QSAR Study on the Relative Activity of Thiazide diuretics in Doping

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Abstract. Electrotopological state indices ($E_k$) of atom types is used to describe the structures of 10 thiazide diuretics in doping and related to the relative activity ($R_a$). The optimal three-parameter ($E_{16}$, $E_{23}$, $E_9$) quantitative structure-activity relationships (QSAR) model is established by using leaps-and-bounds regression analysis for the relative activity ($R_a$) of thiazide diuretics along with the $E_k$. The correlation coefficients ($R^2$) and the leave-one-out (LOO) cross validation $R_{cv}^2$ for the $R_a$ was 0.975 and 0.936, respectively. The model has favorable correlation, as well as robustness and good prediction capability by $F$, $R_{cv}^2$, $AIC$, $F_{IT}$, $V_{IF}$ tests. The results indicate that the molecular structural units: =O (in sulfonyl), =N– and =C< (in aromatic ring) are main factors which can affect the relative activity ($R_a$) bioactivities of these compounds directly.

1. Introduction

Doping (the stimulant has been used in horse racing since 16th century is called doping.) is the general name of all the athletic banned substance in international sports, including all the banned drugs and methods in sports prescribed by international sports organizations. There are five categories of dopings: irritant, anesthetic, anabolic agents, diuretic and peptide and glycoprotein hormone, according to Prohibited Classes of Substances and Prohibited Methods by International Olympics Committee. Thousands of compounds have been involved, including diuretics, a type of stimulant banned since the 1988 Seoul Olympics[1]. Diuretics can increase the amount of urine excretion, and promote the accumulation of excess water in the subcutaneous and abdominal cavity in a short period of time. As a result, sports competition is unfair, fair and banned. At present, the diuretics in the list of prohibited substances of IOC are divided into four categories: loop diuretics, thiazide diuretics, potassium diuretics, carbonic anhydrase inhibitors[2].

Quantitative structure-activity relationship (QSAR)[3-6] can establish the quantitative relationship between the molecular structure and activity of organic compounds, which shows practical significance for predicting the biological activity of new compounds, initial screening and evaluation of activity compounds, and reducing high cost of the activity evaluation. Topological indexes could directly derive from chemical molecular structure independent of the experimental mensuration. Therefore, the appearance of topological indexes in general QSAR models was popular for the development of reliable QSAR models. The electrotopological state (E-state) indexes ($E_k$) put forward by by Kier and Hall[7-9] are one of the most broadly applied topological indexes.

Based on the electrotopological state (E-state) indexes[7-9], the robust QSAR model of relative activity ($R_a$)[10] for 10 thiazide diuretics in doping was established by leaps-and-bounds regression, to estimate and predict $R_a$ of thiazide diuretics and reveal the microstructure that affected its relative activity at the molecular level. It provides a theoretical reference for the effective detection of thiazide diuretics and plays an important role in anti-doping.
2. Material and methods

2.1 Studied compounds and their biological activity data

The compounds studied in this paper are 10 thiazide diuretics[^10]. The parent structure is shown in Fig.1, where R1, R2, R3, X, Y are the substituent groups. Their substituents and their relative activity data (Ra)[^10] are shown in Table 1.

![Figure1. Basic structure of thiazide diuretics](image)

Table 1. The molecule structures and Ra of thiazide diuretics

| No. | R1 | R2         | R3 | X   | Y   | Ra[^10] |
|-----|----|------------|----|-----|-----|---------|
| 1   | Cl | H          | H  | –N= | >C= | 0.80    |
| 2   | Cl | SCH2C6H5   | H  | –N= | >C= | 1.30    |
| 3   | Cl | H          | H  | –NH–| >CH–| 1.40    |
| 4   | Cl | CHCl2      | H  | –NH–| >CH–| 1.70    |
| 5   | Cl | CHCl2      | CH3| –NH–| >CH–| 1.80    |
| 6   | Cl | CH2SCH2CF3 | CH3| –NH–| >CH–| 2.00    |
| 7   | CF3| CH2Cl      | H  | –NH–| >CH–| 1.30    |
| 8   | CF3| CH2C6H5    | H  | –NH–| >CH–| 1.80    |
| 9   | Cl | CH2Cl      | CH3| –NH–| >CH–| 1.80    |
| 10  | CF3| H          | H  | –NH–| >CH–| 1.30    |

2.2 Molecular descriptors

In this paper, electrotopological state indices (Ek[^7-9]) is used to characterize the molecular structure of different classes of organic compounds. Hall et al. have investigated the localization of several famous topologies, and provided Ek which can show the topological, geometrical and electrical characters comprehensively. The Ek of all atoms in the compound molecular skeleton was calculated following the method given by Hall et al, and the details can be found in references[^7-9].

