

QSARs on Antibacterial Activity of Some New not yet Synthesized, Substituted Dihydrodibenzothiepins

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This paper presents result of QSAR (Quantitative Structure Activity Relationship) study realized with the PRECLAV software. The dependent property is activity against Pseudomonas aeruginosa IC13202 strain, measured by MIC values. The calibration set includes 35 dihydrodibenzothiepins. The prediction set contains 100 others not yet synthesized dihydrodibenzothiepins, having unknown observed values of activity. There are no outliers in calibration set. The identification of "significant descriptors" and the final results are influenced by "representative sample" character of calibration set within analyzed database (calibration set + prediction set). In absence of prediction set the predictive quality of QSAR is large ($r^2 = 0.9492$, $F = 186.8$, $r^2_{CV} = 0.9340$). Large percentage, in weight, of CH_3 , C_6H_5 and SO_2 molecular fragments seems to be favorable to inhibitory activity. The logP descriptor is not predictor. In presence of prediction set the predictive quality of QSAR is lower ($r^2 = 0.8708$, $F = 48.9$, $r^2_{CV} = 0.8084$). Within the "most active ten" prediction set dihydrodibenzothiepins nine dihydrodibenzothiepins include halogen atoms and seven dihydrodibenzothiepins include SO_2 group.

Keywords: QSAR, dibenzothiepins, Pseudomonas aeruginosa, PRECLAV

Generally speaking, the prediction set includes molecules having unknown observed values of dependent property. The QSAR studies can be made in absence or in presence of certain prediction set. In the absence of the prediction set, the purpose of QSAR studies is the identification of the molecular features with the highest impact (favorable or unfavorable) on the biochemical activity. In the presence of the prediction set the purpose is to identify the prediction set molecules having the largest computed activity.

Dihydrodibenzothiepine derivatives have been studied extensively and found to have various chemical reactivity and broad spectrum of biological activities including antidepressant, antihistaminic, antipsychotic, anti-inflammatory and antimicrobial. The search for new antimicrobial active compounds represents one of the most important directions of current medicinal chemistry. Therefore, we focused our studies on synthesis and antimicrobial activity evaluation of some new dihydrodibenzothiepine derivatives.

The structures of the prediction set molecules were selected mainly by their possibility to be synthesized in our laboratory conditions and taking into account the commercial availability of the raw materials. Also, we chose preferentially halogen substituted radicals, being known their favorable effect on the antimicrobial activity.

The calibration set and the prediction set

In our QSAR study the dependent property was the antibacterial activity against *Pseudomonas aeruginosa* IC13202 bacteria strain, usually defined:

$$A = \log(k/MIC) \quad (1)$$

where:

MIC is the minimum inhibitory concentration ($\mu\text{g/mL}$)

k - real number (factor)

The analyzed chemical structures and the observed values of MIC are presented in literature [1-4]. These values are within the range [62.5, 500]. The value of k factor in formula (1) is $k = 1000$. Therefore, the observed values of activity are within the range [0.301, 1.204].

Figure 1 and table 1 present the chemical structure of the dibenzothiepins in calibration set and the observed value of activities.

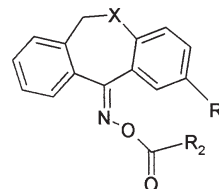


Fig. 1. Structure of the analyzed dibenzothiepins

Figure 1 and table 2 present the chemical structure of the dibenzothiepins in the prediction set.

Methods and formulas

The minimum energy geometry, for each molecule in the calibration and prediction set, was obtained by the PCModel v. 9.0 software [5], using MMX force field [6]. Then the geometry was optimized more rigorously using the quantum mechanics program MOPAC v. 9.103W [7] and included PM6 method [8]. In MOPAC analysis we used the keyword string "pm6 pulay gnorm=0.2 geo-ok bonds vectors".

In the next step, the programs MOPAC, DRAGON v. 5.4 [9] and PRECLAV v. 0907 [10-13, 21, 22] computed, for each molecule, the value for almost 2000 molecular descriptors. The molar refractivities of X, R_1 and R_2 substituents are computed using ChemSketch software [15].

