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Pattern-Based Recognition for the Rapid Determination of Identity, Concentration, and Enantiomeric Excess of Subtly Different Threo Diols

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Abstract

A pattern based recognition approach was developed for the rapid determination of identity, concentration, and enantiomeric excess of chiral vicinal diols, specifically threo diols. A diverse enantioselective sensor array was generated using three chiral boronic acid receptors and three pH indicators. The optical response produced by the sensor array was analyzed by two pattern recognition algorithms: principal component analysis and artificial neural networks. Principle component analysis demonstrated good chemoselective and enantioselective separation of the analytes, and an artificial neural network was used to accurately determine the concentration and enantiomeric excess of five unknown samples with an average absolute error of ± 0.08 mM in concentration and 3.6% in enantiomeric excess. The speed of the analysis was enhanced by using a 96-well plate format, which portents to applications in high-throughput screening for asymmetric catalyst discovery. X-ray crystallography and ^{11}B NMR was utilized to study the enantioselective nature of boronic acid host-2.

Keywords

Enantioselectivity; High-throughput screening; Host-guest system; Indicator displacement assay; Principal component analysis; Artificial neural networks

INTRODUCTION

Animals are constantly surveying their external environment via their olfactory system for chemicals that indicate food sources and habitats, as well as chemical signals controlling social interactions and reproductive behavior.¹ The olfactory system involves a recognition protocol where the detectors/sensors generate a pattern that is recognized and classified by the brain.² Inspired by the olfactory system, the use of multivariate data analysis combined with sensors of partially overlapping selectivity has become a powerful tool in the field of molecular recognition.³⁻²¹

In an analogous fashion, a novel optical recognition scheme is described here for the rapid determination of the identity, enantiomeric excess (*ee*), and concentration ($[\text{G}]_t$) of chiral

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SUPPORTING INFORMATION. Spectroscopic data of the host (*S,S*)-2 with the indicators, UV/vis data for PCA and ANN; Crystallographic information file (CIF) of complex 7; ^{11}B NMR of host (*R,R*)-2 and complex 7 is available free of charge via the Internet at <http://pubs.acs.org>.

vicinal diols particularly three diols of close structural similarity. The differential receptors are chiral boronic acids, which are responsible for the sensors' chemoselective and enantioselective discriminatory properties. The signaling relies upon indicator displacement assays (IDAs) to provide a colorimetric response. The assay was developed by using combinations of three differential receptors and three different indicators within a single array, which resulted in fingerprinting of each diol analyzed.

Indicator displacement assays rely on colorimetric or fluorescence indicators that reversibly interact with a receptor and, upon binding to the receptor, undergo change in their optical properties. A competition between the analyte and the indicator for the binding site of the receptor generates an equilibrium in solution that can be monitored optically and related to $[G]_t$.²² Our group has pioneered the development of enantioselective indicator displacement assays that utilize a chiral receptor and where the color change is related to *ee* and $[G]_t$.^{23,24}

The data generated by the sensor arrays is commonly processed by unsupervised and supervised pattern recognition algorithms. A major unsupervised technique is principle component analysis (PCA), while an artificial neural network (ANN) is a supervised technique. PCA reduces multidimensional and partly correlated data to two, three or more dimensions. This is achieved by projecting the data onto fewer dimensions that represent maximum variance relationships between variables.²⁵ ANN programs are based on a simplified model of the brain. ANN requires a training set that consists of a collection of known parameters. The training set then creates a neural network suitable for the analysis in hand. This network is then used to analyze unknown samples.^{26,27}

The goal of the present study was to demonstrate the ability to distinguish chirality as well chemical identity of subtly different analytes; therefore, diols **3-6** were selected as targets (Scheme 1). Analyte **3** has two phenyl groups, in contrast to **4** and **5**, which contains only one. Analytes **4** and **5** differ in a ketone relative to an ester, respectively. Lastly, analyte **6** has two esters. Hence, we are challenging our array approach by using four analytes and their enantiomers, all possessing electron withdrawing groups of only subtle difference.

