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By Takahiro Goto

Purpose

The purpose of the project is to investigate the utility of data collected at the National Institute of Health Science in Japan for quantitative structure-activity relationship (QSAR) modeling. The hypothesis being tested is that a linear relationship exists between a chemical compound's structural and physicochemical characteristics and its no observed effect level (NOEL).

Methods

The data that was used included a set of intermediate–duration oral NOELs that are publicly accessible on the website of the National Institute of Health Science in Japan (<u>http://www.nihs.go.jp/index-j.html</u>). The software, Leadscope Predictive Data Miner, was used to build an associative model and to analyze the correlation between observed NOELs and calculated NOELs.

Results

The results showed that Leadscope PDM selected 117 structural characteristics (e.g., sulfonate, 1-hydoroxynaphthalen, 3-hydroxy-1-benzensulfonate, etc.) and 7 physicochemical characteristics (e.g., hydrogen bond acceptors, polar surface area, ALogP, etc.) for 218 compounds that were entered into the software. An associative model with one latent descriptor, called a PLS factor, was then developed. The Leadscope PDM program provided visualization of the correlation between observed NOELs and calculated NOELs. The R² was 0.35, and cross-validated R² was 0.21.

Conclusions

Overall, this analysis provided little evidence for utility of the data. This was because the hypothesis could not be proven with the available data and model. This project did demonstrate a good example of QSAR modeling using NOELs and Leadscope software as evidenced by the results of this study. It is anticipated that this project will facilitate further QSAR studies in the near future utilizing the data already generated.

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1. INTRODUCTION

Quantitative structure-activity relationship (QSAR) is a regression of the biological activity of chemical compounds on their structural and physicochemical characteristics. Although the predictive value of the outcome is not always accurate, QSAR is a powerful method because it investigates the biological activities of large numbers of chemical compounds quickly and cost-effectively in the absence of experimental toxicological data. There are many examples of QSAR applications in public health when this method demonstrates a desired forecasting effect [1-3]. QSAR is a cross-chemical extrapolation at a level higher than Benchmark dose (BMD) modeling, chemical-specific adjustment factor (CSAF) modeling, or physiologically based pharmacokinetic (PBPK) modeling. QSAR relies on much smaller amounts of initial information than required by the other methods and, as such, it can be applied to a much larger number of substances needing health guidance values (HGVs) than BMD, CSAF, or PBPK methods can do [4].

Accuracy of toxicological QSARs has been investigated by a number of authors. For example, Rupp et al. have reported on the comparison between experimentally observed no observed adverse effect levels (NOAELs) in repetitive dose animal studies and predicted lowest observed adverse effect levels (LOAELs) obtained using TOPKAT 6.2 version 3.1 [5]. Venkatapathy et al. reported a correlation between observed LOAELs and calculated LOAELs using TOPKAT [6] and a correlation between observed tumor dose $(TD_{50} - , the dose that induces tumors in half of the test animals at the end of a$ standard life span) and calculated the lethal dose (LD₅₀ - , the dose that kills 50% of a Mazzatorta et al. compared oral rat chronic LOAELs with study population [7]. calculated LOAELs that were obtained with an in-house developed procedure that relied on partial least squares (PLS) regression. A genetic-algorithm variable selection was used as implemented in Statistics Toolbox 5.0.1 of MATLAB 7.0.1 [8]. Mombelli analyzed the predictive performance of irritant chemical compounds in three kinds of commercially available software (DEREK, HAZARDEXPERT, and TOPKAT). His conclusion was that only TOPKAT was predictive with regard to the chemicals studied [9].

When the Chemical Substances Control Law (CSCL) took effect in Japan in 1972, approximately 20,000 chemicals had already been in use, manufactured, or imported by Japanese industry in large amounts. These chemicals are called the "Existing Chemicals," with the Japanese government being tasked by law to collect toxicological information about these substances. The data for these studied have been obtained from animal studies. They include single dose oral toxicity study, 28-day repeated oral dose toxicity study, repeated dose and reproductive/developmental toxicity study, simple oral administration reproductive toxicity study, one-generation reproduction toxicity, and 90-day repeated dose toxicity study. Chemical compounds that emerged in Japan after the law took effect in 1972 are labeled "New Chemicals," and manufacturers and importers are required to submit data to the Japanese government about the safety of these New Chemicals. These data include toxicology, degradability, bioaccumulation, and ecological effects. The Japanese government is responsible for safety information about Existing Chemicals. Presently, only a small percent of all target Existing Chemical compounds have appropriate toxicological data due to the cost and time-consuming nature of animal experiments.

Although QSAR has not been able to completely replace the role of animal experiments (although the EU REACH legislation anticipates this), it is useful to estimate a compound's toxicity using QSAR because little time and input of information is required to provide results that are comparable with results of animal experiments. One of the useful ways to utilize QSAR is to compare the observed results in animal experiments with calculated results that QSAR generates. If the results of QSAR studies and animal studies are comparable, the outcome of one confirms the findings of the other. Additionally, QSAR can be applicable to some compounds that do not need testing because of their low toxicity.

The hypothesis being tested is that a linear relationship exists between a chemical compound's structural and physicochemical characteristics and its no observed effect level (NOEL). The relationship between experimental and calculated NOELs is believed to be linear.

The aim of the current project is to investigate the utility of data collected at the National Institute of Health Science (NIHS) in Japan for QSAR modeling. To achieve this aim, an associative QSAR model was developed based on a set of intermediate-duration oral NOEL using the commercially available software, Leadscope Predictive Data Miner (PDM) (Leadscope Inc., Columbus, USA). Only 21 applicable publications were found using the term "Leadscope" in a search of PubMed (US National Library of Medicine, National Institutes of Health as of February 16th, 2013). The 21 publications included ones written by Arvidson [10], Cross et al. [11], Roberts et al. [12], Blower et al. [13-16], and Yang et al. [17-18]. This project, using PDM and NOELs, is expected to bring a new aspect to QSAR modeling by providing a workable model.

2. METHODS

(1) Data Collection and Processing

The data used are a set of intermediate-duration oral NOELs that are publicly accessible on the website of the NIHS in Japan [19]. The Institute's database contains the 340 Existing Chemicals and each chemical compound was researched in a number of studies. The intermediate-duration oral NOELs studies were selected because they had the most chemical compounds and data. The availability and quality of intermediate-duration NOELs favored the use of the database. These NOELs were obtained for a: 1) 28-day repeated oral dose toxicity study for 122 compounds, 2) repeated dose and reproductive/developmental toxicity study for 85 compounds, 3) one-generation reproduction toxicity study for 1 compound, and 4) simple oral administration reproductive toxicity study for 10 compounds. The data were mainly obtained in 28-day repetitive-dose oral animal experiments in rats performed in Japan. The website also had summary pages as well as the original reports of the studies. When a difference was observed between the NOEL information in the summary pages and in the original reports, the original reports were used assuming greater accuracy. A total of two hundred twenty nine (229) chemicals with acceptable NOEL data were reported on the Institute's website and utilized for the study. These chemicals required adjustment before they could be entered in the Leadscope program. All compounds having inorganic atoms, such as P (phosphorus), Si (silicon), or Sn (tin) were eliminated since QSAR methodology is unable to analyze data containing these compounds. The inorganic part of salts was also removed and the compounds were converted to their corresponding hydrolyzed organic forms. For example, Sodium *p*-styrenesulfonate (No. 85 in Appendix I) (a parent compound) was converted into *p*-styrensulfonic acid (a child compound). Also, the hydrates, such as Disodiumsuccinatehexahydrate (No. 128) was converted into the unhydrates form (succinic acid).

Following these adjustments, 218 NOELs remained and were used in the study. Among them, the average duration of the studies was 36.5 days with experiments for the 217 compounds being performed at 11 different institutions within Japan and one compound tested at an institution in the Netherlands.

The NOELs varied from 0.0684 to 11.89 mmol/kg/day (-2.17 to 1.08 mmol/kg/day on the log scale), and they were coded in PDM by their ID numbers. The chemical compounds' structures were converted to Simplified Molecular Input Line Entry Specification (SMILES) notation using the Online SMILES Translator and Structure File

Generator [20]. Then, a structure-data file (SDF), which has two-dimensional information about atoms, bonds, connectivity, and coordinates of molecule on each compound, was created from SMILES using the same website, and was entered in PDM along with NOELs. Also, structurally transformed compounds were entered into PDM with the child's (instead of the parent's compound) structural information. Finally, PDM had 218 compounds' ID with corresponding NOELs and structural information. General information about these compounds is shown in Appendix I.

(2) Analysis plan

Among a number of commercially available software in the area of QSAR, PDM was selected because of its ability to build an associative model. When the data about chemical compounds are entered into PDM, the software generates two kinds of their inherent information; the chemical compounds' structural characteristics (features) and physicochemical characteristics (descriptors). PDM recognizes what functional groups each chemical compound has from its SMILES. The software also automatically calculates each chemical compound's physicochemical characteristics. The structural characteristics and physicochemical characteristics of the compounds are shown below (Table 2-1 and Table 2-2). The software provided an automatic procedure for selection of the characteristic used to build the study model.

Table 2-1: The Examples of Chemical Compounds' Structural Characteristics

Hydroxyl	1, 2-Diol / 1, 3-Diol
Amine	Carboxyl
Carboxylate	Sulfonate
1-(tert)Butylbenzene	1-Benzenesulfonate
1-Hydroxynaphthalene	1-Carbonylbenzene

 Table 2-2: Chemical compounds' Physicochemical Characteristics

Physicochemical	Definitions ^a
Characteristics	
Molecular Weight	The sum of the atomic masses of all the atoms in the molecule
Parent Molecular	The sum of the atomic masses of all the atoms in the parent
Weight	molecule
Parent Atom	The number of all the atoms in the parent molecule
Count	
Rotatable Bonds	The number of single, non-terminal acyclic bonds
Hydrogen Bond	a. Any doubly bonded oxygen
Acceptors (HBA)	b. Any singly bonded oxygen, such as anion A-O- or hydroxyl
	A-OH
	c. Uncharged imine, nitrile, or aromatic N. Examples of
	imines include C=NH, or C=N-Ak; aromatic N includes ARO
	-N-ARO, where both ARO-N bonds are cyclic, aromatic
	d. An ether oxygen in the form C-O-C, where
	• neither C is substituted by a doubly-bonded N, S, or O;
	e.g., neither C is part of a carbonyl
	• at most one C is aromatic
	• the O is acyclic
	e. C-O-C is cyclic, both C are sp^3 hybridized

	f. Any doubly bonded sulfur in thioxomethyl C=S, where S has no other attachments
Hydrogen Bond Donors (HBD)	 a. Any OH not part of an oxo acid (A hydroxyl of an oxo acid is considered ionized at physiological pH and will be recognized as an HBA.) b. Any NH not in a tetrazole or N-trifluoromethyl-sulfonamide
Polar surface area	Molecular polar surface area (PSA) i.e., surface attributed to polar atoms, is a descriptor that has been shown to correlate well with passive molecular transport through membranes and, therefore, allows prediction of transport of drugs.
ALogP	The octanol-water partition coefficient (log P) is the commonly used measure of lipophilicity.
Lipinski Score	 Lipinski's <i>Rule of Five</i> is a simple way to assess a compound's oral bioavailability based on upper limits of the compound's logP, Molecular Weight, and the Number of Hydrogen Bond Donors and Acceptors. This score has a value between 0 – 4 and indicates the number of the following rules that are violated: a. if ALogP is greater than or equal to 5.0 – add 1; b. if Parent Molecular Weight is greater than or equal to 500 – add 1; c. if the sum of N+O atoms is greater than 10 – add 1; d. if the sum of NH + OH is greater than 5 – add 1

^a from Leadscope User Manual [21]

Using a chemical compound's structural and physicochemical characteristics and observed NOEL, PDM generated associative models by PLS regression, and also provided a calculated NOEL derived from the model. The software also provided a correlation plot between the observed NOEL and calculated NOEL with indicators, such as R^2 and cross-validated R^2 (Q^2), which gives information about how well the calculated NOEL correlated with the observed NOEL. R^2 is the coefficient of determination, which indicates how well the deviation of the data can be explained by the equation. Q^2 is a parameter suggesting robustness of the model to perturbations in the data.

