0.1	n.p.	no	no	bacterial pneumonia
	11	74	M	AML	induction	N	0.1	n.p.	no	no	bilateral pleural fluid
	means	58.7	67%	female			0.1	0.3			

Figure 1. Individual discriminant scores derived from exhaled breath profiles of patients with and without invasive pulmonary aspergillosis

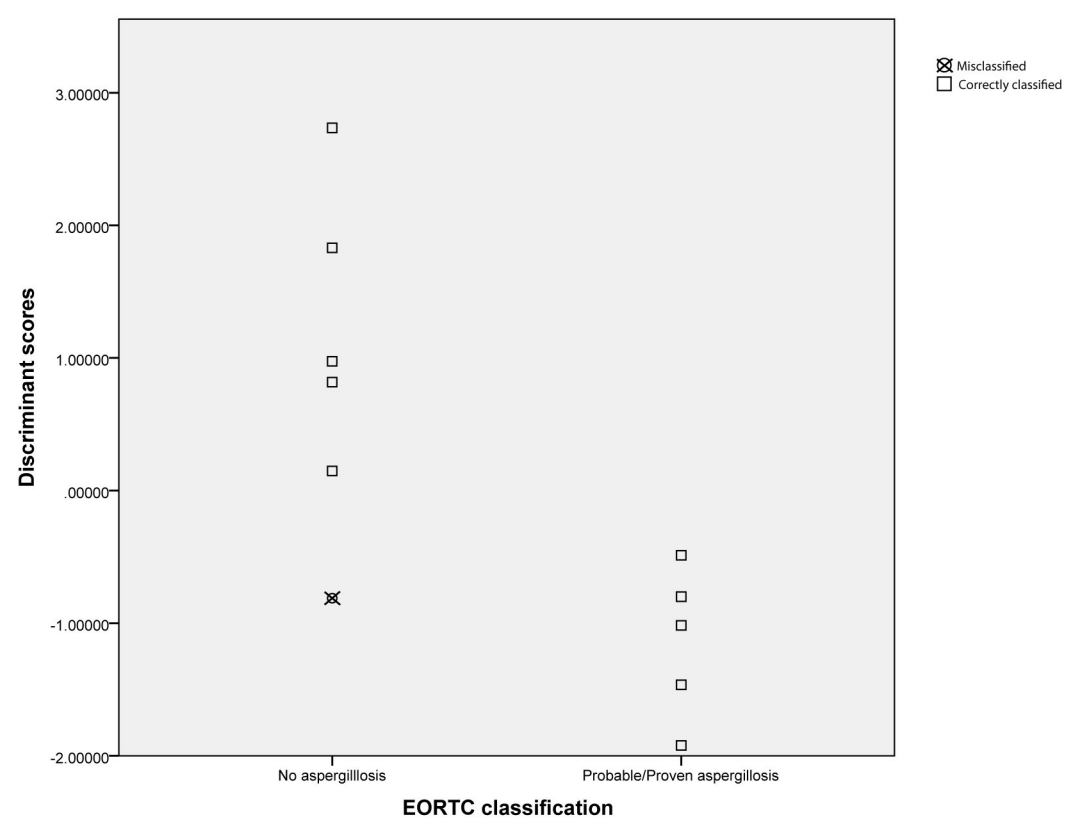


Figure 2. Correlation of galactomannan on BAL and discriminant score based on exhaled breath profiles in subject in whom bronchoscopy was performed