Our previous studies have shown that the use of cross-reactive sensors in a single array enhances the fingerprinting of the analytes, allowing one to determine structural similarities and chirality simultaneously.²⁸ Furthermore, increasing the enantioselectivity of the receptors improves chiral discrimination. Due to previous results that showed that C_2 -symmetric chiral secondary amines on the boronic acid give excellent enantioselectivity, host-**1** was chosen.²⁹ With this guiding principle we now introduce host-**2**, which is also C_2 -symmetric about the pyrrolidine ring (Scheme 1). The aim was to develop a second highly enantioselective host that can be used in conjunction with host-**1** in our pattern recognition protocol.

The use of an additional chiral receptor could allow for concentration as well as *ee* determination. In our previous studies, an achiral receptor has been consistently used to determine concentration.³⁰ In the senses of taste and smell, however, the suites of receptors used are all chiral by virtue of their proteins structure. Yet, both handedness and relative concentrations of tastants and odorants can be determined. Hence, another goal of the work described herein was to show that a suite of chiral synthetic receptors could do the same. Further, in this study we explore the reason behind the enantioselectivities of the host-**2** via x-ray crystallography and ^{11}B NMR.

RESULTS AND DISCUSSION

Synthesis

Host-**1** and host-**2** were synthesized by reductive amination according to a published literature procedure.³¹ The diols used in this analysis are commercially available except for **4**, which was synthesized by a previously established Sharpless asymmetric dihydroxylation reaction.³² Diol (3*S*,4*R*)-3,4 dihydroxy-4-phenyl-butane-2-one (**4a**) was isolated in 70% yield and 93% *ee* while (3*R*,4*S*)-3,4 dihydroxy-4-phenyl-butane-2-one (**4b**) was obtained in a 76% yield and 86% *ee*. The *ee* of **4a** and **4b** were determined by using the James, *et al.* NMR analysis (Figure S7).³³

X-ray Crystallographic Analysis

To study the enantioselective behavior of our boronic acid hosts, an x-ray quality crystal of host (*R,R*)-**2** with (*S,S*)-hydrobenzoin (**3b**) was isolated by diffusing methanol into a concentrated solution of the complex in dichloromethane (Figure 1). Previously, our group has analyzed the free boronic acid host *o*-(pyrrolidin-1-ylmethyl)phenylboronic acid and its complexes with catechol by x-ray crystallography.³⁴ It was shown that in these complexes the presence of the N-B bond is solvent dependent. An aprotic solvent (CHCl₃) facilitates the formation of the N-B bond, while in protic solvent (CH₃OH) a solvolysis pathway is operative. When complex **7** was crystallized from methanol, instead of the expected solvent inserted structure, the boron atom was found to be *sp*² hybridized. The average B-O bond length is 1.36 Å and the O-B-O angle of the cyclic (*S,S*)-hydrobenzoin boronate is 113.23°. As expected in such a case, there is no intramolecular B-N bond; instead an N-B distance of 4.43 Å is observed. The optimal distance for an N-B bond is 1.5-1.8 Å as reported for other boronic acid molecules with similar configurations.^{35,36} The absence of the solvent inserted structure could be due to the more sterically inaccessible nitrogen lone pair in the tertiary amine of host (*R,R*)-**2**; as we have recently found in another sterically congested system.³⁷ Interestingly, the ¹¹B NMR profile of the complex in CD₃OD indicates that the boron atom is solvated in solution (see next section). Therefore, there is a discrepancy between the structure of the complex in solid state and in solution.

Nevertheless, the crystal structure does lend insight into the enantioselectivity of host-**2**. The indicator displacement isotherm for host (*R,R*)-**2** and host (*S,S*)-**2** with **3a** and **3b** using ML (Scheme 1) as an indicator is shown in Figure 2. As required by the principal of chirality, the enantiomeric hosts show opposite enantioselectivity for enantiomeric guests. As shown, host (*R,R*)-**2** has a lower affinity for **3b** than **3a**. This can be explained through the crystal structure in which the phenyl group on the pyrrolidine ring of the host (*R,R*)-**2** has an unfavorably steric interaction with the phenyl group of **3b**. This unfavorable interaction would not be present in the complex of host (*R,R*)-**2** and **3a**.

