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Octopaminergic agonists for the cockroach neuronal octopamine receptor

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Abstract

The compounds 1-(2,6-diethylphenyl)imidazolidine-2-thione and 2-(2,6-diethylphenyl)imidazolidine showed the almost same activity as octopamine in stimulating adenylate cyclase of cockroach thoracic nervous system among 70 octopamine agonists, suggesting that only these compounds are full octopamine agonists and other compounds are partial octopamine agonists. The quantitative structure-activity relationship of a set of 22 octopamine agonists against receptor 2 in cockroach nervous tissue, was analyzed using receptor surface modeling. Three-dimensional energetics descriptors were calculated from receptor surface model/ligand interaction and these three-dimensional descriptors were used in quantitative structure-activity relationship analysis. A receptor surface model was generated using some subset of the most active structures and the results provided useful information in the characterization and differentiation of octopaminergic receptor.

Keywords: Periplaneta americana, receptor surface model, cerius2, octopamine agonist

Abbreviation:

AEA arylethanolamine
AII 2-(arylimino)imidazolidine
AIO 2-(arylimino)oxazolidine
AIT 2-(arylimino)thiazolidine
APAT 2-(α-phenylethylamino)-2-thiazoline
BPAT 2-(β-phenylethylamino)-2-thiazoline
CAO 2-(3-chlorobenzylamino)-2-oxazoline
DCAO 2-(3,5-dichlorobenzylamino)-2-oxazoline
DETS 2-(2,6-diethylphenylimino)-5-methylthiazolidine
DET6 2-(2,6-diethylphenylimino)thiazine
EGTA ethylene glycol bis(β-aminoethyl ether)-N,N′,N′′,N′′'-tetraacetic acid
GFA genetic function approximation
G/PLS genetic partial least squares
IND 2-aminomethyl-2-indanol
LAH lithium aluminum hydride
MCSG maximum common subgroup
MCT6 2-(2-methyl-4-chlorophenylimino)thiazine
OA octopamine
PLS partial least squares
QSAR quantitative structure-activity relationship
SBAT 2-(substituted benzylamino)-2-thiazoline

Abbreviations continued on next page
Octopamine [2-amino-1-(4-hydroxyphenyl)ethanol] which has been found to be present in high concentrations in various insect tissues, is the monohydroxylic analogue of the vertebrate hormone noradrenalin. Octopamine was first discovered in the salivary glands of the octopus by Erspamer and Boretti (1951). It has been found that octopamine is present at a high concentrations in various invertebrate tissues (Axelrod and Saavedra, 1977). This multifunctional and naturally occurring biogenic amine has been well studied and established as 1) a neurotransmitter, controlling the firefly light organ and endocrine gland activity in other insects; 2) a neurohormone, inducing mobilization of lipids and carbohydrates; 3) a neuromodulator, acting peripherally on muscles, fat body, corpora cardiaca and the corpora allata, and 4) a centrally acting neuromodulator, influencing motor patterns, habituation and even memory in various invertebrate species (Evans, 1985, 1993).

The action of octopamine is mediated through various receptor classes. Three different receptor classes OAR1, OAR2A and OAR2B have been distinguished from non-neuronal tissues (Evans, 1981). OAR2 is coupled to G-proteins and is specifically linked to an adenylate cyclase. Thus, the physiological actions of OAR2 have been shown to be associated with elevated levels of cAMP (Nathanson, 1985). In the nervous system of the locust, Locusta migratoria, a particular receptor class was characterized and established as a new class OAR3 by pharmacological investigations of the [3H]OA binding site using various agonists and antagonists (Roeder, 1990, 1992, 1995; Roeder and Gewecke, 1990; Roeder and Nathanson, 1993).

Recently much attention has been directed at the octopamine agonist as a valid target in the development of safer and selective pesticides (Jennings et al., 1988; Hirashima et al., 1992; Ismail et al., 1996). Structure-activity studies of various types of octopamine agonists and antagonists were also reported using the nervous tissue of the migratory locust, L. migratoria (Roeder, 1990, 1992, 1995; Roeder and Gewecke, 1990; Roeder and Nathanson, 1993). However, information on the structural requirements of these octopamine agonists and antagonists for high octopamine-receptor ligands is still limited. It is therefore of critical importance to provide information on the pharmacological properties of octopamine receptor types and subtypes. Our interest in octopaminergic agonists was aroused by the results of quantitative structure-activity relationship (QSAR) studies using various physicochemical parameters as descriptors (Hirashima et al., 1999a; Pan et al., 1997a) and receptor surface modeling (Hirashima et al., 1998a).