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Table 1
SUBSTITUENTS X, R₁ AND R₂ IN CALIBRATION SET MOLECULES
AND OBSERVED VALUE OF ACTIVITIES

X	R ₁	R ₂	Activity	X	R ₁	R ₂	Activity
SO ₂	H	-CH ₂ -CH ₃	1.204	SO ₂	H	-C ₆ H ₃ - 3,4-di- OCH ₃	1.204
SO ₂	H	-C ₆ H ₄ - <i>p</i> -C ₂ H ₅	0.301	SO ₂	H	thiophen- 2-yl	1.204
SO ₂	H	-C ₆ H ₄ - <i>o</i> -F	0.301	SO ₂	CH ₃	-CH ₂ - CH ₃	0.903
SO ₂	H	-C ₆ H ₄ - <i>m</i> -F	0.301	SO ₂	CH ₃	-C ₆ H ₃ - 2,3-di- OCH ₃	1.204
SO ₂	H	-C ₆ H ₄ - <i>p</i> -F	0.301	SO ₂	CH ₃	-C ₆ H ₃ - 2,6-di- OCH ₃	1.204
SO ₂	H	-C ₆ H ₄ - <i>o</i> -Cl	0.602	S	H	-C ₆ H ₄ - <i>m</i> - CH ₃	0.301
SO ₂	H	-C ₆ H ₄ - <i>m</i> -Cl	0.301	S	H	-C ₆ H ₄ - <i>p</i> - C ₂ H ₅	0.301
SO ₂	H	-C ₆ H ₄ - <i>p</i> -Cl	0.301	S	H	-C ₆ H ₄ - <i>o</i> - F	0.301
SO ₂	H	-C ₆ H ₄ - <i>o</i> -Br	0.301	S	H	-C ₆ H ₄ - <i>p</i> - F	0.301
SO ₂	H	-C ₆ H ₄ - <i>m</i> -Br	0.301	S	H	-C ₆ H ₄ - <i>o</i> - Cl	0.301
SO ₂	H	-C ₆ H ₄ - <i>p</i> -Br	1.204	S	H	-C ₆ H ₄ - <i>p</i> - Cl	0.301
SO ₂	H	-C ₆ H ₄ - <i>o</i> -I	0.301	S	H	-C ₆ H ₄ - <i>o</i> - Br	0.301
SO ₂	H	-C ₆ H ₄ - <i>m</i> -I	0.301	S	H	-C ₆ H ₄ - <i>p</i> -I	0.301
SO ₂	H	-C ₆ H ₄ - <i>p</i> -I	0.301	S	H	-C ₆ H ₄ - <i>o</i> - OCH ₃	0.301
SO ₂	H	-C ₆ H ₄ - <i>o</i> -NO ₂	0.301	S	H	-C ₆ H ₄ - <i>m</i> - OCH ₃	0.301
SO ₂	H	-C ₆ H ₄ - <i>m</i> -NO ₂	0.301	S	H	-C ₆ H ₄ - <i>o</i> - NO ₂	0.301
SO ₂	H	-C ₆ H ₃ -2,3-di-OCH ₃	1.204	S	H	-C ₆ H ₄ - <i>m</i> - NO ₂	0.301
SO ₂	H	-C ₆ H ₃ -2,4-di-OCH ₃	1.204				

Specific procedure [14] identified the "significant" molecular fragments. If the bond order value of chemical bond between two heavy (different from hydrogen) atoms is high, these atoms are included within the same molecular fragment. Consequently, the conjugation of certain molecular fragment with neighbor fragment(s) is low. The percents, in weight, of "significant" fragments are well correlated with the activity values.

PRECLAV program computed QSARs.

Specific procedure identified the "significant" descriptors. In presence of the prediction set PRECLAV program verifies if the calibration set (without "high outliers") is "representative sample" in calibration set + prediction set group, from the point of view of analyzed descriptor [11]. For the same calibration set and different prediction sets PRECLAV computes different QSAR equations. Consequently, in presence and absence of prediction set PRECLAV computes different QSAR equations. From this point of view, PRECLAV is fundamentally different from other QSAR computation programs, which *a priori* assume the calibration set to be a representative sample for the analyzed database.