Structural Characterization in Solution by ¹¹B NMR

Structural studies using ¹¹B NMR were carried out to investigate the effect of the solvent on the nature of the N-B interaction of complex **7** in solution. The ¹¹B NMR chemical shift is dependent on hybridization, charge, coordination, and the substituents present on the boron. Therefore, useful information about the structure of the molecule can be obtained by ¹¹B NMR analysis. The ¹¹B NMR signals of the complex and the receptor are assigned based on the comparison with our previous analysis of the *o*-(*N,N*-dialkylaminomethyl)arylboronate system.³⁴

The sample of crystals used to obtain the x-ray structure of complex **7** displays one ¹¹B signal at 31.3 ppm in CDCl₃ (Figure 3A). The signal at 31.3 ppm is assigned to the trigonal planar boron species with no N-B bond. However, in CD₃OD (Figure 3B), the same set of crystals of

complex **7** shows two peaks. The major peak (~90%) at 8.1 ppm is assigned to the fully solvated sp^3 hybridized boron species of complex **7** in solution. While, the minor signal (~20%) at 23.3 ppm corresponds to the ^{11}B NMR signal of the free host-**2** in methanol (See Supplementary Information).

Host-**2** by itself has a chemical shift of 28.8 ppm in CDCl_3 and 23.3 ppm in CD_3OD . The signal at 28.8 ppm was assigned to the trigonal planar boron in CDCl_3 . The signal at 23.3 ppm in CD_3OD could not be assigned to a single boron species in solution because previously, we have determined that in *o*-(*N,N*-dialkylaminomethyl)arylboronate system the N-B coordinated species has a chemical shift of ~ 14 ppm. We postulate that there is a fast exchange between the solvent inserted sp^3 hybridized boron and the sp^2 hybridized boron of host-**2** in methanol. Therefore, the signal at 23.3 ppm could not be assigned to a single structure of host-**2** in methanol.

As expected, complex **7** exist as the B- sp^3 hybridized solvated species in CD_3OD with a chemical shift of 8.1 ppm. However, the crystal structure shows the presence of the B- sp^2 hybridized species of complex **7**. In order to explain this discrepancy we hypothesize that there is an equilibrium between the solvent inserted sp^3 hybridized boron and sp^2 hybridized boron of complex **7** in solution, but the sp^2 hybridized species of complex **7** is what crystallizes out of solution.

In summary, we have studied for the first time the origin of enantioselectivity in our boronic acid based receptors, and a clear steric clash is seen in the unfavorable complex (Figure 1). Insight on the structure of the complex **7** in different solvent systems was also achieved. Complex **7** is B- sp^2 hybridized in CDCl_3 while there is an equilibrium between the sp^3 solvent inserted boron and sp^2 hybridized boron in CD_3OD .

Generating a Chemo- and Enantioselective Array

Before creating our pattern based recognition procedures, we needed to establish the proper pairings of hosts and indicators to best enantiodifferentiate the chiral diol target, as described below. However, we first determined binding constants of the hosts with indicators because we have found that the optical response of an enantioselective IDA is best when the host is about 90% saturated with an indicator.²⁹ Six catechol based indicators were used in the study described herein: alizarin (A), alizarin complexone dihydrate (AC), bromopyrogallol red (BPG), 4-methylscutletin (ML), pyrogallol red (PG), and pyrocatechol violet (PV). Host-**1** has been studied previously, and thus the ratio to achieve 90% of saturated host-**1** with each of the six indicators was calculated from literature binding constants.³⁰ The binding constants for the indicators with host-**2** are given in Table 1 (Figures S1-S6).

To establish the optimal host:indicator pairings for enantiodiscrimination, we screened host and indicator duos for colorimetric differences between the diol analytes. The concentration of each host that yields 90% saturation with each indicator was used in a 96-well plate (Table S1). The receptor/indicator pairs were treated with the diols, (Figure 4) such that the concentration of the diol was 5mM. The results of these plates identified three combinations: host-**1**:PV, host-**2**:PV and host-**2**:ML as good host:indicator duos for enantiomeric discrimination for all the diols, showing a $\Delta\text{Abs} > 0.05$ (Table 2-3). Host-**1**:BPG also shows good discrimination, with ΔAbs of 0.23 and ΔAbs of 0.16, between the enantiomers of **3** and **5** respectively, but the discrimination is poor for the enantiomers of diols **4** and **6**. This process allows us to rapidly identify the best host-indicator combination for the enantiodiscrimination of each chiral diol, and eliminates the need to carry out UV-vis titrations of each chiral diol with each host and indicator combination. The total number of UV-titrations that were circumvented is 96 as there are 8 diols, 2 host and 6 indicators.