2.3 Statistical regression analysis

For QSAR derivation, Ra values[^10] act as the dependent variables, and electrotopological state indices descriptors act as the independent variables. The regression analyses are carried out by using multiple linear regression (MLR), leaps-and-bounds regression (LBR) program. By using Fischer statistics (F-tests) to eliminate insignificant descriptor when entering a new descriptor, stepwise regression can identify the most important descriptors contributing to the relative activity (Ra) of thiazide diuretics.

The correlation between variables in model was estimated by the variance inflation factor (VIF[^11]) value. VIF is defined as follows:

$$VIF = 1/(1 - \beta^2)$$ (1)

in which $\beta$ is the correlation coefficient of multiple regressions between one variable and the others in the equation. $VIF = 1.0$ suggests no self-correlation among each variable; if $VIF$ ranges from 1.0 to 5.0, indicating that there is no obvious autocorrelation between variables, the model is stable; when $VIF$ is larger than 5.0, the regression equation is unstable and recheck of variables correlation coefficient is necessary.

The quality of each regression model was evaluated, using a squared correlation coefficient ($R^2$), cross validation squared correlation coefficient ($R_{cv}^2$). The “leave-one-out” (LOO) cross-validation...
coefficient $R_{cv}^2$ was considered as an indicator of the predictive performance and stability of a QSAR model. As a rule of thumb, the equations with regression coefficients $R^2 > 0.80^{[12]}$ and $R_{cv}^2 > 0.50^{[13]}$ are considered reasonable. Where parameter $R_{cv}^2 > 0.50$ is used as a criterion of both robustness and predictive ability of the model, whereas the conventional correlation coefficient $R^2$ defines the goodness-of-fit.

Another indicator used to evaluate the quality of model is the standard deviation ($S_D$). The model is good and the prediction accuracy is acceptable when the ratio of the standard deviation to the value range (the samples between the maximum and minimum values) is less than 10%$^{[14]}$.

Akaike’s information criterion ($AIC$; Eq.2; the model that produces the minimum AIC value is considered potentially the most useful) and Kubinyi function ($F_{IT}$; Eq.3; the best model will present the highest value of this function)$^{[15,16]}$ are applied to determine if a variable should be included in the model. That is to say, if the Akaike’s information criterion decreases in value when adding an additional variable and the Kubinyi function increases in value, then, the introduction of this new variable is justified.

$$AIC = RSS \times \frac{f + b}{(f - b)^2}$$  \hspace{1cm} (2)

$$F_{IT} = \frac{R^2(f - b - 1)}{(f + b^2)(1 - R^2)}$$  \hspace{1cm} (3)

Where $RSS$ is the residual sum of squares, $f$ is the number of compounds included in the model, $b$ is the number of variables included in the model, $R^2$ is the square of the correlation coefficient.

### 3. Results and discussion

#### 3.1 QSAR equation of the Ra

The electrotopological state indices and relative activity($Ra$) of thiazide diuretics were input into MINITAB14.0 statistical analysis software, and the leaps-and-bounds regression was used to select the best variable combinations, to establish the best QSAR models, the results are shown in Table 2.

| No. | $R^2$   | $R_{cv}^2$ | $AIC$  | $FIT$  | $S_D$  | $F$  | Variables |
|-----|---------|------------|-------|--------|--------|------|-----------|
| 1   | 0.558   | 0.264      | 0.118 | 0.918  | 0.255  | 10.119| $E_16$    |
| 2   | 0.914   | 0.872      | 0.028 | 5.314  | 0.120  | 37.423| $E_16, E_23$ |
| 3   | 0.975   | $0.936$    | 0.020 | 12.316 | 0.070  | 77.254| $E_16, E_23, E_9$ |
| 4   | 0.979   | 0.847      | 0.070 | 8.965  | 0.070  | 59.598| $E_16, E_23, E_9, E_2$ |