3. RESULTS

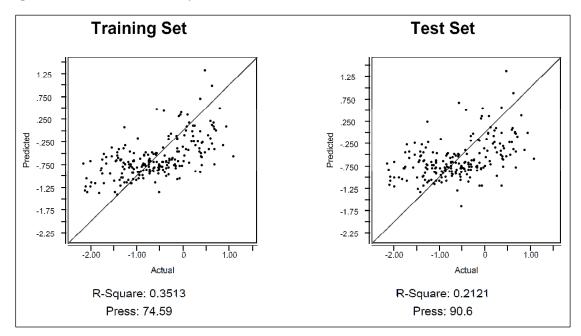
The results of the modeling in PDM showed that 218 compounds were used in the modeling as entered into PDM, and the average number of local neighbors (the compounds having similar structural characteristics) was 2.11 and the number of singletons (the compounds having no similar structural characteristics in the group of 218 compounds) was 131. Also, 117 structural characteristics (e.g., sulfonate, 1-hydoroxynaphthalen, 3-hydroxy-1-benzensulfonate, etc.) and 7 physicochemical characteristics (e.g., hydrogen bond acceptors, polar surface area, ALogP, etc.) were chosen in PDM to optimize building an associative model, and then a model (218Model) was built with one latent variable called the PLS factor (Table3-1). In the process, PDM created the PLS factor from 124 features and descriptors to avoid multicolinearity. This is because features are correlated with each other in many cases and therefore the factor analysis is used in the modeling process to deal with many like features and descriptors. Also, 2% test set (50-fold cross validation) was chosen in cross validation because of the general use for the sample size of 218 compounds. In 50-fold cross validation, the entire set of compounds is randomly divided into 50 portions. Forty-nine (49) portions are first used to build the model, while remaining one portion is used to test the model created with 49 portions. This process is repeated 50 times, using a different portion as

the test set. Also, PDM showed the correlation between observed NOELs and calculated NOELs, where observed NOELs are placed in X axis and calculated NOELs are placed in Y axis. R^2 was 0.35, and Q^2 was 0.21 (Figure 3-1). The least predicted residual error sum of square (PRESS) is 74.59 for training set and 90.6 for the test set.

Table 3-1. Modeling method summary

Method used:	Partial Le	east Squares Regression
Final predictors used:	Total:	124
Structural features:		hierarchy features(116), dynamic features(1)
Calculated descriptors:		Leadscope calculated properties(7)
Number of PLS factors:	1	
Cross-validation:	2% test s	set

Figure 3-1. Model summary



Note: R-Square in Training Set (0.3513) is R^2 and R-Square in Test Set (0.2121) is Q^2 .

Also, out of 124, the 10 best features and descriptors that contributed to the model more than others are shown in Table 3-2 (e.g., hydrogen bond acceptors and polar surface area). Because features that have smaller residuals (shown as % Feature Residuals) are a better contributor to the model, the features are in the order of smaller residuals. Loadings are the correlation coefficients between the features and the PLS factor. Features that have larger absolute total loadings and larger total weight are generally better contributors.

Name	Total Loadings	Total (+) Loadings	Total (-) Loadings	Total Weight	% Feature Residuals
Hydrogen Bond Acceptors	0.285	0.285	0.0	0.259	50.9
Polar Surface Area	0.27	0.27	0.0	0.229	56.0
sulfonate	0.253	0.253	0.0	0.219	61.4
1-benzene-sulfonate	0.249	0.249	0.0	0.239	62.7
sulfonyl group	0.228	0.228	0.0	0.155	68.6
benzene, 1-sulfonyl-	0.228	0.228	0.0	0.168	68.6
2-naphthalene-sulfonate	0.18	0.18	0.0	0.113	80.5
naphthalene, 1-hydroxy-	0.173	0.173	0.0	0.169	82.1
1-benzene-sulfonate, 3-hydroxy-	0.16	0.16	0.0	0.128	84.5
benzene, 1-(alkyl, acyc)-	-0.152	0.0	-0.152	-0.101	86.0

 Table 3-2. Top10 most significant features and descriptors

The residuals in the 218Model suggested that there might be a weak positive correlation between NOELs and residuals (Figure 3-2). The histogram of the frequency of the residuals demonstrated that the residuals were reasonably normally distributed (Figure

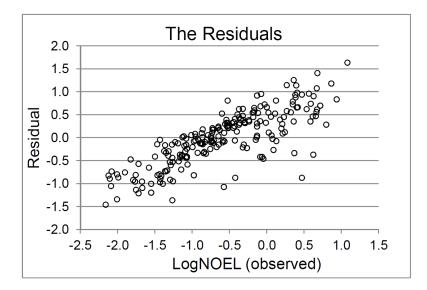
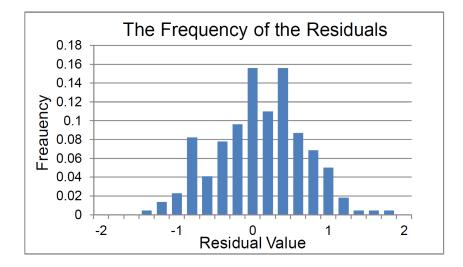


Figure 3-2. The Residuals in the 218Model

Figure 3-3. The Frequency of the Residuals in the 218Model



Next, the average of observed NOELs was compared with calculated NOELs from each

institution. Out of all 12 institutions, 8 institutions conducted studies on more than one compound. Among these 8 institutions, the average in each institution turned out to be reasonably close to each other, and the relative difference did not exceed 33% (Table 3-3).

Table 3-3. The average of observed NOELs and that of calculated NOELs in 8

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No.	Institution	The	The	Absolute	Relative	The
		average of	average of	error	error, %	number
		observed	calculated			of
		log-NOELs	log-NOELs			studies
1	Safety Research Institute for	-0.509	-0.582	0.073	14	28
	Chemical Compounds Co.,					
	Ltd.					
2	Research Institute for Animal	-0.789	-0.612	0.177	22	34
	Science in Biochemistry and					
	Toxicology					
3	Food and Drug Safety	-0.471	-0.513	0.042	9	41
	Center, Hatano Research					
	Institute, Japan					
4	Mitsubishi Chemical	-0.703	-0.680	0.023	3	33
	Medience Corporation					
5	Public Interest Incorporated	-0.476	-0.632	0.156	33	28
	Foundation BioSafety					
	Research Center					
6	Bozo Research Center	-0.604	-0.628	0.024	4	21

7	Panapharm Laboratories Co.,	-0.744	-0.558	0.186	25	13
	Ltd.					
8	Nihon Bioresearch Inc.	-0.550	-0.624	0.074	13	16

Finally, a number of chemical compounds were placed in groups to improve the model. By dividing 218 compounds into two or more groups, it was believed that a better associative model could be developed. In order to achieve this, a model with a high Q^2 The 218Model was created with one compound that had the largest was created. residual in the results being removed. Then another associative model was created with the remaining 217 compounds and again removing one compound with the largest This procedure was repeated until the model achieved a 0.9 or larger Q^2 . residual. When 104 compounds were removed, a Q^2 of the model was 0.9029 with 114 compounds left (114Model) (Figure 3-4). Also, all compounds that were included in the 114Model are shown in Appendix I. In the construction of the 114Model, 115 structural characteristics and 8 physicochemical characteristic were used and the model was created with 10 PLS factors.

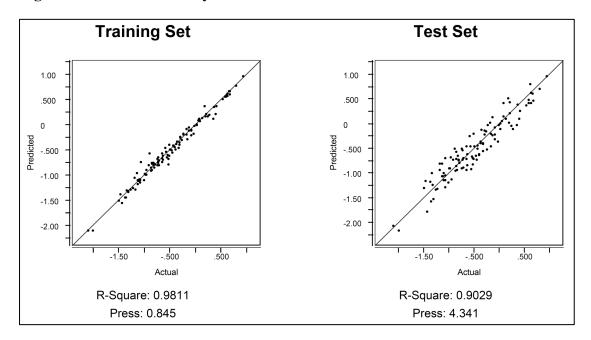


Figure 3-4. Model summary in the 114 Model

4. DISCUSSION

It is difficult to develop a definition of a good associative predictive model. However, large R^2 is considered an acceptable norm. Venkatapathy et al. have reported 0.31 as R^2 in a correlation between observed TD_{50} and LD_{50} calculated using TOPKAT. The authors considered this correlation value to be low [7]. Mazzatorta et al. have reported 0.54 as R^2 in the comparison of oral rat chronic LOAELs with calculated LOAELs obtained with an in-house developed procedure that relied on PLS with a genetic-algorithm variable selection. The study was implemented using Statistics Toolbox 5.0.1 of MATLAB 7.0.1 [8]. In the present project, the results showed that the model created had a relatively small R^2 value (0.35). Future work will be initiated to determine the underlying reason for the relatively small R^2 derived from the models that were created and to make improvements on subsequent models.

(1) Duration

First, emphasis was placed on the duration of the studies from which 218 NOELs were obtained. One hundred twenty two (122) NOELs out of 218 NOELs were obtained in the 28 day experiments, but the other NOELs were not. Average duration was 36.5 days (± 10.5 days of one standard deviation), while the longest duration was 98 days (each

duration is shown in appendix II). An associative model was built using the 122 NOELs that were obtained in the 28 day study. The model did not suggest any improvement. The model demonstrated a smaller R^2 than the 218Model, and the correlation between the observed NOELs and the calculated NOELs looked more scattered than the 218Model (results not shown).