Principal Component Analysis (PCA)

Our first goal was to demonstrate chemo and enantioselective discrimination of our analytes. As identified by the screening, our assay utilized three host:indicator combinations: (*S,S*)-**1**:PV, (*R,R*)-**2**:ML, and (*S,S*)-**2**:PV. Each diol at 5 mM concentration was treated with these three host:indicator combinations. The experiment was repeated four times to ensure reproducibility. A full spectrum of each sample was recorded, and the data was analyzed at nine different wavelengths: (496, 500, 516 nm) for host (*S,S*)-**1**:PV, (362, 366, 374 nm) for host (*R,R*)-**2**:ML, and (496, 500, 516 nm) for host (*S,S*)-**2**:PV (Table S2). These wavelengths were chosen based on the largest change in absorbance. The XLSTAT computer program was used to generate a PCA plot, which showed excellent discrimination for all the diols and their enantiomers (Figure 5). Tight clustering of identical samples and good spatial resolution for all the analytes was achieved. A slight rotation of the PCA axes showed chemoselectivity separated along PC1', while chirality is correlated along PC2'. Guests with *S* stereochemistry on the stereocenter near the phenyl (or ester of **6**) have positive scores along PC2', while those with *R* stereochemistry have negative scores. Furthermore, the extent of enantioselectivity of the sensor array ((*S,S*)-**1**:PV, (*R,R*)-**2**:ML, and (*S,S*)-**2**:PV) to the diols is also graphed on the PCA plot. As shown, the enantiomers of diol **3** are farthest apart with respect to PC2' as compared to the other diols, meaning that the array best discriminates the enantiomers of diol **3**. The enantiomers of diol **6**, however, are closer together with respect to PC2' and they are not as well discriminated by the array. This result is supported by the screening plate analysis where the largest Δ Abs is observed between the enantiomers of diol **3** with (*S,S*)-**1**:PV, (*R,R*)-**2**:ML, and (*S,S*)-**2**:PV combinations as compared to the other diols. Therefore, the PCA plot also graphically displays how well each diol is enantiomerically discriminated by the array.

Having found excellent chemo and enantioselectivity, the next goal was to explore the ability to simultaneously determine $[G]_t$ and *ee*. For this purpose, we arbitrarily chose diol **5**, while using the three different host indicator combinations. The earlier screening experiment discussed above for diol **5** determined that the best enantiomeric discrimination between **5a** and **5b** was achieved by host-**1**:BPG (Δ Abs 0.119), host-**2**:ML (Δ Abs 0.352) and host-**2**:PV (Δ Abs 0.232). Thus, *ee* titrations of diol-**5** with only these three combinations of host and indicators were carried out at three different concentrations (2 mM, 4 mM and 5 mM) and eight different *ee* values (-1.0, -0.6, -0.2, 0, 0.2, 0.4, 0.6, 1.0), where 1.0 is 100% **5a** and -1.0 is 100% **5b**. Each *ee* titration was carried out at one concentration, and repeated three times. The data set consisting of wavelengths (570, 572, 574, 576, 578 nm) for host (*S,S*)-**1**:BPG, (512, 514, 516, 518 nm) for host (*S,S*)-**2**:PV, and (376, 378, 380, 382, 384 nm) for host (*R,R*)-**2**:ML was analyzed using PCA (Table S3). Good spatial resolution was obtained in the PCA plot showing clustering of identical samples and spatial resolution of concentration and *ee* (Figure 6A). The data sets of varying *ee* values at the same concentration were clustered together in smooth curves in Figure 6A. *Ee* values range within the stripe from +1 to -1 left to right respectively (See legend of Figure 6A). In this analysis we did not have to use an achiral host to determine concentration of analytes in contrast to our previous studies. Using data from different host and indicator combinations enabled us to eliminate the achiral host from our sensor array, thus simplifying the analysis.²³