Table 2 shows that with the increase in the number of independent variables in the model, in addition to $R^2$, $S_D$, the rest of the statistical indicators in the ternary equation has a turning point, which $R_{cv}^2$, $F_{IT}$, $F$ has the maximum, $AIC$ has a minimum. So choose the best three-variable QSAR model:

$$Ra = -0.929(\pm 1.154) + 0.070(\pm 0.018)E_9 + 0.230(\pm 0.024)E_{16} - 0.157(\pm 0.016)E_{23}$$  \hspace{1cm} (4)

$$f = 10, R^2 = 0.975, R_{cv}^2 = 0.936, R_{adj}^2 = 0.962, F = 77.254, S_D = 0.070$$

The QSAR equation (4) can be used to predict $Ra$, and the predicted value($Ra_{cal}$) is basically consistent with the experimental data in Table 3.

| No. | $E_2$ | $E_9$ | $E_{16}$ | $Ra_{exp}$ | $Ra_{cal}$ | $Ra_{err}$. |
|-----|-------|-------|----------|------------|------------|-----------|
| 1   | 0.0000 | -0.8650 | 45.545 | 3.7373 | 0.80 | 0.79 | 0.01 |
| 2   | 0.5144 | 0.1280 | 47.827 | 4.2093 | 1.30 | 1.31 | -0.01 |
| 3   | 0.0143 | -0.5125 | 45.684 | 0.0000 | 1.40 | 1.43 | -0.03 |
| 4   | 0.0000 | -0.9296 | 46.740 | 0.0000 | 1.70 | 1.65 | 0.05 |
| 5   | 0.0000 | -0.8591 | 47.652 | 0.0000 | 1.80 | 1.86 | -0.06 |
3.2 Validation of the QSAR equation

The predictive capability of a QSAR model should be tested through model validation. Cross-validation is one of the most often used model validation methods. The $R_{cv}^2$ values of models(4) is 0.936, which is well above 0.5, indicating that the model has good robustness and prediction ability.

The $V_{IF}$ values of the variables($E_9$, $E_{16}$, $E_{23}$) in model (4) are 1.319, 1.069, 1.291, respectively, which is less than 5.0. Indicating that model is statistically significant and have good stability.

The relative activities($R_a$) range of the sample between the maximum and minimum values is 1.20(2.00 – 0.80 = 1.20). The ratio of the standard deviation ($S_D$=0.070) to 1.20 is 5.83%. This is less than 10%, indicating that the model has acceptable predictive accuracy.

3.3 Analysis of the QSAR equation(4)

According to the theory of electrotopological state indices, it can be seen that $E_9$ in the model (4) reflects carbon atom(=C<) in aromatic ring, $E_{16}$ in the model (4) reflects oxygen atom(=O) in sulfuryl, $E_{23}$ in the model (4) reflects nitrogen atom(=N–). Wherein =O and =N– are highly electronegative polar groups that are capable of forming hydrogen bonds. =C< is nonpolar or weak polar groups that is hydrophobic interaction of forming. The diuretic contains =C<, =O and =N–, which is hydrogen bond and hydrophobic with the enzyme in the human body.

In addition, determination coefficient $R^2$ is also called the reduction error ratio. $R^2 = 0.975$ in model (4), indicating that $E_{16}$, $E_{23}$ and $E_9$ constant items together show 97.5% of the factors affecting relative activities($R_a$) of the diuretic, and only 2.5% are random factors, which further prove the correctness of the model.

4. Conclusions

(1) The optimal three-variable QSAR model of relative activity for thiazide diuretics was constructed by using the leaps-and-bounds regression method. The QSAR models have showed good correlation, as well as robustness and prediction ability by statistical indicators:$R^2$, $R_{cv}^2$, $S_D$, $V_{IF}$, $F_{IT}$ and $A_{IC}$ tests.

(2) According to the electrotopological state indices entering the model (4), the main molecular structural units that affect their relative activity are =O(in sulfuryl), =N– and =C<(in aromatic ring).

(3) According to the main molecular structure units that affect relative activity, the interactions between thiazide diuretics molecules and enzyme in the human body are mainly hydrophobic, hydrogen bond.

All in all, this study provides theoretical guidance for diuretic detection, and also has certain guiding effect on anti-doping.

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