An Enantioselective e-Nose: An Array of Nanoporous Homochiral MOF Films for Stereosepecific Sensing of Chiral Odors

Salih Okur, Peng Qin, Abhinav Chandresh, Chun Li, Zejun Zhang, Ulrich Lemmer, and Lars Heinke*

Abstract: Chirality is essential in nature and often pivotal for biological information transfer, for example, via odor messenger molecules. While the human nose can distinguish the enantiomers of many chiral odors, the technical realization by an artificial sensor or an electronic nose, e-nose, remains a challenge. Herein, we present an array of six sensors coated with nanoporous metal–organic framework (MOF) films of different homochiral and achiral structures, working as an enantioselective e-nose. While the achiral-MOF-film sensors show identical responses for both isomers of one chiral odor molecule, the responses of the homochiral MOF films differ for different enantiomers. By machine learning algorithms, the combined array data allow the stereoselective identification of all compounds, here tested for five pairs of chiral odor molecules. We foresee the chiral-MOF-e-nose, able to enantioselectively detect and discriminate chiral odors, to be a powerful approach towards advanced odor sensing.

Chirality and enantioselectivity are fundamental in nature and crucial in various fields, ranging from DNA molecules to pharmaceuticals.[1] In addition, biological messenger molecules like pheromones are often chiral.[2] Typically the bioactivity is stereoselective and the different pheromone enantiomers have different, sometimes opposing, biological functions.[2,3] Equally, most chiral odor molecules have an enantioselective smell, meaning that the perception of the individual enantiomers by the human olfactory system differs significantly. Among the most popular examples is limonene: While the R-isomer has an odor of an orange, the S-isomer smells like lemons.[4] Another example is 1-phenylethanol, the R-isomer has a floral, earthy-green odor, while the S-isomer smells like mild hyacinth with strawberry nuances.[5] The odor of the R-enantiomer of 2-octanol is described as creamy, cucumber, fatty and sour, while the S-enantiomer has more of a mushroom odor.[6] These odors and fragrances are characterized by a panel of judges with “trained noses”.[4] For quantification, the compositions of chiral odor samples are typically analyzed by enantioselective chromatography with homochiral columns.[6,7] When optimized, this method is very precise and sensitive; however, due to its complexity and setup size it is not suited for practical sensor applications.

Several efforts have been made to fabricate enantioselective sensors. Different chiral materials based on chiral polymers,[8] supramolecular chiral systems,[9] cyclodextrins,[10] carbon-nanotubes,[11] or graphene[12] functionalized with chiral molecules were used to discriminate both enantiomers of one chiral molecule in a sensor setup. For example, in individual experiments, meaning one enantiomer pair at a time, the S- and R-enantiomers of α-pinene, β-pinene, and limonene were distinguished by DNA-functionalized carbon-nanotubes.[11a] A powerful technique for determining the molecular uptake is a quartz-crystal microbalance (QCM), where the recorded frequency shift, corresponding to the mass change, is used as sensor signal. Although various chiral materials, like cyclodextrins,[10a,13] chiral polymers,[14] and chiral molecular monolayers,[15] have demonstrated an enantioselective response in combination with a QCM, the enantioselective discrimination of many chiral molecules, for example, by a QCM-based e-nose, has not yet been demonstrated. For the simultaneous discrimination of many molecules, an array of sensors with different, ideally orthogonal, selectivities resulting in a multi-dimensional response is required.

Films of nanoporous homochiral materials, such as homochiral metal–organic frameworks (MOFs), seem perfectly suited for sensing applications. In addition to their record specific surface area resulting in high sensitivities, a striking advantage of MOFs is their huge variety with roughly 100000 published structures[16] and at least hundreds, presumably many thousands of homochiral MOF structures.[17] So far, numerous homochiral MOFs were used to discriminate the enantiomers of one chiral molecule.[18] There, R- and S-enantiomers are typically distinguished by using chromatography or gravimetric techniques taking advantage of their enantioselective adsorption strength in the chiral MOF nanopores.[19] For example, using a lactate-based MOF...
A sensor array which detects and discriminates the enantiomers of various molecules (even in their pure form) simultaneously has not yet been presented. Such a sensor array should work like an electronic nose, e-nose,

composing of different enantioselective sensors, allowing to distinguish various chiral molecules and their enantiomers. The application range of an enantioselective e-nose is very wide, ranging from the sensing of bioactive chiral pheromones, e.g. to measure and inhibit the communication between infesting insects,

and deteriorating of food.

E-nose systems without enantioselectivity are already widely applied in many fields to detect odors, for instance to assess food authenticity and adulteration.

The active sensing materials in such e-noses are typically based on polymers or inorganic materials with large surface areas.

Recently, the potential of MOFs for e-nose applications was demonstrated. So far, enantioselective discrimination of many molecules is realized by the combination of the e-nose with an enantioselective technique such as chiral chromatography or electrophoresis.