Artificial Neural Network (ANN) Analysis

The last goal was to analyze unknown samples for concentration and *ee*. A neural network was generated by using the same data set that was used to generate the PCA plot for diol **5**, but now as an ANN training set (Figure 6B). The repetitive data was not included in the training set as ANN does not work well with redundant values. Statistica Neural Network software was used to develop the ANN. It has an embedded intelligent problem solver (IPS) function that automatically generates several networks that are suitable for the designated problem at hand. The input layer contains the absorbance of each *ee* titration giving a total of 14 absorbance

values. The outputs are $[G]_t$ and % **5a**. ANN does not work well with negative values, therefore % **5a** was used instead of *ee* values. A multilayered perceptron (MLP) network with 14 inputs, 8 processing units in the hidden layer, and 2 outputs was selected for our analysis. The network was trained by back-propagation algorithms, which minimize the discrepancy between the input and the outputs. True unknown samples, prepared completely independent of the training set were then treated with the same sensor array and their $[G]_t$ and % **5a** values were predicted by the network (Table S4). The % **5a** values predicted by the network were converted to *ee* and are listed in Table 4. The average absolute error for $[G]_t$ was ± 0.08 mM and for *ee* was 6.72%. The *ee* error is high because of a single outlier that has an *ee* error of 19.38%. Without including the outlier, the average absolute error in *ee* drops to 3.57%. The *ee* error can be further reduced by using a larger training set as shown previously by our group.³⁰

Overall, the time required for this entire analysis from screening, to training, to analysis by ANN and PCA, required approximately 24 hrs of work. The method is very rapid once an ANN network has been developed by using a training set for a particular analyte. The concentration and *ee* values of 96 unknown samples can be determined in 40 mins, as it takes 30 min for the pipetting system to load a 96-well plate with the host, indicator, and unknown solutions, and 10 min for the 96-well plate reader to record the absorbances of 96 unknown samples at five different wavelengths needed for the ANN analysis. The absorbance data of unknown samples can be channeled to the developed ANN, which in real time gives the concentration and *ee* of unknown samples. In effect, this assay is a powerful tool which we are now transitioning to a high-throughput study to determine identity, $[G]_t$, and *ee* of unknown samples of chiral vicinal diols.

CONCLUSION

We have described a technique that fingerprints chemical identity, concentration, and chirality of chiral vicinal diols. Pattern recognition protocols were used to analyze the data. Excellent fingerprinting was obtained where the diols are chemoselectively and enantioselectively separated. This was achieved by using a diverse sensor array made up of different host and indicator combinations. We have also shown that a highly enantioselective array that does not require the presence of an achiral boronic acid host can be used to calculate concentration of unknown samples. A family of chiral hosts and indicator combinations with ANN can provide both concentration and *ee* values of unknown samples with high accuracy. In addition, a new enantioselective boronic acid host-**2** has been designed and synthesized in one step that showed excellent enantioselectivity for several chiral vicinal diol. To study the enantioselective behaviour of our boronic acid host, we solved an x-ray structure of host (*R,R*)-**2** with (*S,S*)-hydrobenzoin (**3b**) (complex **7**). Studies of the solution phase behaviour of the complex by ¹¹B NMR lent some insight on the enantioselective behaviour of host-**2** and suggests that the hybridization of the boronic acid species depends both on the solvent and the substrate, but also depends on the structure of the boronic acid receptor.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

ACKNOWLEDGEMENT

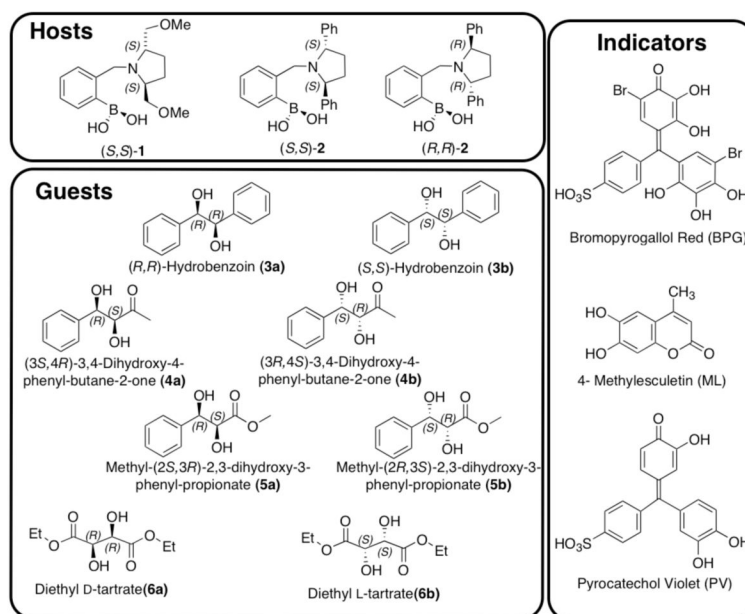
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**Scheme 1.**