(2) Variance among 12 institutions

Second, the 218 experiments were conducted in 12 different facilities. A variance among the facilities was expected. Not all facilities complied with the same guideline that the Japanese government had issued. Some complied, some did not. The effect of variance among facilities was analyzed using SAS (SAS 9.3 TS Level 1M0 W32_7PRO platform, SAS Institute Inc., Cary, North Carolina) with the command of "proc mixed." Eight (8) institutions were entered into SAS rather than the 12 institution because 4 institutions conducted only one experiment. All institutions are shown in appendix II. The results demonstrated that: 1) "Fit statistics" is 165.3 for -2 Res Log Likelihood, 167.3 for AIC (Akaike Information Criterion), 167.4 for AICC (AIC with a correction for finite sample sizes), and 167.4 for BIC (Bayesian Information Criterion), 2) p-value of "Solution for Fixed Effects" is less than 0.0001 for both intercept and observed, 3)

Solution for Random Effects are 0 for all eight institutions, and 4) p-value of Type 3 Tests of Fixed Effects was less than 0.0001. Therefore, the results of "proc mixed" suggested that there were no random effects among the institutions, meaning that the results of the correlation between the observed NOELs and the calculated NOELs were not dependent on the variance of the institutions.

Also, as shown in table 3-3, the result may suggest that there is a signal in at least some of the institutions rather than noises, but it is not very helpful to perform further analysis with stratified data, because the number of studies in each institution is not very large (13 to 41).

(3) Grouping the compounds

Finally, the 114Model was created in order to develop better associative models. This was done by dividing 218 compounds into groups (Figure 3-4). No useful chemical structural relationship among the selected 114 compounds or among the removed 104 compounds was found. No structural commonality was found in the 114 compounds (e.g., the number of benzene rings or the size of the compounds), nor did there seem to be any structural commonality in the removed 104 compounds. Although an effort was

made to predict the NOELs by the removal of 104 compounds using the 114Model, the results showed that the 114Model failed to predict. The reason this did not occur was because the 114 compounds were "cherry-picked" compounds according to the residuals. A model with the 104 compounds did not show good associative relationship. The model had small R², and correlation between observed NOELs and calculated NOELs seemed to be scattered. If the model had shown useful relationship, the results would have suggested that there existed two proper chemical compounds groups in the 218 compounds. Although the 114Model suggests that there may be some signal involving some of 218 compounds, the noise that caused the distorted relationship could not be identified and thus removed.

In data mining, how to remove the noise is a common issue. Julian-Ortiz et al. report interesting results [22]. At first, they built the model of chronic LOAEL by means of a multilinear regression (MLR) with 229 compounds that were originated from the U.S. Environmental Protection Agency (EPA) and the National Toxicology Program (NTP) database. They suggested that the model did not show great correlation between observed LOAEL and calculated LOAEL. However, when they used 86 compounds that originated from EPA (but not NTP) and built a model, the model showed a great improvement compared with the original model with all 229 compounds as 1) the residuals got smaller; 2) heteroscedasticity (dependence of the variance on the value of the experimental datum) disappeared; 3) the standard error of the estimates reduced; 4) correlation was improved; and 5) Fisher-Snedecor parameter F rose. Interestingly, they added a model with compounds that originated from NTP but this did not show useful improvement. They concluded that only data obtained from EPA could be properly modeled by means of MLR, while data obtained from NTP database introduced noise and did not provide good models by themselves or in combination with the EPA data. Also, they stated that better models could be built by using only reliable data.

5. CONCLUSION

In this project, the utility of the NOELs of the 218 compounds was tested by building models using QSAR modeling technique with PDM. The results suggested that although the model showed some association between observed NOELs and calculated NOELs, the model provided little evidence for good utility of the data. Also, the reason why the model could not achieve good associative relationship between the observed NOELs and calculated NOELs was considered, and the results analyzed with SAS. An effort was made to group the chemical compounds. The results suggest that, perhaps, both the "signal" and "noise" are present in the analyzed data, but efforts to filter out the noise was not successful. Once only the signal-contacting data are left, the model can be significantly improved. This work will be undertaken by the author in the future. In this respect, the present project will help to further the study of the QSAR methodology.

6. REFERENCES

- 1. Goldstein BD. Advances in Risk Assessment and Communication. *Annu. Rev. Public Health.* 2005;26:141–63.
- 2. Demchuk E, Ruiz P, Pohl HR, et al. Computational toxicology: the public health perspective. *Toxicologist*. 2008;102:413–414.
- 3. Demchuk E, Ruiz P, Chou S, Fowler BA. SAR/QSAR methods in public health practice. *Toxicol. Appl. Pharmacol.* 2011;254:192–197.
- 4. Demchuk E, Ruiz P, Wilson JD, et al. Computational Toxicology Methods in Public Health Practice. *Toxicology Mechanisms and Methods*. 2008;18:119–135.
- Rupp B, Appel KE, Gundert-Remy U. Chronic oral LOAEL prediction by using a commercially available computational QSAR tool. *Arch Toxicol*. 2010;84:681–688.
- Venkatapathy R, Moudgal CJ, Bruce RM. Assessment of the Oral Rat Chronic Lowest Observed Adverse Effect Level Model in TOPKAT, a QSAR Software Package for Toxicity Prediction. J. Chem. Inf. Comput. Sci. 2004;44:1623–1629.
- 7. Venkatapathy R, Wang CY, Bruce RM, Moudgal C. Development of quantitative structure–activity relationship (QSAR) models to predict the carcinogenic potency of chemicals I. Alternative toxicity measures as an estimator of carcinogenic potency. *Toxicology and Applied Pharmacology*. 2009;234:209–221.
- Mazzatorta P, Estevez MD, Coulet M, Schilter B. Modeling Oral Rat Chronic Toxicity. J. Chem. Inf. Model. 2008;48: 1949–1954.
- Mombelli E. An Evaluation of the Predictive Ability of the QSAR Software Packages, DEREK, HAZARDEXPERT and TOPKAT, to Describe Chemically-induced Skin Irritation. *ATLA*. 2008;36:15–24.

- 10. Arvidson KB. FDA toxicity databases and real-time data entry. *Toxicology and Applied Pharmacology*. 2008;233:17–19.
- Cross KP, Myatt G, Yang C, Fligner MA, Verducci JS, Blower PE Jr. Finding Discriminating Structural Features by Reassembling Common Building Blocks. J. Med. Chem. 2003;46:4770–4775.
- Roberts G, Myatt GJ, Johnson WP, Cross KP, Blower PE Jr. LeadScope: Software for Exploring Large Sets of Screening Data. J. Chem. Inf. Comput. Sci. 2000;40:1302–1314.
- Blower P, Fligner M, Verducci J, Bjoraker J. On Combining Recursive Partitioning and Simulated Annealing To Detect Groups of Biologically Active Compounds. J. Chem. Inf. Comput. Sci. 2002;42:393–404.
- Blower PE, Yang C, Fligner MA, et al. Pharmacogenomic analysis: correlating molecular substructure classes with microarray gene expression data. *The Pharmacogenomics Journal*. 2002;2:259–271.
- 15. Blower PE Jr, Cross KP, Fligner MA, Myatt GJ, Verducci JS, Yang C. Systematic Analysis of Large Screening Sets in Drug Discovery. *Current Drug Discovery Technologies*. 2004;1:37–47.
- 16. Blower PE, Cross KP. Decision Tree Methods in Pharmaceutical Research. *Current Topics in Medicinal Chemistry.* 2006;6:31–39.
- Yang C, Cross K, Myatt GJ, Blower PE, Rathman JF. Building Predictive Models for Protein Tyrosine Phosphatase 1B Inhibitors Based on Discriminating Structural Features by Reassembling Medicinal Chemistry Building Blocks. *J. Med. Chem.* 2004;47:5984–5994.
- Yang C, Hasselgren CH, Boyer S, et al. Understanding Genetic Toxicity Through Data Mining: The Process of Building Knowledge by Integrating Multiple Genetic Toxicity Databases. *Toxicology Mechanisms and Methods*. 2008;18:277–295.

- The National Institute of Health Science, Japan. Japan Existing Chemical Database. Available at <u>http://www.nihs.go.jp/index-j.html</u>. Accessed February 20, 2013.
- 20. Online SMILES Translator and Structure File Generator. National Cancer Institute. Available at <u>http://cactus.nci.nih.gov/translate/</u>. Accessed February 20, 2013.
- 21. Leadscope User Manual. Enterprise Version 2.2. Personal Version 3.2.
- 22. Julian-Ortiz JV, Garcia-Domenech R, Galvez J, Pogliani L. Predictability and prediction of lowest observed adverse effect levels in a structurally heterogeneous set of chemicals. SAR and QSAR in Environmental Research. 2005;16:263–272.

APPENDICES

I. General Information for the 218 Chemical Compounds (Parent compounds)

No.	Name	Compound ID ^a	SMILES	MW ^b	Formula	с
1	4-Vinylpyridine	Taka100-43-6	C1=NC=CC(=C1)C=C	105.14	C ₇ H ₇ N	1
2	3-Cyanopyridine	Taka100-54-9	C1=CC(=CN=C1)C#N	104.12	$C_6H_4N_2$	1
3	2-Vinylpyridine	Taka100-69-6	C1=C(C=C)N=CC=C1	105.15	C ₇ H ₇ N	1
4	4-Ethylmorpholine	Taka100-74-3	N1(CCOCC1)CC	115.18	C ₆ H ₁₃ NO	1
5	4,4'-Methylenebis(2-chloro	Taka101-14-4	C2=C(CC1=CC(=C(N)C=C1)Cl)C=	267.15	$C_{13}H_{12}Cl_2N_2$	
	aniline)		CC(=C2Cl)N			
6	Dicyclohexylamine	Taka101-83-7	C2C(NC1CCCCC1)CCCC2	181.32	C ₁₂ H ₂₃ N	1
7	1,3-Diphenylguanidine	Taka102-06-7	C1=CC=CC=C1NC(N)=NC2=CC=	211.26	$C_{13}H_{13}N_3$	1
			CC=C2			
8	Glyceroltriacetate	Taka102-76-1	C(C(OC(=O)C)COC(C)=O)OC(C)=	218.20	$C_9H_{14}O_6$	1
			0			
9	2-(Di-n-butylamino)ethanol	Taka102-81-8	N(CCCC)(CCCC)CCO	173.30	C ₁₀ H ₂₃ NO	
10	Bis(2-ethylhexyl)nonanedio	Taka103-24-2	C(OC(=O)CCCCCCC(OCC(CCC	412.66	$C_{25}H_{48}O_4$	1
	ate		C)CC)=O)C(CCCC)CC			
11	2-ethylhexylvinylether	Taka103-44-6	C(OC=C)C(CCCC)CC	156.27	$C_{10}H_{20}O$	
12	N-Ethylaniline	Taka103-69-5	C1=C(NCC)C=CC=C1	121.20	$C_8H_{11}N$	
13	N,N-Dimethylbenzylamine	Taka103-83-3	c1(cccc1)CN(C)C	135.23	C ₉ H ₁₃ N	1
14	1,4-Diethylbenzene	Taka105-05-5	c1(ccc(CC)cc1)CC	134.22	$C_{10}H_{14}$	1
15	2-(Diethylamino)ethylmeth	Taka105-16-8	C(N(CC)CC)COC(C(C)=C)=O	185.27	$C_{10}H_{19}O_2N$	1
	acrylate					
16	Methylacetoacetate	Taka105-45-3	C(C(OC)=O)C(C)=O	116.12	$C_5H_8O_3$	1
17	Dibutyladipate	Taka105-99-7	C(C)CCOC(CCCC(OCCCC)=0)=	258.40	$C_{14}H_{26}O_4$	1
			0			
18	1,4-Dibromobenzene	Taka106-37-6	C1=C(C=CC(=C1)Br)Br	235.92	$C_6H_4Br_2$	
19	4-Chlorophenol	Taka106-48-9	C1=C(C=CC(=C1)Cl)O	128.56	C ₆ H ₅ ClO	1
20	2,3-Epoxypropylmethacryla	Taka106-91-2	C(OC(C(C)=C)=O)C1CO1	142.17	$C_7 H_{10} O_3$	
	te					
21	3-Methylphenol	Taka108-39-4	C1=C(C)C=CC=C1O	108.14	C ₇ H ₈ O	1
22	Propyleneglycolmonometh	Taka108-65-6	C(OC)C(OC(C)=O)C	132.16	$C_{6}H_{12}O_{3}$	1
	yletheracetate					
23	3,5-Dimethylaniline	Taka108-69-0	C1=C(C)C=C(C=C1C)N	121.20	$C_8H_{11}N$	1