Herein, we present an enantioselective e-nose, which is based on QCM sensors coated with six different nanoporous homochiral and achiral MOF thin films. The MOF thin films were directly prepared on the QCM sensors in a layer-by-layer fashion, resulting in surface-mounted MOFs, SURMOFs. The e-nose was tested for five pairs of chiral odor molecules: (R)- and (S)-limonene, (R)- and (S)-2-octanol, (R)- and (S)-1-phenylethanol (R)- and (S)-1-phenylethylamine and methyl (R)/S lactate, 10 volatile organic compounds (VOCs) in total. While the sensors with the achiral MOF structures show very similar responses for both isomers and can only distinguish the different molecules, the sensors coated with the homochiral MOF structures can enantioselectively distinguish the chiral molecules. By machine learning algorithms based on k-nearest-neighbor (kNN) analysis, the sensor array can discriminate all molecules and their isomers with 96% confidence.

The e-nose is composed of six quartz-crystal microbalance (QCM) sensors coated with SURMOFs of six different structures, comprising three homochiral and three achiral architectures. The homochiral SURMOFs are Cu$_2$(DCam)$_2$-(dabco), Cu$_2$(DCam)$_2$(BiPy), and Cu$_2$(DCam)$_2$(BiPyB), denoted as chirMOF1, chirMOF2, and chirMOF3, grown in (100) orientation perpendicular to the substrate. The structures of the SURMOFs are shown in Figure 1 and S3. The X-ray diffractograms (Figure S4) show that all samples are crystalline and have the targeted MOF structures. In addition, the diffraction patterns show that the samples are grown in an oriented way on the substrate. In detail, the chiral pillared-layer SURMOFs are grown in (001) orientation and the achiral SURMOFs are grown in (100) orientation perpendicular to the substrate.

The response of each QCM sensor, which is the frequency shift caused by the molecular uptake by the SURMOF films, is proportional to the mass change caused by the uptake of the guest molecule in the MOF film. When the QCM-SURMOF-array is exposed to the different odor molecules, the frequency shifts of all sensors are recorded simultaneously. The sensor array response to limonene and 1-phenylethanol, with various concentrations in the ppm range, is shown in Figure 2. The data of the response to the other molecules, which are 1-phenylethylamine, 2-octanol, and methyl lactate, are shown in Figure S6. Although the data for the different molecules appear qualitatively similar, some quantitative differences in the response can be found. For example, a detailed inspection shows that while the uptake of (S)-limonene by chirMOF1 is more than three times as much as by chirMOF2, this ratio is only approximately 10% for (R)-limonene. Similarly, the uptake of (S)-1-phenylethanol by chirMOF3 is less than by achirMOF1, for (R)-1-phenylethanol it is opposite.

Radar plots of the response to the different pure enantiomers at a concentration of 50 ppm are shown in Figure 3. The radar plots at 10 and 100 ppm, Figures S7, are qualitatively similar to the plot at 50 ppm, however with different scaling. The response of the achiral SURMOFs differs for different molecules but is essentially identical for both enantiomers, as found for all investigated molecules.
This means that the molecules can be distinguished but, as expected, these sensors cannot distinguish the enantiomers. In detail, the achiral sensor responses for the \(R\)- and \(S\)-enantiomers are within a range of a few percent, see Table S1, and the average difference is 3.0%. On the other hand, the sensors coated with the homochiral MOF films show clear differences between the enantiomers of the same molecule. For instance, in addition to the limonene example discussed above, the response of 1-phenylethylamine at the sensor chirMOF1 is 60% larger for the \(R\)-enantiomer in comparison to \(S\).

The chiral sensors show an average relative signal difference between both enantiomers of 40.4%. In addition, while some molecules often show a stronger signal for the \(R\)-isomers, like for methyl lactate and 1-phenylethylamine, the signals of limonene in all three chiral sensors are stronger for the \(S\)-isomer. For 2-octanol, the signal of the \(R\)-isomer is stronger than for \(S\) in sensors chirMOF1 and chirMOF3 but opposite in chirMOF2.

Figure S8 shows the sensor response versus the odor concentration. The plots show essentially linear responses in the investigated range. Noteworthy, each sensor shows a different slope for the same molecule and, more importantly, different molecules show different slopes in the same sensor.

Since all isomers of the odor molecules have characteristic patterns on the radar plots, their identification is possible in principle. For a more detailed understanding and quantification of the sensor data and for a performance investigation of the enantioselective e-nose, the data from the sensor array was analyzed by a machine learning algorithm. Here a kNN algorithm was used, see SI.

The confusion matrices, summarizing the performance of the classification algorithm, are shown in Figure 4 and S11. The true classes are the rows and the predicted classes are the columns. The accuracy of classification of the data points by kNN is shown as percentage in the matrix. Correct classifications are on the main diagonal of the matrix; misclassifications are the other values. The confusion matrix (Figure 4a) shows that all odors can be distinguished by their QCM-array responses. Most enantiomers can be discriminated with perfect accuracy; only the discrimination of the 2-octanol enantiomers shows an overlap of the \(R\)- and \(S\)-isomers with about 80% correct and 20% false classifications. On average, the classification accuracy of distinguishing all 10 isomers is 96.1%. This means a very high precision for detection and discrimination was realized.