Structures of hosts, guests/analytes and indicators used in this study.

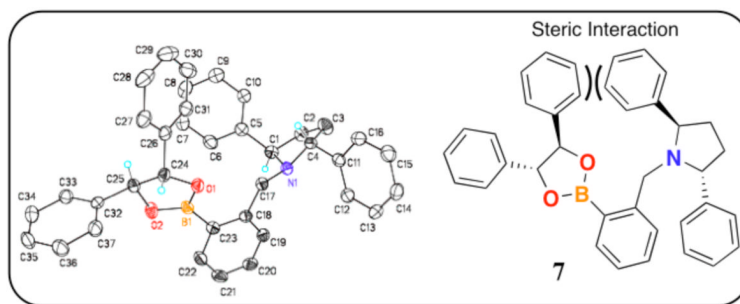


Figure 1. Left: Crystal structure of host (*R,R*)-**2** complex with (*S,S*)-hydrobenzoin (**3b**). Right: ChemDraw representation of the crystal structure (**7**).

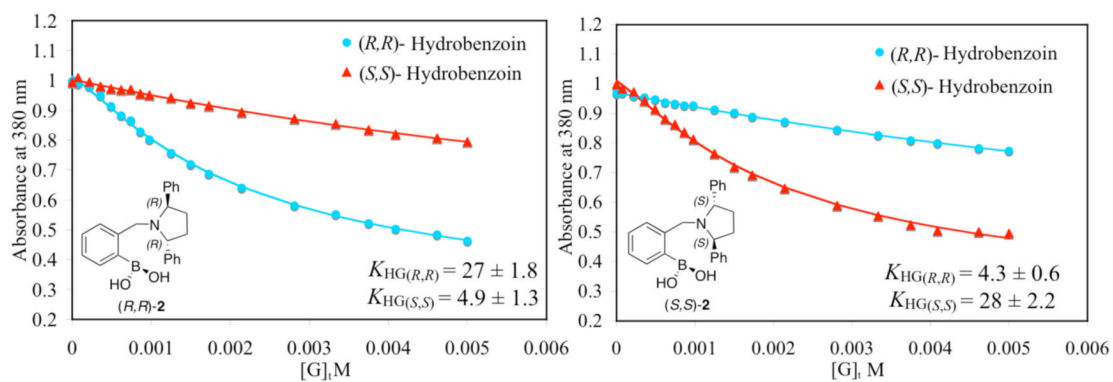


Figure 2.

Enantioselective indicator displacement assay (eIDA) of host-2 ($440 \mu\text{M}$) /ML ($75 \mu\text{M}$) with (R,R) -hydrobenzoin and (S,S) -hydrobenzoin. All titrations were carried out in 100% MeOH, 10 mM *para*-toluenesulfonic acid and Hunig's base buffer, pH 7.4, association constant K_{HG} (10^2M^{-1}). All measurements were taken at 25°C . The solid lines are calculated curves resulting from iterative data fitting for a displacement assay.³⁸