	[1			1	-
24	1,3,5-Trihydroxybenzene	Taka108-73-6	C1=C(0)C=C(0)C=C10	126.11	C ₆ H ₆ O ₃	
25	Isocyanuricacid	Taka108-80-5	O=C1NC(NC(N1)=O)=O	129.09	$C_3H_3N_3O_3$	
26	4-Methylpyridine	Taka108-89-4	C1=CN=CC=C1C	93.10	C ₆ H ₇ N	
27	1,3-Dibromopropane	Taka109-64-8	C(CBr)CBr	201.89	$C_3H_6Br_2$	
28	1-Bromo-3-chloropropane	Taka109-70-6	C(CCl)CBr	157.44	C ₃ H ₆ BrCl	1
29	Thiophene	Taka110-02-1	C1=CC=CS1	84.14	C_4H_4S	1
30	1,2-Bis(stearoylamino)etha	Taka110-30-5	C(NC(CCCCCCCCCCCCCCC)	593.02	$C_{38}H_{76}N_2O_2$	
	ne		=0)CNC(CCCCCCCCCCCCCCC			
			C)=O			
31	Tetrahydromethy-1,3-isobe	Taka11070-44-3	CC12C(=CCCC1)C(OC2=O)=O	166.18	$C_9H_{10}O_3$	1
	nzofuranedione					
32	3,3'-Thiobispropanoicacid	Taka111-17-1	C(C(O)=O)CSCCC(O)=O	178.21	$C_6H_{10}O_4S$	1
33	N-(Aminoethyl)ethanolami	Taka111-41-1	N(CCN)CCO	104.15	$C_4H_{12}N_2O$	1
	ne					
34	Methyldodecanoate	Taka111-82-0	C(C)CCCCCCCC(OC)=O	214.35	$C_{13}H_{26}O_2$	1
35	1-Octanethiol	Taka111-88-6	C(CCCCS)CCC	146.30	$C_8H_{18}S$	
36	Undecane	Taka1120-21-4	C(C)CCCCCCCC	156.31	$C_{11}H_{24}$	
37	1,2-Bis(2-chloroethoxy)eth	Taka112-26-5	C(OCCCl)COCCCl	187.07	$C_6H_{12}Cl_2O_2$	1
	ane					
38	Docosanoicacid	Taka112-85-6	C(CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC	340.59	$C_{22}H_{44}O_2$	
			C(=O)O			
39	2-Amino-2-ethyl-1,3-propa	Taka115-70-8	C(C(N)(CO)CO)C	119.16	$C_5H_{13}NO_2$	1
	nediol					
40	Pentaerythritol	Taka115-77-5	C(C(CO)(CO)CO)O	136.17	$C_5H_{12}O_4$	
41	2,6-Dichlorotoluene	Taka118-69-4	C1=CC=C(C(=C1Cl)C)Cl	161.03	$C_7H_6Cl_2$	1
42	2,4,6-Tribromophenol	Taka118-79-6	c1(c(cc(Br)cc1Br)Br)O	330.80	C ₆ H ₃ Br ₃ O	
43	4,4'-Methylenebis(2,6-di-te	Taka118-82-1	C(c1c(c(C(C)(C)C)cc(c1)Cc1cc(C(424.65	$C_{29}H_{44}O_2$	
	rt-butylphenol)		C)(C)C)c(O)c(c1)C(C)(C)C)O)(C)(
			C)C			
44	Ditridecylphthalate	Taka119-06-2	C1=CC(=C(C=C1)C(OCCCCCCCC	530.83	$C_{34}H_{58}O_4$	
			000000000000000000000000000000000000000			
			CC)=O			
45	2,2'-Methylenebis(6-tert-bu	Taka119-47-1	C1=C(C=C(C(=C1C(C)(C)C)O)CC	340.51	$C_{23}H_{32}O_2$	1
	tyl-p-cresol)		2=CC(=CC(=C2O)C(C)(C)C)C)C			
46	2-Methyl-5-nitrobenzenesul	Taka121-03-9	C1=C(C=CC(=C1[S](O)(=O)=O)C)	217.20	C7H7NO5S	1

	fonicacid		[N+](=O)[O-]			
47	3-Aminobenzenesulfonicac id	Taka121-47-1	C1=C(C=C(C=C1)N)[S](=O)(O)=O	173.20	C ₆ H ₇ NO ₃ S	1
48	4-chlorobenzoylchloride	Taka122-01-0	c1(C(=O)Cl)ccc(Cl)cc1	175.01	C ₇ H ₄ Cl ₂ O	1
49	4-Ethylphenol	Taka123-07-9	C1=C(CC)C=CC(=C1)O	122.16	C ₈ H ₁₀ O	1
50	4-Methoxybenzaldehyde	Taka123-11-5	C1=CC(=CC=C1C=O)OC	136.15	$C_8H_8O_2$	1
51	4-Aminophenol	Taka123-30-8	c1(ccc(O)cc1)N	109.14	C ₆ H ₇ NO	1
52	Diacetonealcohol	Taka123-42-2	C(C(C)=O)C(C)(O)C	116.18	$C_{6}H_{12}O_{2}$	
53	Paraldehyde	Taka123-63-7	C1(OC(OC(O1)C)C)C	132.16	C ₆ H ₁₂ O ₃	
54	Azodicarboxamide	Taka123-77-3	O=C(N)N=NC(=O)N	116.08	$C_2H_4N_4O_2$	1
55	2,2-Dimethyl-1,3-propaned iol	Taka126-30-7	C(C(C)(C)CO)O	104.15	$C_{5}H_{12}O_{2}$	1
56	Tetrahydrothiophene-1,1-di oxide	Taka126-33-0	O=[\$]1(=O)CCCC1	120.17	$C_4H_8O_2S$	1
57	Sodium3-nitrobenzenesulfo nate	Taka127-68-4	c1(cc(ccc1)[N+](=O)[O-])S(=O)(=O)	225.15	C ₆ H ₄ NNaO ₅ S	1
58	3 Sodium4-amino-1-naphthal Taka130-13-2 C1=CC=CC2=C(C		C1=CC=CC2=C(C=CC(=C12)[S](=	245.24	C ₁₀ H ₈ NNaO ₃	1
	enesulfonate		O)(=O)[O-])N.[Na+]		S	
59	Divinylbenzene	Taka1321-74-0	C1=CC=CC(=C1C=C)C=C	130.19	$C_{10}H_{10}$	1
60	Sorbitanmonooctadecanoat	Taka1338-41-6	[C@@H]1(OC[C@H]([C@@H]1O	430.62	$C_{24}H_{46}O_{6}$	
	e)0)C(0)COC(=0)CCCCCCCCC CCCCCC			
61	Sodium2-naphthol-3,6-disu	Taka135-51-3	C1=C2C(=CC(=C1[S](=O)(=O)[O-]	348.26	C ₁₀ H ₆ Na ₂ O ₇ S	
	lfonate)O)C=CC(=C2)[S](=O)(=O)[O-].[N a+].[Na+]		2	
62	p-tert-Octylphenol	Taka140-66-9	C1=C(C(CC(C)(C)C)(C)C)C=CC(= $C1)O$	206.32	C ₁₄ H ₂₂ O	
63	1,3,5-Tri-tert-butylbenzene	Taka1460-02-2	C1=C(C(C)(C)C)C=C(C=C1C(C)(C)C)C(C)(C)C	246.43	C ₁₈ H ₃₀	
64	1,3-Bis(aminomethyl)benze ne	Taka1477-55-0	C1=C(CN)C=CC=C1CN	136.19	$C_8H_{12}N_2$	1
65	3,3-Bis(p-dimethylaminoph enyl)-6-dimethylaminophth alide	Taka1552-42-7	C1=CC(=CC4=C1C(C2=CC=C(N(C)C)C=C2)(C3=CC=C(N(C)C)C=C 3)OC4=O)N(C)C	415.54	$C_{26}H_{29}N_3O_2$	
		1				1