For comparison, if the data from only the chiral sensors are used (Figure 4b, left) the accuracy is slightly decreased to
In addition to the small misclassification of (R/S)-octanol, (S)-octanol is also misclassified as (S)-limonene with 22%. Using only the data from the three achiral sensors, Figure 4b, right, shows that the molecules can be classified; however, the discrimination of the enantiomers is not possible. The chance of classifying the enantiomer correctly is about 1:1, i.e. 50%. This is also expected, since the achiral MOFs possess no chiral moieties allowing an enantioselective response. Noteworthy, the molecules (not the enantiomers) are classified with 100% accuracy.

The sensor responses during the exposure to the odor molecules can be described with mono-exponential decay functions with time constants in the range of a few minutes, typically 3 to 10 min, see Table S2. In Figures 4c and S11c, the transient behavior of the accuracies during the odor exposures is shown. The data show that the classification accuracy of the entire sensor array varies slightly in between about 90 and almost 100%. Only at the beginning of the odor exposure (within the first 3 min) the accuracy is significantly smaller than the final value, which is caused by the initial large signal change per time and the accompanying large scattering of the data as well as the fact that all data curves, that is, all frequency shifts, start at 0 Hz. Nevertheless, the data of the sensor array allow already a decent classification of the chiral molecules right at the beginning, although the uptake curves are far from equilibrium. The classification accuracies based on only half the sensor array, either the three chiral or the three achiral sensors, also show slight changes with time and are always smaller than the accuracy from the entire array. While the chiral sensors allow a precise classification with only a few percent less than the entire sensor array, the accuracy based on only the achiral sensors is in the region of 50%, in detail, in the range between 40 and 74% for all concentrations and times. Please note that the kNN-results of the data from 10, 50, and 100 ppm are very similar, with only minor changes of the classification accuracies, see Figure S11.

The sensor array also allows the analysis of the odor signals from different (pure) vapor concentrations simultaneously, Figure S12. Most signals can be unequivocally described to the isomer and the concentration, however, a few signals show cross-sensitivities with different concentrations.

In addition to the kNN analysis, the results of linear discriminant analysis (LDA) and principal component analysis (PCA) are shown in Figures S13 and S14. The LDA scatterplots as well as the PCA score and loading plots also allow a straightforward visual enantioselective grouping and classification of the sensor responses.

In comparison to QCM-based chiral recognition using cyclodextrins[10a,13] or chiral polymers[14] as enantioselective sensor material, using chiral MOF films presents the advantage of sufficiently different selectivities enabled by the different MOF structures. Since the chemical space of MOFs is very large, the adsorption properties, and thus the sensor response, can be tuned. Ideally, a sensor array with very different or even orthogonal responses for all enantiomers should be realized. We like to stress that although the used chiral MOFs possess many structural similarities, the enantioselective response of these sensors are very different. Taking into account that roughly 100000 MOF structures are currently known, including many chiral structures, the possibilities of using MOFs as sensor material seem unlimited.

91.2%. In addition to the small misclassification of (R/S)-octanol, (S)-octanol is also misclassified as (S)-limonene with 22%. Using only the data from the three achiral sensors, Figure 4b, right, shows that the molecules can be classified;
In conclusion, a QCM sensor array coated with different homochiral and achiral MOF thin films is presented which works as an electronic nose for chiral odor molecules. While the achiral MOF sensors cannot distinguish the different isomers of the chiral odors, each chiral MOF sensor shows a different response to the different enantiomers. As a result, the combined data of the sensor array allow the enantioselective detection and discrimination of chiral odor molecules. The e-nose was tested for five pairs of chiral odor molecules, where all ten enantiopure compounds were detected and distinguished with an average accuracy of 96%.

In the future, the number of sensors in the array should be enlarged with MOF films with well-chosen selectivities to increase the confidence of the sensing[33] and also to increase the range of detectable odors. In addition, the detection of applications is possible.

A handy sensor size suitable for practical and economic low sensor costs, we believe significant miniaturization to interaction. Due to the small size of MOF-QCM sensors and is required for a detailed understanding of the complex interaction. For the molecular enantioselective interaction in MOFs is only thoroughly explored for a few examples[37] and more research is required for a detailed understanding of the complex interaction. Due to the small size of MOF-QCM sensors and the low sensor costs, we believe significant miniaturization to a handy sensor size suitable for practical and economic applications is possible.

Acknowledgements

We acknowledge funding through the Alexander von Humboldt Foundation (Philipp Schwartz Fellowship), the China Scholarship Council (CSC), and the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) under Germany’s Excellence Strategy-2082/1-390761711 and DFG HE 7036/5. We also thank Stefan Bräse (KIT) for stimulating discussions. Open access funding enabled and organized by Projekt DEAL.

Conflict of interest

The authors declare no conflict of interest.

Keywords: chirality · electronic nose · enantioselectivity · metal–organic frameworks · sensors