Furthermore, molecular modeling and conformational analyses were performed in Catalyst/Hypo to gain a better understanding of the interactions between octopaminergic antagonists and OAR3 in order to determine the conformations required for binding activity (Pan et al., 1997b). A similar procedure was repeated using octopamine agonists (Hirashima et al., 1999b). However, binding activity is not enough for evaluating octopamine-agonist activity, because it is difficult to determine the different activities of octopamine agonists and antagonists. In drug discovery, it is common to have measured activity data for a set of compounds acting upon a particular protein but not to have knowledge of the three-dimensional structure of the protein active site. In the absence of such three-dimensional information, one can attempt to build a hypothetical model of the receptor site that can provide insight about receptor site characteristics. Such a model is known as a receptor surface model, which provides compact and quantitative descriptors that capture three-dimensional information about a putative receptor site. Thus, the current work is aimed to perform 3D receptor surface modeling on a set of octopamine agonists using the thoracic nervous system of the American cockroach, Periplaneta americana, in which octopamine-agonist action is thought to be due to cAMP elevation at OAR2 (Hirashima et al., 1992).

Materials and Methods

Synthesis of OA agonists

AEAs 1-7, IND 39 and THI 40 were synthesized from trimethylsilyl cyanide and the corresponding aldehyde in the presence of catalytic amount of anhydrous zinc iodide, followed by reduction with LAH (Hirashima et al., 1990). AEAs 23 and 24 were obtained by reducing optically active mandelic amides, which was obtained from optically active manellic acids via their esters, with LAH (Waet al., 1989). AITs 9-11, 41-43, MCT6 44, DET5 45, DET6 46, SBATs 12-14, 47-57, APATs 58-59 and BPAT 60 were synthesized by cyclization of the corresponding thiourea with conc. hydrogen chloride (Hirashima et al., 1994). AIOs 15-16, 61-65, CAO 17 and DCAO 66 were obtained by cyclodesulfurizing the corresponding thiourea with yellow mercuric oxide (Hirashima et al., 1996). SPITs 18 and 67-70 were synthesized by the cyclization of monoethanolamine hydrogen sulfate with arylisothiocyanate in the presence of sodium hydroxide as described in the previous report (Hirashima et al., 1998b). AIs 19-22 were prepared according to a reported method by refluxing the corresponding substituted aniline and 1-acetyl-2-imidazolidone in phosphoryl chloride followed by hydrolysis (Nathanson and Kaugars, 1989). The structures of the compounds were confirmed by 1H and 13C NMR measured with a JEOL JNM-EX400 spectrometer at 400 MHz, TMS being used as an internal standard for 1H NMR and elemental analytical data.

Chemicals

Octopamine, theophylline (1,3-dimethylxanthine), tyramine 41 and EGTA were purchased from Nacalai Tesque, www.nacalai.co.jp/en/; GTP and DL-synephrine 8 was from Sigma Chemical Co., www.sigmaaldrich.com; ATP disodium salt was from Kojjin Co., www.kojjin.co.jp/english/; LAH was from Chemetall GmbH, www.chemetall.com. The cAMP radioimmunoassay kit (cord RPA 509) was purchased from Amershams International.

Insect culture

Males and females of *Periplaneta americana* were used indiscriminately, as their nervous systems exhibited no gross structural or neurochemical differences. The insects were reared under crowded conditions in this laboratory at 28 °C with a photoperiod of 12 h light:12 h dark and at a relative humidity of 65-70% for more than 7 years. They were provided with an artificial mouse diet (Oriental Yeast Co., Chiba, Japan, www.oyc.co.jp/e/) and water ad libitum.