67	4-Chloro-o-cresol	Taka1570-64-5	C1=C(C(=CC(=C1)Cl)C)O	142.60	C7H7ClO	
68	3,4,5,6-Tetrachlorophthali	Taka1571-13-7	C1(=C(Cl)C(=C(C2=C1C(NC2=O)	284.91	C ₈ HCl ₄ NO ₂	1
	mide		=O)Cl)Cl)Cl			
69	Perfluorooctadecanoicacid	Taka16517-11-6	O=C(C(C(C(C(C(C(C(C(C(C(C(C(C(C(C(C(C(C(914.14	C ₁₈ HF ₃₅ O ₂	1
			C(C(C(C(C(F)(F)F)(F)F)(F)F)(F)F)(F)F)(F			
			F)F)(F)F)(F)F)(F)F)(F)F)(F)F)(F)F)(F)F			
			F)F)(F)F)(F)F)(F)F)(F)F)O			
70	4,4'-Bis(chloromethyl)-1,1'-	Taka1667-10-3	C1=CC(=CC=C1C2=CC=C(C=C2)	251.15	$C_{14}H_{12}Cl_2$	
	biphenyl		CCI)CCI			
71	Ethylcyclohexane	Taka1678-91-7	C(C)C1CCCCC1	112.21	$C_{8}H_{16}$	
72	6-tert-Butyl-2,4-xylenol	Taka1879-09-0	C1=C(C(=C(C=C1C)C(C)(C)C)O)C	178.30	$C_{12}H_{18}O$	1
73	tert-pentylbenzene	Taka2049-95-8	c1(C(CC)(C)C)ccccc1	148.24	$C_{11}H_{16}$	
74	1-Chloro-2,5-dimethoxyben	Taka2100-42-7	C1=C(OC)C=CC(=C1Cl)OC	172.61	$C_8H_9O_2Cl$	1
	zene					
75	isobutyl2-naphthylether	Taka2173-57-1	C1=C2C(=CC=C1OCC(C)C)C=CC	200.28	$\mathrm{C}_{14}\mathrm{H}_{16}\mathrm{O}$	1
			=C2			
76	1-Methoxynaphthalene	thalene Taka2216-69-5 C1=CC=C2C(=C1OC)C=CC=C2 1		158.20	$C_{11}H_{10}O$	1
77	2,3,6-Trimethylphenol	Taka2416-94-6 C1=C(C(=C(O)C(=C1)C)C)C 13		136.19	$C_9H_{12}O$	
78	2-(Dimethylamino)ethylacr	Taka2439-35-2	C(OC(C=C)=O)CN(C)C	143.21	$C_7H_{13}NO_2$	
	ylate					
79	Tripropyleneglycol	Taka24800-44-0	C(C(O)OCC(C)OC(CC)O)C	192.29	$C_9H_{20}O_4$	1
80	Nonylphenol	Taka25154-52-3 C1=C(C(=CC=C1)[O-])CCCCCCC		220.36	$C_{15}H_{24}O$	
			CC.[K+]			
81	Diisopropylbenzene	Taka25321-09-9	C1=CC=CC(=C1C(C)C)C(C)C	162.27	$C_{12}H_{18}$	
82	1,3-Cyclohexanedimethana	Taka2579-20-6	[C@H]1(C[C@@H](CCC1)C[NH3	142.24	$C_8H_{18}N_2$	1
	mine		+])C[NH3+]			
83	3-Methyl-4-nitrophenol	Taka2581-34-2	C1=C([N+]([O-])=O)C(=CC(=C1)O	153.15	$C_7H_7NO_3$	1
)C			
84	Dibenzyltoluene	Taka26898-17-9	C1=C(C(=C(C=C1)C)CC2=CC=CC	272.38	$C_{21}H_{20}$	1
			=C2)CC3=CC=CC=C3			
85	Sodiump-styrenesulfonate	Taka2695-37-6	C1=C([S]([O-])(=O)=O)C=CC(=C1	206.19	C ₈ H ₇ NaO ₃ S	
)C=C.[Na+]			
86	1,3,5-Tris(3,5-di-tert-butyl-	Taka27676-62-6	n1(Cc2cc(c(O)c(c2)C(C)(C)C)C(C)(784.10	$C_{48}H_{69}N_3O_6$	
	4-hydroxybenzyl)isocyanur		C)C)c(=O)n(Cc2cc(C(C)(C)C)c(c(c			
	icacid		2)C(C)(C)C)O)c(=O)n(Cc2cc(C(C)(

			C)C)c(c(c2)C(C)(C)C)O)c1=O			
87	Tricyclo[3.3.1.13,7]decane	Taka281-23-2	C12CC3CC(C1)CC(C2)C3	136.23	C ₁₀ H ₁₆	1
88	Diethylbiphenyl	Taka28575-17-9	C1=C(C=C(C=C1)CC)C2=CC(=CC =C2)CC	210.31	C ₁₆ H ₁₈	1
89	2-(Dimethylamino)ethylme thacrylate	Taka2867-47-2	C(OC(C(C)=C)=O)CN(C)C	157.24	C ₈ H ₁₅ NO ₂	1
90	Monoisopropylnaphthalene	Taka29253-36-9	c1(cccc2ccccc12)C(C)C	170.25	C ₁₃ H ₁₄	1
91	Perfluorooctane	Taka307-34-6	FC(C(C(C(F)(C(F)(F)C(C(C(F)(F)F))(F)F)(F)(438.06	C_8F_{18}	
92	1,2-Dichloro-3-nitrobenzen e	Taka3209-22-1	C1=C([N+]([O-])=O)C(=C(Cl)C=C 1)Cl	192.00	$C_6H_3NO_2Cl_2$	
93	zenetricarboxylate		C1=C(C(=CC=C1C(OCC(CCCC)C C)=O)C(OCC(CCCC)CC)=O)C(OC C(CCCC)CC)=O	546.87	C ₃₃ H ₅₄ O ₆	1
94	3,5,5-Trimethylhexan-1-ol	Taka3452-97-9	C(C(CCO)C)C(C)(C)C	143.26	C ₉ H ₂₀ O	
95	3-Phenoxytoluene Taka3586-14-9		C2=C(OC1=CC=CC=C1)C=CC=C2 C	184.25	C ₁₃ H ₁₂ O	1
96	6 Diheptylphthalate Taka3648-21-3		c1(c(C(OCCCCCCC)=O)cccc1)C(O CCCCCCC)=O	362.51	C ₂₂ H ₃₄ O ₄	
97	Sodium1-methoxycarbonyl pentadecane-2-sulfonate	Taka4016-24-4	C(C([S]([O-])(=O)=O)C(OCC)=O) CCCCCCCCCCCC.[Na+]	372.50	C ₁₇ H ₃₃ NaO ₅ S	
98	2,6-Bis(1,1-dimethylethyl)- 4-ethylphenol	Taka4130-42-1	C1=C(C=C(C(=C1C(C)(C)C)O)C(C)(C)C)CC	234.38	C ₁₆ H ₂₆ O	1
99	Thioureadioxide	Taka4189-44-0	O=[S]([O-])C(N)=N	108.12	CH ₄ N ₂ O ₂ S	
100	2,2',3,3'-Tetrachloro-4,4'-di aminodiphenylmethane	Taka42240-73-3	C1=CC(=C(C(=C1CC2=CC=C(C(= C2Cl)Cl)N)Cl)Cl)N	336.04	$C_{13}H_{10}Cl_4N_2$	
101	4-(1-Methylethenyl)phenol	Taka4286-23-1	CC(=C)C1=CC=C(C=C1)O	134.18	C ₉ H ₁₀ O	1
102	2,2,4,4,6,8,8-Heptamethyln onane	Taka4390-04-9	C(C(CC(CC(C)(C)C)C)(C)C)C(C)(C)C	226.44	C ₁₆ H ₃₄	
103	3-Methyl-1,5-pentanediol	Taka4457-71-0	C(C(CCO)C)CO	118.20	C ₆ H ₁₄ O ₂	
104	Methoxymethanol	Taka4461-52-3	C(0)OC	62.07	C ₂ H ₆ O ₂	
105	Cyanoguanidine	Taka461-58-5	C(=NC#N)(N)N	84.08	$C_2H_4N_4$	
106	N,N-Dicyclohexyl-2-benzot Taka4979-32-2 hiazolesulfenamide		C1=CC=CC4=C1N=C(SN(C2CCC CC2)C3CCCCC3)S4	346.59	$C_{19}H_{26}N_2S_2$	1
107	(Methacryloyloxyethyl)trim	Taka5039-78-1	C([N+](C)(C)C)COC(C(C)=C)=O.[207.70	C ₉ H ₁₈ ClNO ₂	

	ethylammoniumchloride		Cl-]			
108	DisperseYellow42	Taka5124-25-4	C1=C([N+]([O-])=O)C(=CC=C1[S]	369.40	$C_{18}H_{15}N_3O_4S$	1
			(=0)(=0)NC2=CC=CC=C2)NC3=C			
			C=CC=C3			
109	2,4-Dinitrophenol	Taka51-28-5	C1=C(C(=CC=C1[N+](=O)[O-])O)[184.11	$C_6H_4N_2O_5$	
			N+](=O)[O-]			
110	Ethylcarbamate	Taka51-79-6	C(OC(N)=O)C	89.03	C ₃ H ₇ NO ₂	
111	2-Bromo-2-nitropropane-1, 3-diol	Taka52-51-7	C(C([N+](=O)[O-])(CO)Br)O	199.99	C ₃ H ₆ BrNO ₄	
112	2,3-Dibromosuccinicacid	Taka526-78-3	O=C(C(C(C(O)=O)Br)Br)O	275.88	$C_4H_4O_4Br_2$	
113	Dicyclohexylcarbodiimide	Taka538-75-0	C2C(N=C=NC1CCCCC1)CCCC2	206.33	$C_{13}H_{22}N_2$	
114	Citral	Taka5392-40-5	C(\C(=C\C=0)C)CC=C(C)C	152.23	C ₁₀ H ₁₆ O	
115	chlorocyclohexane	Taka542-18-7	C1(CCCCC1)Cl	118.60	C ₆ H ₁₁ Cl	1
116	n-Hexadecane	Taka544-76-3	C(CCCCCCCC)CCCCCCC	226.44	C ₁₆ H ₃₄	
117	Monosodium4-amino-5-hy	Taka5460-09-3	C1=C2C(=C(C=C1[S](O)(=O)=O)N	341.29	C ₁₀ H ₈ NNaO ₇	
	droxy-2,7-naphthalenedisul)C(=CC(=C2)[S](=O)(=O)[O-])O.[S_2	
	fonate		Na+]			
118	Methane, isothiocyanato- Taka 556-61-6 S=		S=C=NC	73.12	C ₂ H ₃ NS	
119	3-Methoxy-3-methyl-1-but	Taka56539-66-3	C(C(OC)(C)C)CO	118.18	$C_6H_{14}O_2$	1
	anol					
120	Benzyltrimethylammonium	Taka56-93-9	C1=CC=CC=C1C[N+](C)(C)C.[C]-	185.70	C ₁₀ H ₁₆ ClN	1
	chloride]			
121	4-Ethyl-1,1'-biphenyl	Taka5707-44-8	C2=C(C1=CC=CC=C1)C=CC(=C2)	182.26	$C_{14}H_{14}$	
			CC			
122	1,2-Butanediol	Taka584-03-2	C(C)C(CO)O	90.12	$C_4 H_{10} O_2$	1
123	tert-Butylmethacrylate	Taka585-07-9	CC(OC(C(C)=C)=O)(C)C	142.20	$C_8H_{14}O_2$	1
124	4-Chloro-m-cresol	Taka59-50-7	C1=C(C(=CC=C1O)Cl)C	142.58	C7H7ClO	1
125	4-(1-Methyl-1-phenylethyl)	Taka599-64-4	C2=C(C(C1=CC=CC=C1)(C)C)C=	212.29	C ₁₅ H ₁₆ O	1
	phenol		CC(=C2)O			
126	3-Nitrophthalicacid	Taka603-11-2	C1=C(C(=C(C=C1)[N+](=O)[O-])C	211.13	C ₈ H ₅ NO ₆	1
			(=O)O)C(=O)O			
127	1-Naphthol-4-sulfonicacids	Taka6099-57-6	C1=CC=CC2=C(C=CC(=C12)[S](=	160.17	C ₁₀ H ₇ NaO ₄ S	1
	odiumsalt		O)(=O)[O-])O.[Na+]			
128	Disodiumsuccinatehexahyd	Taka6106-21-4	C(CC([0-])=0)C([0-])=0.0.0.0	270.14	C ₄ H ₁₆ Na ₂ O ₁₀	
	rate		.O.O.[Na+].[Na+]			