We obtained tens thousands multilinear type (2) QSAR equations.

$$A = C_0 + \sum_{i=1}^p C_i \cdot D_i \quad (2)$$

where:

- A is activity
- C₀ - intercept
- C_i - weighting factors
- D_i - (values of) "significant" descriptors
- p - number of QSAR descriptors

Specific formula computed the quality of equations [10]. Any descriptor included in the maximum predictive power QSAR is named "predictor". Specific procedure identified the "high outlier" molecules in the calibration set [13] and specific formula computed the relative "utility" U of predictors [10]. The predictors which present a high value (> 500) of "utility" can be considered very useful in estimating the activity, because they correlate very well with activity and do not correlate with other predictors. Each "useful" predictor offers plenty of information about why activity varies from molecule to molecule. Moreover, each "useful" predictor offers a different kind of information from the other predictors.

After computing the A_{calc} values of the activity for the prediction set molecules, PRECLAV sorts these molecules according to the estimated (computed) values. It is computed an average value A_{calc}^m for the estimated values and a standard deviation s of the estimated values. The program considers "high values" the activity values fulfilling the criterion (3) and "low values" the activity values fulfilling the criterion (4).

$$A_{\text{calc}} > A_{\text{calc}}^m + 0.5 \cdot \sigma \quad (3)$$

$$A_{\text{calc}} < A_{\text{calc}}^m - 0.5 \cdot \sigma \quad (4)$$

Therefore, certain prediction set molecule is "high", "moderate" or "low" active relating only to the other molecules in prediction set. In addition, the computed activity of prediction set molecules can be compared with the computed activity of calibration set molecules.

To obtain reliable enough conclusions from computed QSAR we used:

Table 2
SUBSTITUENTS X, R₁ AND R₂ IN PREDICTION SET MOLECULES

MolID	X	R ₁	R ₂	MolID	X	R ₁	R ₂
TH20	S	H	CH ₃	SH20	SO ₂	H	CH ₃
TH40	S	H	<i>n</i> -C ₃ H ₇	SH40	SO ₂	H	<i>n</i> -C ₃ H ₇
TH41	S	H	iso-C ₃ H ₇	SH41	SO ₂	H	iso-C ₃ H ₇
TH50	S	H	<i>n</i> -C ₄ H ₉	SH50	SO ₂	H	<i>n</i> -C ₄ H ₉
TH51	S	H	<i>sec</i> -C ₄ H ₉	SH51	SO ₂	H	<i>sec</i> -C ₄ H ₉
TH52	S	H	iso-C ₄ H ₉	SH52	SO ₂	H	iso-C ₄ H ₉
TH53	S	H	<i>tert</i> -C ₄ H ₉	SH53	SO ₂	H	<i>tert</i> -C ₄ H ₉
TH3N	S	H	CH=CH ₂	SH3N	SO ₂	H	CH=CH ₂
TH2M	S	H	CH ₂ OCH ₃	SH2M	SO ₂	H	CH ₂ OCH ₃
TH2C	S	H	CH ₂ Cl	SH2C	SO ₂	H	CH ₂ Cl
TH2C3	S	H	CCl ₃	SH2C3	SO ₂	H	CCl ₃
TH3F1	S	H	CH(F)CH ₃	SH3F1	SO ₂	H	CH(F)CH ₃
TH3F2	S	H	(CH ₂) ₂ F	SH3F2	SO ₂	H	(CH ₂) ₂ F
TH3C1	S	H	CH(Cl)CH ₃	SH3C1	SO ₂	H	CH(Cl)CH ₃
TH3C2	S	H	(CH ₂) ₂ Cl	SH3C2	SO ₂	H	(CH ₂) ₂ Cl
TH3B1	S	H	CH(Br)CH ₃	SH3B1	SO ₂	H	CH(Br)CH ₃
TH3B2	S	H	(CH ₂) ₂ Br	SH3B2	SO ₂	H	(CH ₂) ₂ Br
TH3M	S	H	(CH ₂) ₂ OCH ₃	SH3M	SO ₂	H	(CH ₂) ₂ OCH ₃
TH5R	S	H	cyclopentyl	SH5R	SO ₂	H	cyclopentyl
TH5MP	S	H	CH ₂ -cyclopentyl	SH5MP	SO ₂	H	CH ₂ -cyclopentyl
TH6R	S	H	cyclohexyl	SH6R	SO ₂	H	cyclohexyl
TH70	S	H	CH ₂ -C ₆ H ₅	SH70	SO ₂	H	CH ₂ -C ₆ H ₅
THH1	S	H	2-furanyl	SHH1	SO ₂	H	2-furanyl
THH2	S	H	3-furanyl	SHH2	SO ₂	H	3-furanyl
THH3	S	H	3-thiophenyl	SHH3	SO ₂	H	3-thiophenyl
TM20	S	CH ₃	CH ₃	SM20	SO ₂	CH ₃	CH ₃
TM40	S	CH ₃	<i>n</i> -C ₃ H ₇	SM40	SO ₂	CH ₃	<i>n</i> -C ₃ H ₇
TM41	S	CH ₃	iso-C ₃ H ₇	SM41	SO ₂	CH ₃	iso-C ₃ H ₇
TM50	S	CH ₃	<i>n</i> -C ₄ H ₉	SM50	SO ₂	CH ₃	<i>n</i> -C ₄ H ₉
TM51	S	CH ₃	<i>sec</i> -C ₄ H ₉	SM51	SO ₂	CH ₃	<i>sec</i> -C ₄ H ₉
TM52	S	CH ₃	iso-C ₄ H ₉	SM52	SO ₂	CH ₃	iso-C ₄ H ₉
TM53	S	CH ₃	<i>tert</i> -C ₄ H ₉	SM53	SO ₂	CH ₃	<i>tert</i> -C ₄ H ₉
TM3N	S	CH ₃	CH=CH ₂	SM3N	SO ₂	CH ₃	CH=CH ₂
TM2M	S	CH ₃	CH ₂ OCH ₃	SM2M	SO ₂	CH ₃	CH ₂ OCH ₃
TM2C	S	CH ₃	CH ₂ Cl	SM2C	SO ₂	CH ₃	CH ₂ Cl
TM2C3	S	CH ₃	CCl ₃	SM2C3	SO ₂	CH ₃	CCl ₃
TM3F1	S	CH ₃	CH(F)CH ₃	SM3F1	SO ₂	CH ₃	CH(F)CH ₃
TM3F2	S	CH ₃	(CH ₂) ₂ F	SM3F2	SO ₂	CH ₃	(CH ₂) ₂ F
TM3C1	S	CH ₃	CH(Cl)CH ₃	SM3C1	SO ₂	CH ₃	CH(Cl)CH ₃
TM3C2	S	CH ₃	(CH ₂) ₂ Cl	SM3C2	SO ₂	CH ₃	(CH ₂) ₂ Cl
TM3B1	S	CH ₃	CH(Br)CH ₃	SM3B1	SO ₂	CH ₃	CH(Br)CH ₃
TM3B2	S	CH ₃	(CH ₂) ₂ Br	SM3B2	SO ₂	CH ₃	(CH ₂) ₂ Br
TM3M	S	CH ₃	(CH ₂) ₂ OCH ₃	SM3M	SO ₂	CH ₃	(CH ₂) ₂ OCH ₃
TM5R	S	CH ₃	cyclopentyl	SM5R	SO ₂	CH ₃	cyclopentyl
TM5MP	S	CH ₃	CH ₂ -cyclopentyl	SM5MP	SO ₂	CH ₃	CH ₂ -cyclopentyl
TM6R	S	CH ₃	cyclohexyl	SM6R	SO ₂	CH ₃	cyclohexyl
TM70	S	CH ₃	CH ₂ -C ₆ H ₅	SM70	SO ₂	CH ₃	CH ₂ -C ₆ H ₅
TMH1	S	CH ₃	2-furanyl	SMH1	SO ₂	CH ₃	2-furanyl
TMH2	S	CH ₃	3-furanyl	SMH2	SO ₂	CH ₃	3-furanyl
TMH3	S	CH ₃	3-thiophenyl	SMH3	SO ₂	CH ₃	3-thiophenyl

- the physical meaning of predictors, according to MOPAC / PRECLAV / DRAGON documentation;
- the algebraic sign of predictors in QSAR;
- the computed value of utility;
- the result of virtual fragmentation.