Adenylate-cyclase assay

The adenylate-cyclase assay was conducted on adult *P. americana* as described previously (Hirashima et al., 1992, 1994, 1996, 1998b). Thoracic nerve cords of *P. americana* were homogenized (15 mg/ml) in a 6 mM Tris-maleate buffer (pH 7.4) using a chilled microtube homogenizer (S-203, Ikeda Sci., Tokyo, Japan, www.biorobotics.com/Pages/ikeda.html) as described previously. The homogenate was diluted (1 mg/ml) in 6 mM Tris-maleate, centrifuged at 120,000 g at 4 °C for 20 min. The supernatant was discarded, the pellet was resuspended by homogenizing (1 mg/ml) in the buffer, and again centrifuged at 120,000 g at 4 °C for 20 min. The resulting pellet (P2) was resuspended in the buffer equivalent to the starting amount (15 mg/ml). The adenylate-cyclase activity was measured according to Nathanson’s procedure under optimal conditions (Hirashima et al., 1992, 1994, 1996, 1998b) in a test tube containing 200 µl of 120 mM Tris-maleate (pH 7.4, including 15 mM theophylline, 12 mM MgCl₂, and 0.75 mM EGTA), 60 µl of the P2 fraction and 30 µl of each synthesized compound solution in polyethylene glycol. An appropriate solvent control was run in parallel. The enzyme reaction (5 min at 30°C) was initiated by adding 10 µl of a mixture of 3 mM GTP and 60 mM ATP, stopped by heating at 90 °C for 2 min and then centrifuged at 1000 g for 15 min to remove the insoluble material. The cAMP level in the supernatant was measured using a radioimmunoassay (Hirashima et al., 1992, 1994, 1996, 1998b). Protein concentration was determined by the Lowry method (Lowry et al., 1951), using bovine serum albumin (Sigma) as the standard. Enzyme activity in each assay was corrected using octopamine as a reference. The Vmax values (mostly at 0.1 mM) were calculated relative to octopamine (100%) and control (0%).

Computational details

Molecular alignment: All experiments were conducted with Cerius2 3.8 QSAR environment from Accelrys, www.accelrys.com, on a Silicon Graphics O2, running under the IRIX 6.5 operating system. Multiple conformations of each molecule were generated using the Boltzmann Jump as a conformational search method. The upper limit of the number of conformations per molecule was 150. Each conformer was subjected to an energy minimization procedure to generate the lowest energy conformation for each structure. Alignment of structures through pair-wise superpositioning placed all structures in the study compounds in the same frame of reference as the shape reference compounds, which were selected as a conformer of the most active octopamine agonists. The method used for performing the alignment was MCGS. This method looks at molecules as points and lines, and uses the techniques of graph theory to identify patterns. It finds the largest subset of atoms in the shape reference compound that is shared by all the structures in the study table and uses this subset for alignment. A rigid fit of atom pairings was performed to superimpose each structure so that it overlays the shape reference compound.

Receptor surface modeling: Receptor surface models are predictive and sufficiently reliable to guide the chemist in the design of novel compounds. These descriptors are used for predictive QSAR (quantitative structure-activity relationship) models. This approach is effective for the analysis of data sets where activity information is available but the structure of the receptor site is unknown. Receptor surface modeling attempts to postulate and represent the essential features of a receptor site from the aligned common features of the molecules that bind to it. This method generates multiple models that can be checked easily for validity. Once a reasonable receptor surface model has been defined, a series of structures can be evaluated against the model. When a receptor model has been generated and the models have been aligned, a QSAR can be built using data from the receptor-structure interactions. The results of the minimization procedure were used as descriptors either to refine the model or to predict activity. Three-dimensional energetics descriptors were calculated from receptor surface model/ligand interaction. These three-dimensional descriptors were used in QSAR analysis.

A receptor surface model represents the global volume that can accommodate one or more molecules and can be seen as the shape of an active site built from the ligands that fit into it in their “active” conformation. The descriptors used in this study account for phenomena that occur at the contact surface between the ligands and the protein active site. A receptor surface model represents essential information about the hypothetical receptor site as a three-dimensional surface with associated properties mapped onto the surface model. The location and shape of the surface represent information about the steric nature of the receptor site: the associated properties represent other information of interest, such as hydrophobicity, partial charge, electrostatic potential and hydrogen-bonding propensity. The isosurface procedure produces a surface that entirely encloses the molecules over which it is generated. The surface has no holes and is known as a closed model. Receptor surface models are best constructed from a set of the most active analogues that are chosen to cover the variety of steric and electrostatic variations likely to appear in the test data. The approach we took was to automatically build a set of different receptor surface models from different combinations of the most active analogues, and then use a variable-selection technique such as G/PLS to discover the receptor surface model whose descriptors yield the best QSARs of the full training set. G/PLS allows the discovery and use of nonlinear descriptors by using spline-based terms.