129	1-Chloro-2-(chloromethyl)	Taka611-19-8	C1=CC=CC(=C1CCl)Cl	161.03	C ₇ H ₆ Cl ₂	
	benzene					
130	3-Ethylphenol	Taka620-17-7	C1=C(C=CC=C1CC)O	122.16	$C_8H_{10}O$	
131	4,4'-Methylenediphenol	Taka620-92-8	c1(Cc2ccc(O)cc2)ccc(O)cc1	200.23	$C_{13}H_{12}O_2$	1
132	1,4-Dicyanobenzene	Taka623-26-7	C1=CC(=CC=C1C#N)C#N	128.14	$C_8H_4N_2$	1
133	n-Pentadecane	Taka629-62-9	C(CCCCCCC)CCCCCCC	212.41	$C_{15}H_{32}$	
134	1,3-Benzenedicarboxylicaci	Taka6362-79-4	C1=C(C=C(C=C1[S](=O)(=O)[O-])	268.18	C ₈ H ₅ NaO ₇ S	
	d,5-sulfo-,monosodiumsalt		C(O)=O)C(O)=O.[Na+]			
135	Phenylurea	Taka64-10-8	C1=C(C=CC=C1)NC(N)=O	136.15	$C_7H_8N_2O$	1
136	C.I.PigmentRed22	Taka6448-95-9	C1=C2C(=CC=C1)C=C(C(=O)\C2=	426.43	$C_{24}H_{18}N_4O_4$	
			N/NC3=C(C=CC(=C3)[N+]([O-])=			
			O)C)C(=O)NC4=CC=CC=C4			
137	PigmentOrange16	Taka6505-28-8	C1=CC=CC=C1NC(=O)C(N=NC2=	620.65	$C_{34}H_{32}N_6O_6$	1
			CC=C(C=C2OC)C3=CC(=C(C=C3)			
			N=NC(C(=O)C)C(=O)NC4=CC=C			
			C=C4)OC)C(=O)C			
138	Sodiump-toluenesulfonate	Taka657-84-1	C1=C([S](=O)(=O)[O-])C=CC(=C1	194.18	C7H7NaO3S	
)C.[Na+]			
139	N-(Carboxymethyl)-N,N-di	Taka683-10-3	C([N+](CCCCCCCCCC)(C)C)C	271.44	C ₁₆ H ₃₃ NO ₂	
	methyl-1-dodecanaminium,		([O-])=O.[Cl-].[Na+]			
	innersalt					
140	2,2,4-Trimethyl-1,3-pentan	Taka6846-50-0	C(C(C(C(C(C)C)OC(=O)C(C)C)(C)C))	286.46	$C_{16}H_{30}O_4$	
	edioldiisobutyrate		OC(=O)C(C)C			
141	2-Ethylhexylmethacrylate	Taka688-84-6	C(C(CCCC)CC)OC(C(C)=C)=O	198.34	$C_{12}H_{22}O_2$	1
142	4-Methyl-1-pentene	Taka691-37-2	C(C(C)C)C=C	84.16	C ₆ H ₁₂	1
143	Pentaerythritoltetra(2-ethyl	Taka7299-99-2	C(OC(=O)C(CCCC)CC)C(COC(=O	640.94	$C_{37}H_{68}O_8$	1
	hexanoate))C(CCCC)(CO)(COC(=0)C(CCCC)			
			CC)COC(=0)C(CCCC)CC			
144	Trimethylamine	Taka75-50-3	N(C)(C)C	59.11	C ₃ H ₉ N	1
145	2-tert-Butoxyethanol	Taka7580-85-0	C(OC(C)(C)C)CO	118.17	$C_6H_{14}O_2$	
146	3,4-Dichloro-1-butene	Taka760-23-6	C(C(C=C)Cl)Cl	124.99	$C_4H_6Cl_2$	
147	Triphenylchloromethane	Taka76-83-5	C3=C(C(C1=CC=CC=C1)(C2=CC=	278.78	C ₁₉ H ₁₅ Cl	
			CC=C2)Cl)C=CC=C3			
148	Triisobutylene	Taka7756-94-7	C(C)C	168.30	$C_{12}H_{24}$	
149	1,1,1-Tris(hydroxymethyl)e	Taka77-85-0	C(C(C)(CO)CO)O	120.15	C ₅ H ₁₂ O ₃	1

	thane					
150	2-Ethyl-2-hydroxymethyl-1 ,3-propanediol	Taka77-99-6	C(C(CO)(CO)CO)C	134.17	$C_{6}H_{14}O_{3}$	1
151	2,5-Dimethyl-2,5-di(t-butyl	Taka78-63-7	C(C(OOC(C)(C)C)(C)C)CC(OOC(290.44	$C_{16}H_{34}O_4$	1
	peroxy)hexane		C)(C)C)(C)C			
152	2-Hydroxypropanenitrile	Taka78-97-7	CC(C#N)O	71.08	C ₃ H ₅ NO	
153	Tetrabromoethane	Taka79-27-6	C(C(Br)Br)(Br)Br	345.65	C ₂ H ₂ Br ₄	
154	N-(1,3-Dimethylbutyl)-N'-p	Taka793-24-8	C2=C(NC1=CC=CC=C1)C=CC(=C	268.40	$C_{18}H_{24}N_2$	
	henyl-p-phenylenediamine		2)NC(CC(C)C)C			
155	4,4'-Isopropylidenebis(2,6-	Taka79-94-7	C2=C(C(C1=CC(=C(0)C(=C1)Br) :		$C_{15}H_{12}Br_4O_2$	
	dibromophenol)		Br)(C)C)C=C(Br)C(=C2Br)O			
156	4,4'-Sulfonyldiphenol	Taka80-09-1	C2=C([S](C1=CC=C(O)C=C1)(=O)	250.27	$C_{12}H_{10}O_4S$	
			=O)C=CC(=C2)O			
157	Dicmylperoxide	Taka80-43-3	C1=CC=CC=C1C(OOC(C2=CC=C	270.37	$C_{18}H_{22}O_2$	1
			C=C2)(C)C)(C)C			
158	4,4'-Oxybis(benzenesulfon	Taka80-51-3	C1=CC(=CC=C1[S](NN)(=O)=O)O	358.40	$C_{12}H_{14}N_4O_5S_2$	
	ylhydrazide)		C2=CC=C([S](NN)(=O)=O)C=C2			
159	i9 2-Amino-1-naphthalenesulf Taka81-16-3 onicacid		c1(c2c(ccc1N)cccc2)S(O)(=O)=O	223.26	C ₁₀ H ₉ NO ₃ S	
160	Benzanthrone	Taka82-05-3	C1=CC=C4C3=C1C2=CC=CC=C2	230.26	C ₁₇ H ₁₀ O	1
			C(C3=CC=C4)=O			
161	p-Nitrophenolsodiumsalt	Taka824-78-2	C1=CC(=CC=C1[N+](=O)[O-])[O-]	161.09	C ₆ H ₄ NNaO ₃	
			.[Na+]			
162	Acenaphthene	Taka83-32-9	c12c3CCc1cccc2ccc3	154.22	$C_{12}H_{10}$	1
163	1,3,5-Tris(2-hydroxyethyl)-	Taka839-90-7	C(N1C(N(C(=O)N(C1=O)CCO)CC	261.23	$C_9H_{15}N_3O_6$	
	1,3,5-triazine-2,4,6-(1H,3H		O)=O)CO			
	,5H)-trione					
164	Dimethyl2,6-naphthalenedi	Taka840-65-3	C1=C2C(=CC=C1C(OC)=O)C=C(C	244.25	$C_{14}H_{12}O_4$	1
	carboxylate		(OC)=O)C=C2			
165	Potassium7-hydroxy-1,3-na	Taka842-18-2	C1=C([S]([O-])(=O)=O)C=C2C(=C	380.48	$C_{10}H_{6}K_{2}O_{7}S_{2}$	
	phthalenedisulfonate		1[S]([O-])(=O)=O)C=C(O)C=C2.[K			
			+].[K+]			
166	Phthalimide	Taka85-41-6	C1=CC=CC2=C1C(NC2=O)=O	147.13	C ₈ H ₅ NO ₂	1
167	1-Naphthylaceticacid	Taka86-87-3	C1=CC=CC2=CC=CC(=C12)CC(0	186.22	$C_{12}H_{10}O_2$	
)=O			

168	7-Amino-4-hydroxy-2-naph	Taka87-02-5	C1=CC(=CC2=CC(=CC(=C12)O)[S	239.25	C ₁₀ H ₉ NO ₄ S	1
	thalenesulfonicacid](O)(=O)=O)N			
169	2,6-Dimethylaniline	Taka87-62-7	C1=CC=C(C(=C1C)N)C	121.18	$C_8H_{11}N$	1
170	Chloropentabromocyclohex	Taka87-84-3	C1(C(C(C(C(C1Br)Br)Br)Br)Br)Cl	513.09	C ₆ H ₆ Br ₅ Cl	
	ane					
171	2-Ethylbutyricacid	Taka88-09-5	C(C(C(O)=O)CC)C	116.16	$C_6H_{12}O_2$	
172	2-tert-Butylphenol	Taka88-18-6	C1=CC=CC(=C1C(C)(C)C)O	150.22	$C_{10}H_{14}O$	1
173	o-Toluenesulfonamide	Taka88-19-7	c1(c(cccc1)C)S(=O)(=O)N	171.22	C ₇ H ₉ NO ₂ S	
174	2-Amino-5-methylbenzenes	Taka88-44-8	C1=C(C)C=CC(=C1[S](=O)(=O)O)	187.22	C7H9NO3S	1
	ulfonicacid		Ν			
175	2-Amino-5-chloro-4-methy	Taka88-53-9	C1=C([S](O)(=O)=O)C(=CC(=C1C	221.66	C7H8CINO3S	1
	l-benzenesulfonicacid		1)C)N			
176	6-tert-Butyl-m-cresol	Taka88-60-8	C(c1c(cc(C)cc1)O)(C)(C)C	164.25	$C_{11}H_{16}O$	1
177	2,4-Dimethylbenzenesulfon	Taka88-61-9	C1=CC(=CC(=C1[S](O)(=O)=O)C)	186.23	$C_8H_{10}O_3S$	1
	icacid		С			
178	2,4,6-Trinitrophenol	Taka88-89-1	C1=C([N+]([O-])=O)C=C([N+]([O-	229.10	$C_6H_3N_3O_7$	
])=O)C(=C1[N+]([O-])=O)O			
179	1,2,4-Benzenetricarboxylic	Taka89-04-3	C1=C(C(=CC=C1C(OCCCCCCCC)	546.78	C ₃₃ H ₅₄ O ₆	
	acid,trioctylester		=0)C(0CCCCCCC)=0)C(0CCC			
			CCCCC)=O			
180	benzene-1,2,4,5-tetracarbox	Taka89-05-4	c1(c(cc(C(O)=O)c(c1)C(O)=O)C(O)	254.15	$C_{10}H_6O_8$	1
	ylic		=O)C(O)=O			
181	1,4-Dichloro-2-nitrobenzen	Taka89-61-2	C1=C(Cl)C=CC(=C1[N+](=O)[O-])	192.00	C ₆ H ₃ Cl ₂ NO ₂	1
	e		Cl			
182	o-sec-Butylphenol	Taka89-72-5	c1(c(cccc1)O)[C@@H](CC)C	150.22	$C_{10}H_{14}O$	
183	Thymol	Taka89-83-8	C1=C(C(=CC(=C1)C)O)C(C)C	150.22	$C_{10}H_{14}O$	1
184	2-Hydroxybenzaldehyde	Taka90-02-8	C1=CC=CC(=C1C=O)O	122.12	C ₇ H ₆ O ₂	
185	1-Chloronaphthalene	Taka90-13-1	c12c(c(ccc1)Cl)cccc2	162.62	C ₁₀ H ₇ Cl	1
186	1,2-Dicyanobenzene	Taka91-15-6	C1=CC=CC(=C1C#N)C#N	128.13	$C_8H_4N_2$	1
187	DecahydroNaphthalene	Taka91-17-8	C12C(CCCC1)CCCC2	138.25	C10H18	
188	2,4-Diamino-6-phenyl-s-tri	Taka91-76-9	C1=CC=CC=C1C2=NC(=NC(=N2)	187.20	C ₉ H ₉ N ₅	
	azine		N)N			
189	AzoicCC5	Taka91-96-3	C2=C(C1=CC(=C(NC(CC(C)=O)=	380.44	$C_{22}H_{24}N_2O_4$	
			O)C=C1)C)C=CC(=C2C)NC(CC(C			
)=O)=O			