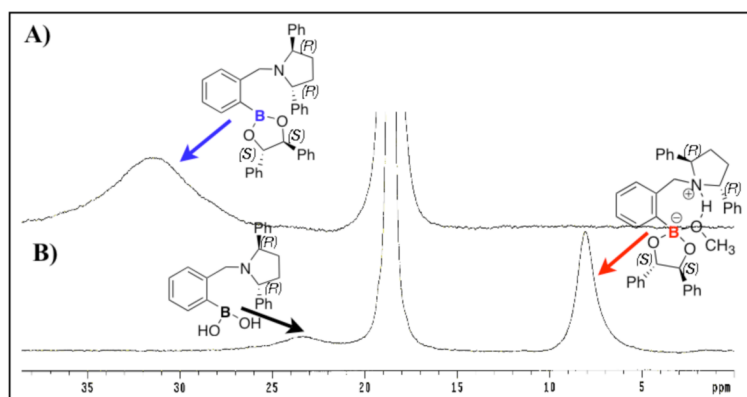


Figure 3. ^{11}B NMR spectra of complex **7** CDCl_3 (A) and CD_3OD (B). The signal at 18.6 ppm is the external standard trimethyl borate.

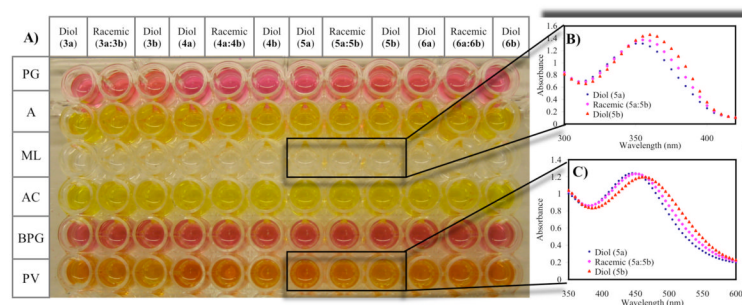


Figure 4.

A) Method used to screen for host (*R,R*)-**2** with indicators: pyrogallol red (PG), alizarin (A), 4-methylesculetin (ML), alizarin complexone dihydrate (AC), bromopygallol red (BPG), and pyrocatechol violet (PV). The concentration of the indicators and the host are listed in Table S1. (*R,R*)-hydrobenzoin (**3a**), (*S,S*)-hydrobenzoin (**3b**), (*3S,4R*)-2,3-dihydroxy-4-phenyl-butane-2-one (**4a**), (*3R,4S*)-2,3-dihydroxy-4-phenyl-butane-2-one (**4b**), methyl(*2S,3R*)-2,3-dihydroxy-3-phenyl-propionate (**5a**), methyl(*2R,3S*)-2,3-dihydroxy-3-phenyl-propionate (**5b**), diethyl *D*-tartrate (**6a**), diethyl *L*-tartrate (**6b**) **B)** Absorption spectra of host (*R,R*)-**2**:ML with diol **5a**;Racemic mixture (**5a:5b**); diol **5b**. **C)** Absorption spectra of host (*R,R*)-**2**:PV with diol **5a**;Racemic mixture (**5a:5b**); diol **5b**. All studies were carried out in 100% MeOH, 10 mM buffered solution of Hunig's base and *para*-toluenesulfonic acid to pH 7.4 at 25°C.

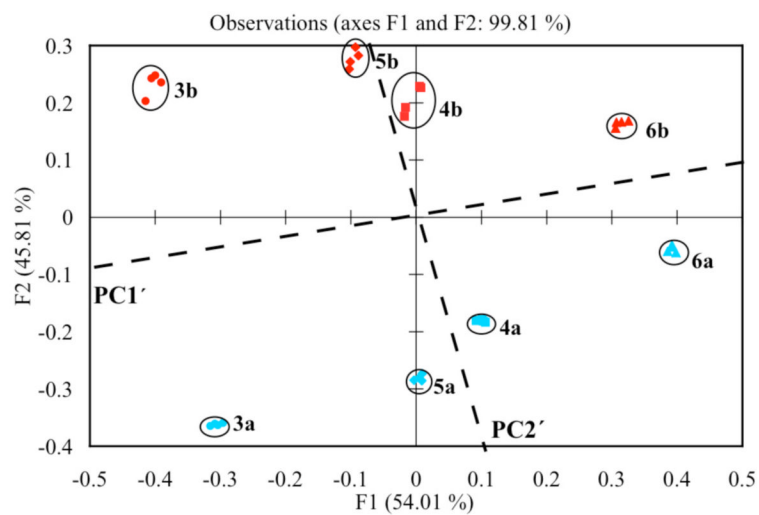


Figure 5. PCA of the enantiomers of the four diols analyzed with host (*S,S*)-**1**:PV host (*R,R*)-**2**:ML, and host (*S,S*)-**2**:PV. The diols are labeled on the PCA plot. All studies were carried out in 100% MeOH, 10 mM buffered solution of Hunig's base and *para*-toluenesulfonic acid to pH 7.4 at 25 °C.