Genetic partial least squares: Due to the large number of points used as independent variables, genetic partial least squares (G/PLS) was used to derive the QSAR models. G/PLS, a variation of genetic function approximation (GFA), can be run as an alternative to the standard GFA algorithm. G/PLS is derived from the best
features of two methods: GFA and partial least squares (PLS). Both GFA and PLS have been shown to be valuable analysis tools in cases where the data set has more descriptors than samples. In PLS, variables might be overlooked during interpretation or in designing the next experiment even though cumulatively they are very important. This phenomenon is known as “loading spread”. In GFA, equation models have a randomly chosen proper subset of the independent variables. As a result of multiple linear regression on each model, the best ones become the next generation and two of them produce an offspring. This was repeated 10000 (default 5000) times. For other settings, all defaults were used. Loading spread does not occur because at most one of a set of co-linear variables is retained in each model. G/PLS combines the best features of GFA and PLS (Cerius2 tutorial. Accelrys Inc., www.accelrys.com/cerius2). G/PLS retains the ease of interpretation of GFA by back-transforming the PLS components to the original variables.

Results

A set of 22 molecules, shown in Table 1, whose stimulatory activities of adenylate cyclase were tested using the receptor surface model in the cockroach central nervous tissue, and whose Vmax values were 40% or larger than 40%, were selected as the target training set. The molecular structures and experimental biological values were 40% or larger than 40%, were selected as the target model in the cockroach central nervous tissue, and whose Vmax activities of adenylate cyclase were tested using the receptor surface model designed for log dose-probit activity analyses. The maximal response relative to octopamine and was expressed as

Table 1. Effect of OA agonists on the adenylate-cyclase activity in homogenates of American cockroach nerve cords in study set a

<table>
<thead>
<tr>
<th>Compound</th>
<th>Vmax (%</th>
<th>Ka (µM)</th>
<th>Observed</th>
<th>Calculated</th>
<th>Deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Synephrine</td>
<td>57.6±5.6</td>
<td>150.0±125(77-179.70)</td>
<td>3.8</td>
<td>3.9</td>
<td>-0.08</td>
</tr>
<tr>
<td>2-Me,4-Br</td>
<td>41.1±1.1</td>
<td>23.6±20.19-27.83</td>
<td>4.6</td>
<td>4.5</td>
<td>0.09</td>
</tr>
<tr>
<td>2,4-Me2</td>
<td>55.0±3.9</td>
<td>13.6±12.80-14.54</td>
<td>4.8</td>
<td>4.8</td>
<td>0.01</td>
</tr>
<tr>
<td>2,6-Et2</td>
<td>50.2±2.2</td>
<td>26.1±25.04-27.26</td>
<td>4.5</td>
<td>4.5</td>
<td>0.03</td>
</tr>
<tr>
<td>2,6-Me2</td>
<td>57.9±1.8</td>
<td>16.8±13.92-20.19</td>
<td>4.7</td>
<td>4.8</td>
<td>-0.06</td>
</tr>
<tr>
<td>2-Et,6-Cl</td>
<td>42.5±0.8</td>
<td>13.16(10.39-16.54)</td>
<td>4.8</td>
<td>4.8</td>
<td>0.04</td>
</tr>
<tr>
<td>CAO</td>
<td>40.0±1.1</td>
<td>5.07±(4.12-66.14)</td>
<td>5.2</td>
<td>5.1</td>
<td>0.14</td>
</tr>
<tr>
<td>5,6-Cl2</td>
<td>45.7±4.6</td>
<td>16.2±13.41-19.80</td>
<td>4.7</td>
<td>4.7</td>
<td>-0.06</td>
</tr>
<tr>
<td>2,6-Et2</td>
<td>104.0±1.6</td>
<td>8.08±6.32-10.26</td>
<td>5.09</td>
<td>5.16</td>
<td>0.07</td>
</tr>
<tr>
<td>2,6-Me2</td>
<td>47.7±7.6</td>
<td>15.0±13.18-19.92</td>
<td>4.8</td>
<td>4.9</td>
<td>-0.06</td>
</tr>
<tr>
<td>2-Me,6-Et</td>
<td>47.6±2.7</td>
<td>9.5±7.48-12.10</td>
<td>5.02</td>
<td>5.04</td>
<td>0.02</td>
</tr>
<tr>
<td>2,6-Et2</td>
<td>104.6±1.4</td>
<td>7.06±6.49-11.13</td>
<td>6.12</td>
<td>6.12</td>
<td>0</td>
</tr>
<tr>
<td>2,4,6-Me2</td>
<td>59.1±10.2</td>
<td>8.40±5.68-12.34</td>
<td>5.08</td>
<td>5.06</td>
<td>0.02</td>
</tr>
</tbody>
</table>

*See the text for experimental details. The adenylate-cyclase activity of P. americana was measured according to Nathanson’s procedure and the cAMP levels were measured by a RIA. Basal (control) and OA (100 µM)-stimulated (Vmax) adenylate cyclase activities were 116.2±0.8, 13.16 (10.39-16.54) and 4.88, 4.84, 0.04 µmol cAMP/min/mg of protein, respectively. In parentheses, 95% confidence limits are shown. Ka and 95% confidence limit values were calculated with the Macintosh personal computer system using a sigmoidal curve-fitting program designed for log dose-probit activity analyses.