190	2-Hydroxypropylmethacryl ate	Taka923-26-2	C(OC(C(C)=C)=O)C(C)O	144.17	C ₇ H ₁₂ O ₃	
191	AzoicCC2	Taka92-77-3	O=C(Nc1ccccc1)c1c(O)cc2c(cccc2) c1	263.29	C ₁₇ H ₁₃ NO ₂	1
192	4,4'-Biphenyldiol	Taka92-88-6	C2=C(C1=CC=C(O)C=C1)C=CC(= C2)O	186.21	$C_{12}H_{10}O_2$	
193	o-Acetoacetotoluidide	Taka93-68-5	C1=CC=CC(=C1NC(CC(C)=O)=O) C	191.23	C ₁₁ H ₁₃ NO ₂	
194	2-(4-Morpholinyldithio)ben zothiazole	Taka95-32-9	C1=CC=CC3=C1N=C(SSN2CCOC C2)S3	284.43	$C_{11}H_{12}N_2OS_3$	1
195	N-Cyclohexyl-2-benzothiaz olesulfenamide	Taka95-33-0	c12c(sc(n1)SNC1CCCCC1)cccc2	264.43	$C_{13}H_{16}N_2S_2$	1
196	o-Dichlorobenzene	Taka95-50-1	C1=C(C(=CC=C1)Cl)Cl	147.00	$C_6H_4Cl_2$	1
197	2-Chlorophenol	Taka95-57-8	C1=C(C(=CC=C1)Cl)O	128.56	C ₆ H ₅ ClO	1
198	1,2,4-Trimethylbenzene	Taka95-63-6	C1=C(C)C=CC(=C1C)C	120.20	C ₉ H ₁₂	
199	3,4-Dimethylaniline	Taka95-64-7	c1(c(ccc(c1)N)C)C	121.20	C ₈ H ₁₁ N	1
200	Ethylmethylketoxime	Taka96-29-7	29-7 C(\C(=N\O)C)C		C ₄ H ₉ NO	
201	2-Imidazolidinethione	nidazolidinethione Taka96-45-7 C1NC(NC1)=S		102.16	$C_3H_6N_2S$	
202	4,4'-Thiobis(6-tert-butyl-m-	Taka96-69-5	C1=C(C(C)(C)C)C(=CC(=C1SC2= $CC(=C(C=C2C)O)C(C)(C)C)C)O$	358.54	C ₂₂ H ₃₀ O ₂ S	1
203	2,4-Di-tert-butylphenol	Taka96-76-4	C1=C(C=CC(=C1C(C)(C)C)O)C(C) (C)C	206.32	C ₁₄ H ₂₂ O	1
204	1,3-Bis(2-methylphenyl)gu anidine	Taka97-39-2	C2=C(NC(N)=NC1=CC=CC=C1C) C(=CC=C2)C	239.32	C ₁₅ H ₁₇ N ₃	1
205	4-Nitro-o-anisidine	Taka97-52-9	C1=C(C=CC(=C1OC)N)[N+](=O)[O-]	168.17	C ₇ H ₈ N ₂ O ₃	1
206	Butylmethacrylate	Taka97-88-1	C(OC(C(C)=C)=O)CCC	142.20	C ₈ H ₁₄ O ₂	
207	Tetrahydrofurfurylalcohol	Taka97-99-4	C(C1CCC01)0	102.13	C ₅ H ₁₀ O ₂	1
208	Trifluoromethylbenzene	Taka98-08-8	C1=C(C=CC=C1)C(F)(F)F	146.11	$C_7H_5F_3$	1
209	Benzenesulfonamide	Taka98-10-2	c1(ccccc1)S(N)(=O)=O	157.19	C ₆ H ₇ NO ₂ S	
210	p-tert-Butyltoluene	Taka98-51-1	C1=C(C(C)(C)C)C=CC(=C1)C	148.24	C ₁₁ H ₁₆	
211	p-tert-Butylphenol	Taka98-54-4	C1=C(C(C)(C)C)C=CC(=C1)O	150.22	C ₁₀ H ₁₄ O	1
212	1-Methylethenylbenzene	Taka98-83-9	c1(cccc1)C(C)=C	118.19	C ₉ H ₁₀	1
213	3-Methylbenzoicacid	Taka99-04-7	C1=C(C=CC=C1C)C(O)=O	136.15	C ₈ H ₈ O ₂	
214	3-Nitrobenzenamine	Taka99-09-2	C1=C(C=CC=C1[N+](=O)[O-])N	138.13	$C_6H_6N_2O_2$	

215	4-(1-Methylpropyl)phenol	Taka99-71-8	c1(ccc(O)cc1)[C@@H](CC)C	150.24	C ₁₀ H ₁₄ O	1
216	4-(1-Methylethyl)aniline	Taka99-88-7	C1=C(C(C)C)C=CC(=C1)N	135.21	$C_9H_{13}N$	1
217	4-Methylbenzoicacid	Taka99-94-5	c1(C(O)=O)ccc(C)cc1	136.15	$C_8H_8O_2$	
218	4-Hydroxybenzoic	Taka99-96-7	C1=CC(=CC=C1C(O)=O)O	138.13	$C_7H_6O_3$	

^a Compound ID is "Taka" + CAS number. CAS numbers are unique numerical identifiers

assigned by the Chemical Abstracts Service (CAS)

^b Molecular weight

^c 1 if the compound is contained in the 114Model

II. Experimental Data

No.			Obse	rved Data		Calculate	ed Data
	Study ^a	Duration	Institu-	NOEL	NOEL	Predicted	Residual
		(days)	tion ^b	(mg/kg/day)	(mmol/kg/day)	Value	Value
					(log scale)		
1	А	28	1	20	-0.7207	-0.8985	0.1778
2	А	28	7	5	-1.3186	-1.0270	-0.2916
3	А	28	5	12.5	-0.9249	-0.8985	-0.0264
4	А	28	3	50	-0.3624	-0.6880	0.3256
5	А	28	2	2	-2.1257	-1.3050	-0.8207
6	А	28	3	20	-0.9574	-0.9838	0.0264
7	А	28	5	10	-1.3248	-0.9358	-0.3890
8	В	48	4	1000	0.6611	0.0994	0.5618
9	А	28	3	25	-0.8409	-0.5402	-0.3007
10	В	42	3	300	-0.1385	-0.1548	0.0163
11	А	28	6	8	-1.2908	-0.6992	-0.5916
12	А	28	2	1	-2.0835	-1.3510	-0.7325
13	А	28	5	50	-0.4321	-0.8484	0.4163
14	В	51	5	30	-0.6507	-0.8981	0.2474
15	В	54	6	50	-0.5688	-0.4947	-0.0741
16	В	49	7	1000	0.9351	0.0980	0.8371
17	А	28	4	1000	0.5877	-0.1566	0.7443
18	А	28	2	4	-1.7707	-0.8198	-0.9509
19	А	28	7	20	-0.8081	-0.7730	-0.0351
20	В	47	0	10	-1.1528	-0.4613	-0.6915
21	А	28	1	100	-0.0340	-0.7656	0.7316
22	В	45	2	300	0.3560	0.0014	0.3546
23	А	28	7	10	-1.0835	-0.9283	-0.1552
24	А	28	2	300	0.3764	-0.5628	0.9392
25	В	48	2	600	0.6673	-0.4126	1.0800
26	В	46	8	5	-1.2700	-0.8367	-0.4333
27	А	28	6	10	-1.3051	-0.7803	-0.5248
28	А	28	3	20	-0.8961	-0.7808	-0.1153
29	В	42	3	25	-0.5271	-0.7871	0.2600

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30	А	28	5	1000	0.2269	-0.2295	0.4564
31	В	49	7	30	-0.7435	-0.6815	-0.0620
32	А	28	4	200	0.0501	-0.4086	0.4587
33	А	28	6	60	-0.2395	-0.5961	0.3566
34	В	55	5	1000	0.6689	0.0472	0.6217
35	В	35	4	10	-1.1652	-0.7398	-0.4254
36	В	46	1	100	-0.1940	-0.7443	0.5503
37	А	28	5	50	-0.5730	-0.6837	0.1107
38	В	42	3	1000	0.4678	-0.4757	0.9435
39	В	48	6	500	0.6228	-0.2871	0.9099
40	В	46	1	100	-0.1341	-0.1785	0.0444
41	В	42	3	30	-0.7298	-0.9053	0.1755
42	В	48	5	100	-0.5196	-0.7484	0.2288
43	А	28	2	8	-1.7249	-1.1570	-0.5679
44	В	42	3	10	-1.7250	-0.5161	-1.2090
45	D	52	8	12.5	-1.4352	-1.3920	-0.0432
46	В	50	6	175	-0.0938	0.3203	-0.4141
47	А	28	1	300	0.2386	0.1124	0.1262
48	В	48	6	100	-0.2431	-0.6254	0.3823
49	А	28	1	100	-0.0869	-0.7727	0.6858
50	В	42	3	20	-0.8330	-0.7410	-0.0920
51	А	28	4	20	-0.7370	-0.6566	-0.0804
52	В	45	2	30	-0.5880	-0.5307	-0.0573
53	А	28	6	100	-0.1211	-0.6656	0.5445
54	С	98	3	300	0.4124	-0.2560	0.6684
55	В	46	5	100	-0.0177	-0.3467	0.3290
56	А	28	2	60	-0.3016	-0.6424	0.3408
57	А	28	3	300	0.1246	0.1907	-0.0661
58	А	28	3	300	0.0875	0.3518	-0.2643
59	В	53	8	30	-0.6375	-0.8179	0.1804
60	В	42	3	1000	0.3659	0.7002	-0.3343
61	А	28	7	300	-0.0648	0.3523	-0.4171
62	А	28	4	15	-1.1385	-0.7399	-0.3986
63	В	52	2	2	-2.0907	-1.0430	-1.0480
64	D	45	5	50	-0.4352	-0.6876	0.2524