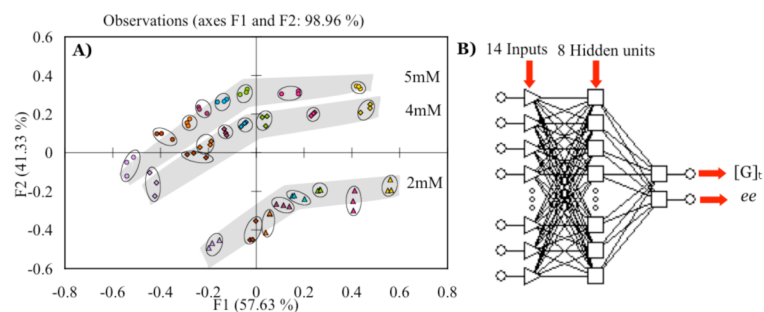


Figure 6.

A) PCA of diol **5** *ee* titration at three different concentrations Δ : 2 mM, \diamond : 4 mM, \circ : 5 mM. *ee*, yellow: -1, pink: -0.6, green: -0.2, blue: 0, red: 0.2, orange: 0.4, brown: 0.6, purple: 1, where 1 is 100% methyl(2*S*,3*R*)-(-)-2,3-dihydroxy-3-phenylpropionate (**5a**) and -1 is 100% methyl(2*R*,3*S*)-(-)-2,3-dihydroxy-3-phenylpropionate (**5b**) **B)** Multilayer perceptron (MLP) artificial neural network generated for the determination of *ee* and analyte total concentration ($[G]_t$). All studies were carried out in 100% MeOH, 10mM buffered solution of Hunig's base and *para*-toluenesulfonic acid to pH 7.4 at 25°C.

Table 1Binding constant K_{HI} ($10^4 M^{-1}$) of the host (S,S)-**2** with the indicators.

Indicators	A	AC	BPG	ML	PG	PV
(S,S)- 2	8.42±1.76	6.41±2.21	191±131	3.61±0.13	1.02±00.7	1.36±0.02

Table 2

Screening plate results. Δ Abs between the enantiomeric guests (5 mM) with host (*S,S*)-**1** and the indicators calculated for: A at 450 nm, AC at 550 nm, ML at 380 nm, BPG at 570 nm, PG at 520 nm and PV at 520 nm.

Indicators	Diol (3)	Diol (4)	Diol (5)	Diol (6)
A	0.007	0.035	0.024	0.007
AC	0.026	0.025	0.008	0.014
BPG	0.233	0.054	0.119	0.038
ML	0.145	0.061	0.102	0.043
PG	0.031	0.031	0.030	0.007
PV	0.107	0.080	0.072	0.057

Table 3

Screening plate results. Δ Abs between the enantiomeric guests (5 mM) with host (*R,R*)-**2** and the indicators calculated for: A at 450 nm, AC at 550 nm, ML at 380 nm, BPG at 570 nm, PG at 520 nm and PV at 520 nm.

Indicators	Diol (3)	Diol (4)	Diol (5)	Diol (6)
A	0.176	0.092	0.158	0.043
AC	0.161	0.042	0.107	0.002
BPG	0.251	0.046	0.156	0.018
ML	0.357	0.208	0.352	0.126
PG	0.005	0.010	0.031	0.112
PV	0.216	0.166	0.232	0.139

Table 4

Artificial Neural Network Analysis of unknown solutions

$[G]_t$, mM	ANN $[G]_t$, mM	Error $[G]_t \pm$ mM	% ee	ANN % ee	Error % ee
3.00	3.07	0.07	50.00	69.38	19.38
3.00	3.04	0.04	30.00	33.10	3.10
3.00	3.05	0.05	10.00	8.47	1.53
3.00	3.04	0.04	-30.00	-20.64	9.36
3.00	2.79	0.21	-50.00	-50.27	0.27