*Calculated by Equation (1).

*In case predicted activity is underestimated, deviation is obtained by subtracting experimental value from predicted value and indicated by minus. In case predicted activity is overestimated, deviation is obtained by subtracting predicted value from experimental value.

activities are listed in Fig. 1a and Table 1, respectively. Vmax means the maximal response relative to octopamine and was expressed as the mean of 4 independent experiments. Values of pKa (log of the reciprocal of Ka) were used as octopamine-agonist activity index. Ka is the concentration of octopamine agonist necessary for half-maximal activation of adenylate cyclase. The Ka values were calculated from at least 5 concentrations ranging from 10^-7 M to 10^3 M using a Macintosh personal computer system, and a sigmoidal curve-fitting program designed for log dose-probit activity analyses. 2-(2,6-Diethylphenyl)imidazolidine 21 showed the highest activity followed by CAO 17, OA 5 and 1-(2,6-diethylphenyl)imidazolidine-2-thione 18 in study compounds. The compounds 18 and 21 showed the almost same Vmax as octopamine, suggesting that only these compounds are full octopamine agonists and other compounds are partial octopamine agonists.

A receptor surface model was generated (Figs. 2 and 3)
The octopamine agonist 21 with the highest activity and OA 5 embedded in a receptor surface model generated from 5, 18 and 21 computed using the Wyvill steric function colored by H-bond contribution. The purple sign stands for a positive contribution of H-bond and a light-blue sign stands for a negative contribution of H-bond.

Figure 3. The octopamine agonist 18 and OA 5 embedded in a receptor surface model generated from 5, 18 and 21 computed with the Wyvill steric function.

The best model generated using the descriptors from the closed receptor surface model is given in Equation (1), which was statistically significant and used to correctly predict the activities of a set of training molecules ranging over 3 orders of magnitude (max. pKa 6.12 and min. pKa 2.82), indicating that these models could be useful tools to design active octopamine agonists. The energies of interaction between the receptor surface model and each molecular model were added to the study table as new columns, which were used for generating QSARs. Instead of one total number which is the sum of the interactions evaluated between each point on the surface and each molecular model, leading to one extra column in the study table, the energies at each surface point are available. Depending on the size of the drug molecules, this is potentially a great number of surface points. The number of variables for Equation (1) was 986. Ten percent of all new significant columns of variables were automatically used as independent X variables in the generation of QSAR. In order to quantitatively understand the dependence of biological activities on receptor surface model parameters of octopamine agonists, regression analysis was applied to representative 22 study compounds (Vmax > 40%) listed in Fig. 1a and Table 1, leading to Equation (1).

\[ p_{Ka} = 4.82948 + 2.46196(ELE/567 - 0.137016)^2 + 3.92126(ELE/1565 - 1.19472)(ELE/1670 - 3.59779)(ELE/2177 + 3.86016)(ELE/3002 - 7.92195)(VDW/678)^2 - 0.898844(VDW/1404 + 0.010941) - 3.36632(TOT/678)^2 - 1.68885(TOT/1866 + 0.243149) - 4.86016(TOT/2746 + 0.245914) - 2.94225(TOT/2746 + 0.114432) + 3.76445(TOT/2755 + 0.283752) + 4.13854(TOT/2755 + 0.267161) \]