Results and discussions

In absence of prediction set the number of "significant" descriptors is 442 and we obtained type (2) QSAR equation where:

$$C_0 = 0.6171$$

$$C_1 = -0.3126$$

$$D_1 \text{ is number of CH groups in benzene rings [9] (U = 1000)}$$

$$C_2 = -0.0137$$

$$D_2 \text{ is product of molar refractivities computed for X and R}_1 \text{ substituents (U = 685)}$$

$$C_3 = 4.5435$$

$$D_3 \text{ is lowest eigenvalue \#8 of Burden matrix weighted by atomic Sanderson electronegativities [9, 16] (U = 947)}$$

The quality of above QSAR is large enough ($r^2 = 0.9492$; $F = 186.8$; $r^2_{cy} = 0.9340$). There are no outliers in calibration set. The minimum correlation descriptor/activity is computed for D_3 ($r^2 = 0.2257$). The maximum intercorrelation between descriptors is computed for D_1/D_2 pair ($r^2 = 0.3552$).

The percents, in weight, of some molecular fragments are well correlated (directly or inversely) with the values of activity: C_6H_4 ($r = -0.7468$), CH_3 ($r = 0.6730$), C_6H_3 ($r = 0.5416$) and SO_2 ($r = 0.4662$).

Because of the structure of the computed QSAR and the result of virtual fragmentation we think:

- the presence of unsubstituted benzene rings is unfavorable for the antibacterial activity;
- the presence of substituted benzene rings by alkyl groups is favorable to antibacterial activity;
- there is somehow synergetic effect of X and R₁ groups
- the presence of SO₂ (instead S) groups is favorable for the antibacterial activity;

Table 3
THE "MOST ACTIVE" TEN MOLECULES IN PREDICTION SET

MolID	Estimated activity	MolID	Estimated activity
TM2C3	1.524	SM2C3	1.505
SM3C2	1.489	SM3C1	1.455
SM2M	1.454	TM3C2	1.431
TM3B2	1.427	SM2C	1.410
SM3F2	1.405	SM3B2	1.401

-the influence of molecular shape is large but difficult to interpret;

-there is no influence of lipophilicity on biochemical activity.

In the presence of the prediction set the number of "significant" descriptors is only 44. Because of much smaller number of "significant" descriptors we think the calibration set is not quite "representative sample" in calibration set + prediction set group. We obtained different enough type (2) QSAR equation where:

$$C_0 = -6.6595$$

$$C_1 = 2.3319$$

D_1 is R autocorrelation of lag #5 weighted by atomic polarizabilities [9, 17, 18]

$$C_2 = 0.5023$$

D_2 is 3D-MoRSE signal #15 [9, 19]

$$C_3 = 2.0692$$

D_3 is information content index (neighborhood symmetry of 1-order) [9, 20]

$$C_4 = -0.2698$$

D_4 is topological charge index of order 3 [9]

The quality of above four predictors QSAR is good enough ($r^2 = 0.8708$; $F = 48.9$; $r_{CV}^2 = 0.8084$). Using this equation the maximum activity computed for calibration set molecules is 1.225, the average activity computed for calibration set molecules is 0.511 ± 0.354 and the average activity computed for prediction set molecules is 1.010 ± 0.333 .

According to criterion (3), this equation identified 39 molecules in prediction set having "high values" of activity. The large number of molecules identified as "high active" in prediction set is, probably, the statistical effect of small gap between maximum and minimum observed value of activity in calibration set, only 0.9 logarithmic units. The Table 3 includes the molecules identified as "the most active 10 molecules" in prediction set and the computed values of activity for these molecules. In table 2 prediction set 32% molecules include halogen atoms and 50% molecules include SO_2 chemical group. However, in the table 3 "most active" molecules 90% molecules include halogen atoms and 70% include SO_2 group. Favorable effect of halogen atoms and SO_2 group on activity is obvious.