where \( n = 22, r^2 = 0.971, CV-r^2 = 0.720 \), predicted sum of squares = 2.998 and Bsr\(^2\) = 0.968±0.001. The descriptors ELE/567, ELE/1565...
etc are the electrostatic interaction energy of the molecule with the receptor at points 567, 1565 etc. The descriptors VDW/678 and VDW/1404 are the Van der Waals interaction energy of the molecule with the receptor at point 678 and 1404 etc. The descriptors TOT/678, TOT/1866 etc are the added energy of both electrostatic interaction energy and Van der Waals interaction energy at point 678, 1866 etc. The term \( n \) means the number of data points; \( r \)-squared (\( r^2 \)), the square of the correlation coefficient, which is used to describe the goodness of fit of the data of the study compounds to the QSAR model; cross-validated \( r^2 \) (CV-\( r^2 \)), a squared correlation coefficient generated during a validation procedure using the equation: 
\[
CV-r^2 = \frac{(SD - predicted sum of squares)}{SD}; 
\]
SD, the sum of squared deviations of the dependent variable values from their mean; predicted sum of squares, the sum of overall compounds of the squared differences between the actual and the predicted values for the dependent variables. The predicted sum of squares value is computed during a validation procedure for the entire training set. The larger the predicted sum of squares value, the more reliable is the equation. A CV-\( r^2 \) is usually smaller than the overall \( r^2 \) for a QSAR equation. It is used as a diagnostic tool to evaluate the predictive power of an equation generated using the G/PLS method. Cross-validation is often used to determine how large a model (number of terms) can be used for a given data set. For instance, the number of components retained in G/PLS can be selected to be that which gives the highest CV-\( r^2 \). Bootstrap \( r^2 \) (Bsr\( r^2 \)) is the average squared correlation coefficient calculated during the validation procedure (Cerius2 tutorial, Accelrys Inc., www.accelrys.com/cerius2). A Bsr\( r^2 \) is computed from the subset of variables used one-at-a-time for the validation procedure. It can be used more than one time in computing the \( r^2 \) statistic.

Table 1 depicts structures of octopamine agonists, their experimental \( K_a \) values, calculated \( pK_a \) values using Equation (1), and differences between observed and calculated \( pK_a \) values. In case predicted activity is overestimated, deviation is obtained by calculating predicted activity subtracted by experimental value and indicated by a minus sign. In case predicted activity is underestimated, deviation is obtained by calculating experimental activity subtracted by predicted value. Residuals (observed versus calculated from Table 1) are plotted in Fig. 4.

Once the desired receptor surface model had been constructed, all the structures in the test sets were evaluated against the model. The evaluation consists of computing several energetic descriptors that are based upon the interactions between ligand and model. By using receptor data to develop a QSAR model, the goodness of fit can be evaluated between a candidate structure and a postulated pseudo-receptor. The predictive character of the QSARs was further assessed using test molecules, whose Vmax values are less than 40% and whose structures are shown in Fig. 1b, outside of the training set. The best statistically significant Equation (1) was applied to access these octopamine agonists. The predicted values of these molecules are listed in Table 2. The octopamine agonists showed reasonable predicted activities according to Equation (1) in intracellular cAMP production. Residuals (observed versus calculated from Table 2) are plotted in Fig. 5. These results may imply that the process of calculating a receptor surface model treats these structures reasonably, although they have low Vmax values.

Discussion

QSAR modeling is an area of research pioneered by Hansch and Fujita (1964), Hansch and Leo (1995), and Golender and Vorpagel (1993). QSAR attempts to model the activity of a series of compounds using measured or computed properties of the compounds. Receptor surface models are quantitative and differ from pharmacophore models, which are qualitative, in that the former...
Table 2. Effect of OA agonists on the adenylate-cyclase activity in homogenates of American cockroach nerve cords in test set

<table>
<thead>
<tr>
<th>Compound</th>
<th>Vmax (%)</th>
<th>( K_a ) (( \mu )M)</th>
<th>( \mu K_a )</th>
<th>Observed</th>
<th>Calculated</th>
<th>Deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>AAT</td>
<td>20.9±1.0</td>
<td>29.51 (17.86-56.69)</td>
<td>4.53</td>
<td>3.6</td>
<td>0.93</td>
<td></td>
</tr>
<tr>
<td>H</td>
<td>24.3±1.5</td>
<td>192.02 (105.91-315.60)</td>
<td>3.72</td>
<td>3.84</td>
<td>-0.12</td>
<td></td>
</tr>
<tr>
<td>2-F</td>
<td>29.1±0.6</td>
<td>162.88 (122.07-230.02)</td>
<td>3.79</td>
<td>3.55</td>
<td>0.24</td>
<td></td>
</tr>
<tr>
<td>3-CF$_3$</td>
<td>26.1±3.3</td>
<td>69.55 (44.89-102.85)</td>
<td>4.16</td>
<td>4.58</td>
<td>-0.42</td>
<td></td>
</tr>
<tr>
<td>4-MeO</td>
<td>28.5±0.6</td>
<td>4.65 (3.06-6.99)</td>
<td>5.33</td>
<td>5.37</td>
<td>-0.04</td>
<td></td>
</tr>
<tr>
<td>2,4-Cl$_2$</td>
<td>34.3±0.9</td>
<td>27.55 (18.99-38.81)</td>
<td>4.56</td>
<td>4.52</td>
<td>0.04</td>
<td></td>
</tr>
<tr>
<td>2,4-Cl$_4$</td>
<td>29.2±1.9</td>
<td>22.63 (18.86-27.11)</td>
<td>4.65</td>
<td>3.99</td>
<td>0.66</td>
<td></td>
</tr>
<tr>
<td>2,5-Cl$_2$</td>
<td>27.9±2.2</td>
<td>8.91 (5.15-14.35)</td>
<td>5.05</td>
<td>4.37</td>
<td>0.68</td>
<td></td>
</tr>
<tr>
<td>2,6-(MeO)$_2$</td>
<td>8.1±0.6</td>
<td>59.09 (23.25-141.57)</td>
<td>4.23</td>
<td>3.34</td>
<td>0.89</td>
<td></td>
</tr>
<tr>
<td>3-CI,4-F</td>
<td>12.3±0.7</td>
<td>10.74 (3.93-22.72)</td>
<td>4.97</td>
<td>4.24</td>
<td>0.73</td>
<td></td>
</tr>
<tr>
<td>3,4-F$_2$</td>
<td>19.3±2.9</td>
<td>80.36 (59.83-111.14)</td>
<td>4.09</td>
<td>4.89</td>
<td>-0.8</td>
<td></td>
</tr>
<tr>
<td>i-APAT</td>
<td>37.3±0.7</td>
<td>609.25 (387.24-556.93)</td>
<td>3.33</td>
<td>3.36</td>
<td>-0.03</td>
<td></td>
</tr>
<tr>
<td>d-APAT</td>
<td>16.0±1.2</td>
<td>328.80 (252.17-432.13)</td>
<td>3.82</td>
<td>3.5</td>
<td>0.32</td>
<td></td>
</tr>
<tr>
<td>BPAT</td>
<td>22.7±0.6</td>
<td>185.65 (127.84-286.26)</td>
<td>3.73</td>
<td>4.44</td>
<td>-0.71</td>
<td></td>
</tr>
<tr>
<td>AIO</td>
<td>38.6±1.1</td>
<td>71.24 (44.89-102.85)</td>
<td>3.88</td>
<td>4.24</td>
<td>-0.36</td>
<td></td>
</tr>
</tbody>
</table>

\[ \text{Compound no.} \times 100 \] \( \mu \text{M} \)

\[ \text{AAT} \]

\[ \text{H} \]

\[ \text{2-Cl} \]

\[ \text{2-F} \]

\[ \text{3-CF}_3 \]

\[ \text{4-MeO} \]

\[ \text{2,4-Cl}_2 \]

\[ \text{2,4-Cl}_4 \]

\[ \text{2,5-Cl}_2 \]

\[ \text{2,6-(MeO)_2} \]

\[ \text{3-CI,4-F} \]

\[ \text{3,4-F}_2 \]

\[ \text{i-APAT} \]

\[ \text{d-APAT} \]

\[ \text{BPAT} \]

\[ \text{AIO} \]

\[ \text{4-I} \]

\[ \text{2-Me,4-Br} \]

\[ \text{2,4-(MeO)_2} \]

\[ \text{2,6-(Pr)_2} \]

\[ \text{2,3,4,5-CI}_4 \]

\[ \text{DCAO} \]

\[ \text{SIP} \]

\[ \text{2-Me,4-CI} \]

\[ \text{2,4-Me}_2 \]

\[ \text{2-Me,5-CI} \]

\[ \text{2,6-(Pr)_2} \]

\[ \text{See the text for experimental details. The adenylate-cyclase activity of } P. \text{ americana} \]

(\[ \text{a} \]) \( \text{was measured according to Nathanson's procedure and the cAMP levels} \]

(\[ \text{b} \]) \( \text{were measured by a RIA. Basal (control) and OA (100 (\( \mu \)M)-stimulated} \]

(\[ \text{c} \]) \( \text{Vmax adenylate cyclase activities were 116.2±5.4 and 1,473.3±59.3 pmol cAMP/min/} \]

mg of protein, respectively. In parentheses, 95% confidence limits are shown.} \( \text{Ka} \)

(\[ \text{d} \]) \( \text{and 95% confidence limit values were calculated with the Macintosh} \]

(\[ \text{e} \]) \( \text{personal computer system using a sigmoidal curve-fitting program designed for log dose-probit} \]