Conclusions

Large percentage, in weight, of CH_3 , C_6H_5 and SO_2 molecular fragments seems to be favorable for the activity.

Large percentage, in weight, of C_6H_4 molecular fragments seems to be unfavorable for the activity.

The influence of molecular shape is ambiguous and difficult to interpret.

The logP descriptor is not predictor.

The "representative sample" feature of calibration set in calibration set + prediction set group have a large influence on predictive power of computed QSAR.

The most active ten molecules in prediction set include halogen atoms and a SO_2 group.

References

1. STECOZA, C. E., CĂPROIU, M. T., DRĂGHICI, C., ILIE, C., CHIRIȚĂ, I. C., Rev. Chim. (București), **59**, no. 12, 2008, p. 1348
2. STECOZA, C. E., CĂPROIU, M. T., DRĂGHICI, C., CHIFIRIUC, M. C., DRĂCEA, N. O., Rev. Chim. (București), **60**, no. 2, 2009, p. 137
3. ILIE, C., STECOZA, C. E., CĂPROIU, M. T., HĂU, R., GUȚĂ, R., NĂNĂU-ANDRESCU, D., Rev. Chim. (București), **60**, no. 6, 2009, p. 588
4. CHIFIRIUC, M. C., STECOZA, C. E., DRĂCEA, O., LARION, C., ISRAIL, A. M., accepted for publication, Rom. Biotech. Lett., 2009
5. ** The last version of PCModel is available from Serena Software, Box 3076, Bloomington, IN 47402-3076, USA, Internet page <http://www.serenasoft.com/>
6. *** Improved version of MM2, see Allinger N. L., J. Am. Chem. Soc., **99**, 1977, p. 8127
7. *** The last version of MOPAC is available from Internet page <http://www.openmopac.net/>
8. STEWART, J. J. P., J. Mol. Model., **13**, 2007, p. 1173
9. *** The last version of DRAGON (documentation included) is available from Talete srl., via V. Pisani, 13-20124, Milan, Italy, Internet page <http://www.taletemi.it>
10. *** The last version of PRECLAV (documentation included) is available from Center of Organic Chemistry – Bucharest, email: ltarko@cco.ro or pfilip@cco.ro
11. TARKO, L., Rev. Chim. (București), **56**, no. 6, 2005, p. 639
12. TARKO, L., LUPESCU I., GROPOSILA - CONSTANTINESCU D., ARKIVOC, 2005, X, p. 254
13. TARKO, L., STECOZA C. E., ILIE C., CHIFIRIUC M. C. Rev. Chim. (București), **60**, no. 5, 2009, p. 476
14. TARKO, L., Rev. Chim. (București), **55**, no. 7, 2004, p. 539
15. ** The last version of ChemSketch software is available from Advanced Chemistry Development (ACD), 110 Yonge Street, 14th floor, Toronto, Ontario, Canada M5C 1T4, see website <http://www.acdlabs.com/download/chemsketch/>
16. BURDEN, F.R., J. Chem. Inf. Comput. Sci. **29**, 1989, p. 225
17. CONSONNI, V., TODESCHINI, R., PAVAN, M., J. Chem. Inf. Comput. Sci., **42**, 2002, p. 682
18. CONSONNI, V., TODESCHINI, R., PAVAN, M., GRAMATICA, P., J. Chem. Inf. Comput. Sci., **42**, 2002, p. 693
19. SCHUUR, J.H., SELZER, P., GASTEIGER, J., J. Am. Chem. Soc., **36**, 1996, p. 334
20. MAGNUSON, V.R., HARRISS, D.K., BASAK, S.C., Physical and Theoretical Chemistry, R.B. King (Ed.), Elsevier, Amsterdam, **1983**, p. 178
21. TARKO, L., PINTILIE, L., NEGUT, C., ONISCU, C., CAPROIU, M.T., Rev. Chim. (București), **59**, no. 2, 2008, p. 185
22. DONE, R., MANDRILA, G., TARKO, L., Rev. Chim. (București), **60**, no. 10, 2009, p. 992

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