(\[ \text{f} \]) \( \text{activity analyses.} \)

(\[ \text{g} \]) \( \text{Calculated by Equation (1).} \)

(\[ \text{h} \]) \( \text{In case predicted activity is overestimated, deviation is obtained by calculating predicted} \]

(\[ \text{i} \]) \( \text{activity subtracted by experimental value and indicated by minus. In case predicted} \]

(\[ \text{j} \]) \( \text{activity is underestimated, deviation is obtained by calculating experimental activity} \]

(\[ \text{k} \]) \( \text{value.} \)

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tries to capture essential information about the receptor, while the latter only captures information about the commonality of compounds that bind. Receptor surface models tend to be geometrically overconstrained (and topologically neutral) since, in the absence of steric variation in a region, they assume the tightest steric surface that fits all training compounds. Receptor surface models do not contain atoms, but try to directly represent the essential features of an active site by assuming complementarity between the shape and properties of the receptor site and the set of binding compounds. The receptor surface model application uses 3D surfaces that define the shape of the receptor site by enclosing the most active members (after appropriate alignment) of a series of compounds. The global minimum of the most active compound 21 in the study compounds (based on the value in the activity column) was made as the active conformer. It really is just one of possibly many self-consistent models that fit the biological-activity data.

Octopamine is not likely to penetrate either the cuticle or the central nervous system of insects effectively, since it is fully ionized at physiological pH. Derivatization of the polar groups would be one possible solution to this problem in trying to develop potential pest-control agents. Generally speaking, antagonists are hydrophobic and they are not suitable for insecticides, because they can not penetrate through insect cuticle. Agonists also show higher binding affinity for the receptor (pKi: 9.54-3.92, Hirashima et al., 1999c) than antagonists (pKi: 8.99-4.09, Roeder, 1990). Therefore, it may be more applicable to develop agonists than antagonists as insecticides. Actually chlordimeform (pKi: 6.91, Roeder and Nathanson, 1993), whose active form N-demethylchlordimeform (pKi: 8.74) shows agonist activity (Nathanson and Hunnicutt, 1981), and Amitraz (pKi: 7.67) were developed as acaricides. Although All 21 showed the highest activity both in octopamine agonist (pKa: 6.12) and binding (pKi: 9.54) activities, it showed no acaridal activity, probably because it can not penetrate through insect cuticle, because SPITs 18 and 67, which contain sulfur atom in the molecule, had acaridal activity (Hirashima et al., 1992). Additionally, chlordimeform and 67 stimulated juvenile-hormone esterase followed by increase of ecydysteroids leading to delay of pupation in Tribolium freemani (Hirashima et al., 1998c). Furthermore, the octopamine agonists reported here inhibited calling behavior (Hirashima et al., 2003a) and in vitro pheromone synthesis of Plodia interpunctella (Hirashima et al., 2003b), whose action may be via a tyramine receptor (unpublished data), suggesting possible cross-reactivity of those compounds with tyramine receptor.

There are several receptors that can be activated by octopamine in the central nervous system that can either activate or inhibit adenyl cyclase activity (Brodly and Cravchik, 2000). In addition, at least one of these receptors may be expressed in up to six alternatively spliced forms that may have different functional properties. Maximal stimulation of nerve cord adenylate-cyclase activity by 18, 21 and 67 was inhibited by several antagonists, including mianserin, cyproheptadine, chlorpromazine and gramine (Hirashima et al., 1992). The rank-order ability of these antagonists to block the maximal adenylate cyclase activation by 18, 21 and 67 was identical to the rank-order ability of the same antagonists to block the enzyme activation by an optimally effective concentration of octopamine. The β-adrenergic antagonist propranolol was less potent in this respect. Thus, the action of these octopamine agonists is same as that of octopamine, due to CamP elevation at OAR2.

The above receptor surface model studies show that full agonists with 2,6-diethyl substituents can be potential ligands to octopamine receptors. Phenyl ring-substitution requirements for partial octopamine agonists differ substantially from each other and other various types of octopamine agonists could be potent, although the type of compounds tested here are too limited to draw any conclusions. These derivatives could provide useful information in the characterization and differentiation of the octopaminergic receptor. They may help to point the way towards developing extremely potent, and relatively specific, octopamine ligands, leading to potential insecticides, although further research on the comparison of the 3D QSAR from agonists is necessary. In order to optimize the activities of these compounds as octopamine ligands, more detailed experiments are in progress.

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