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65	A	28	4	120	-0.5394	-0.8681	0.3287
66	A	28	12	10	-1.1373	-0.7357	-0.4016
67	A	28	2	60	-0.3760	-1.0020	0.6260
68	Α	28	1	300	0.0224	-0.0945	0.1169
69	В	42	1	40	-1.3590	-0.6269	-0.7321
70	В	46	8	62.5	-0.6041	-1.1340	0.5299
71	А	28	2	40	-0.4480	-0.7925	0.3445
72	В	48	5	6	-1.4730	-1.3350	-0.1380
73	А	28	1	100	-0.1710	-0.8264	0.6554
74	А	28	1	40	-0.6350	-0.8103	0.1753
75	А	28	5	20	-1.0006	-0.7969	-0.2037
76	А	28	8	30	-0.7221	-0.6121	-0.1100
77	А	28	3	100	-0.1341	-1.0560	0.9218
78	В	43	4	4	-1.5539	-0.5558	-0.9981
79	В	49	8	200	0.0171	-0.5349	0.5520
80	А	28	6	15	-1.1670	-0.7979	-0.3691
81	А	28	7	30	-0.7331	-0.9466	0.2135
82	В	42	4	60	-0.3749	-0.6662	0.2913
83	D	46	1	100	-0.1851	-0.6366	0.4515
84	А	28	4	20	-1.1341	-1.1480	0.0139
85	А	28	1	100	-0.3143	-0.1698	-0.1445
86	А	28	6	1000	0.1056	-0.4240	0.5296
87	А	28	2	40	-0.5322	-0.7913	0.2591
88	А	28	3	60	-0.5447	-0.9353	0.3906
89	В	43	4	40	-0.5945	-0.4955	-0.0990
90	А	28	1	10	-1.2311	-1.1390	-0.0921
91	В	53	6	1000	0.3585	-0.9006	1.2590
92	В	44	2	5	-1.5843	-0.9384	-0.6459
93	А	28	5	1000	0.2621	-0.3205	0.5826
94	В	46	1	12	-1.0769	-0.6665	-0.4104
95	А	28	4	20	-0.9644	-1.0530	0.0886
96	А	28	1	62.5	-0.7634	-0.5212	-0.2422
97	В	55	2	20	-1.2701	0.0890	-1.3590
98	А	28	6	15	-1.1938	-1.0770	-0.1168
99	В	49	7	4	-1.4318	-0.5400	-0.8918

100	А	28	5	100	-0.5264	-1.3420	0.8156
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101	A	28	1	30	-0.6506	-0.6826	0.0320
102	A	28	3	100	-0.3550	-0.7559	0.4010
103	В	49	6	300	0.4045	-0.5796	0.9841
104	В	47	4	12	-0.7137	-0.6587	-0.0550
105	D	46	4	1000	1.0753	-0.5634	1.6390
106	В	51	2	25	-1.1419	-0.7716	-0.3703
107	А	28	3	500	0.3815	-0.4989	0.8804
108	А	28	3	300	-0.0904	-0.4550	0.3646
109	А	28	2	10	-1.2651	-0.7419	-0.5232
110	А	28	2	20	-0.6485	-0.5033	-0.1452
111	А	28	2	10	-1.3010	-0.3910	-0.9100
112	А	28	4	1000	0.5593	-0.4076	0.9669
113	А	28	4	100	-0.3146	-0.7083	0.3937
114	D	46	1	200	0.1185	-0.6946	0.8131
115	В	42	4	10	-1.0741	-0.8371	-0.2370
116	А	28	8	40	-0.7529	-0.7422	-0.0107
117	А	28	2	1000	0.4669	1.3380	-0.8711
118	В	42	3	0.5	-2.1651	-0.7118	-1.4530
119	А	28	2	60	-0.2944	-0.6196	0.3252
120	А	28	5	30	-0.7917	-0.8518	0.0601
121	А	28	3	20	-0.9597	-0.9051	-0.0546
122	В	45	4	200	0.3462	-0.3852	0.7314
123	А	28	4	20	-0.8519	-0.5176	-0.3343
124	А	28	2	60	-0.3759	-0.8269	0.4510
125	А	28	1	100	-0.3269	-0.9612	0.6343
126	А	28	10	100	-0.3245	-0.1145	-0.2101
127	А	28	3	1000	0.7954	0.5063	0.2891
128	В	52	5	100	-0.4316	-0.3753	-0.0563
129	В	48	4	2	-1.9059	-1.1520	-0.7539
130	А	28	1	300	0.3902	-0.7504	1.1410
131	А	28	3	60	-0.5234	-0.8452	0.3218
132	А	28	1	1.25	-2.0108	-1.2200	-0.7908
133	А	28	5	1000	0.6728	-0.7426	1.4150
134	D	53	3	100	-0.4284	0.4393	-0.8677

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135	A	28	1	100	-0.1340	-0.3839	0.2499
136	В	47	4	300	-0.1527	-0.1095	-0.0432
137	А	28	3	1000	0.2072	0.1095	0.0977
138	А	28	5	1000	0.7118	0.0069	0.7049
139	В	42	4	10	-1.4337	-0.4655	-0.9682
140	В	53	5	30	-0.9799	-0.1669	-0.8130
141	В	49	8	30	-0.8203	-0.4723	-0.3480
142	В	42	1	40	-0.3230	-0.8084	0.4853
143	В	54	3	1000	0.1932	-0.0117	0.2049
144	В	42	3	40	-0.1696	-0.7706	0.6010
145	В	47	4	4	-1.4704	-0.6591	-0.8114
146	В	46	2	2	-1.7958	-0.8781	-0.9178
147	А	28	6	12	-1.3661	-1.2070	-0.1591
148	А	28	5	30	-0.7490	-0.7882	0.0392
149	В	42	3	300	0.3974	-0.2626	0.6600
150	В	52	5	200	0.1734	-0.2226	0.3960
151	А	28	2	40	-0.8610	-0.7244	-0.1366
152	В	45	4	6	-1.0736	-0.6448	-0.4288
153	А	28	2	6	-1.7605	-0.9502	-0.8103
154	А	28	3	4	-1.8267	-1.3550	-0.4717
155	А	28	4	1000	0.2645	-0.8891	1.1540
156	D	46	6	10	-1.3984	-0.3999	-0.9985
157	А	28	5	60	-0.6538	-0.8719	0.2181
158	А	28	1	10	-1.5544	-0.3625	-1.1920
159	В	49	8	200	-0.0478	0.4105	-0.4583
160	А	28	1	15	-1.1861	-0.6754	-0.5107
161	А	28	6	160	-0.0029	-0.6711	0.6682
162	А	28	6	12	-1.1090	-0.9341	-0.1749
163	В	49	8	300	0.0601	0.0068	0.0533
164	В	53	8	1000	0.6122	0.1407	0.4715
165	А	28	1	100	-0.5803	0.4899	-1.0700
166	В	46	1	500	0.5313	-0.0930	0.6242
167	А	28	5	25	-0.8721	-0.7576	-0.1145
168	А	28	7	1000	0.6211	0.9887	-0.3675
169	В	55	2	10	-1.0834	-1.0710	-0.0124

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170	А	28	4	140	-0.5641	-0.8672	0.3031
171	В	42	3	10	-1.0651	-0.4839	-0.5812
172	А	28	4	20	-0.8757	-0.9778	0.1021
173	А	28	3	20	-0.9325	-0.6086	-0.3239
174	А	28	2	300	0.2048	-0.1006	0.3054
175	А	28	5	1000	0.6543	0.1803	0.4740
176	В	42	3	12.5	-1.1186	-1.1500	0.0314
177	В	46	11	100	-0.2700	-0.0505	-0.2196
178	А	28	4	4	-1.7580	-0.6247	-1.1330
179	В	42	3	30	-1.2607	-0.3163	-0.9444
180	В	46	6	300	0.0720	0.0961	-0.0240
181	D	49	8	20	-0.9823	-0.9093	-0.0730
182	В	42	3	12	-1.0975	-0.8043	-0.2933
183	В	43	4	8	-1.2736	-1.1300	-0.1436
184	В	46	6	10	-1.0868	-0.6614	-0.4254
185	А	28	9	30	-0.7341	-0.8531	0.1191
186	В	46	4	1	-2.1077	-1.2200	-0.8877
187	А	28	1	40	-0.5386	-0.7963	0.2577
188	В	49	7	4	-1.6702	-0.5792	-1.0910
189	А	28	6	8	-1.6772	-0.7243	-0.9529
190	В	49	8	300	0.3182	-0.2376	0.5559
191	А	28	2	1000	0.5796	0.2196	0.3600
192	В	52	3	8	-1.3669	-0.6423	-0.7246
193	В	45	2	25	-0.8836	-0.7584	-0.1252
194	D	52	8	100	-0.4540	-0.6774	0.2234
195	А	28	2	80	-0.5192	-0.7321	0.2129
196	А	28	3	20	-0.8663	-0.8804	0.0141
197	А	28	4	40	-0.5070	-0.7730	0.2660
198	А	28	8	100	-0.0799	-1.0400	0.9601
199	А	28	5	10	-1.0835	-0.9296	-0.1539
200	А	28	1	4	-1.3381	-0.6290	-0.7091
201	А	28	4	1	-2.0093	-0.6680	-1.3410
202	А	28	6	15	-1.3784	-1.0860	-0.2925
203	А	28	7	20	-1.0135	-1.0770	0.0635
204	А	28	7	7.5	-1.5039	-1.0970	-0.4069

205	А	28	5	30	-0.7486	-0.6532	-0.0954
206	В	45	2	30	-0.6758	-0.4727	-0.2031
207	А	28	2	40	-0.4071	-0.5411	0.1340
208	В	49	7	20	-0.8636	-1.0490	0.1854
209	А	28	2	6	-1.4183	-0.5919	-0.8264
210	А	28	8	1.5	-1.9949	-1.1340	-0.8609
211	В	42	3	60	-0.3986	-0.9103	0.5117
212	В	45	4	40	-0.4705	-0.7944	0.3239
213	В	45	2	100	-0.1340	-0.4523	0.3183
214	D	42	3	5	-1.4413	-0.6481	-0.7932
215	А	28	3	100	-0.1768	-0.7368	0.5600
216	В	48	5	6	-1.3529	-1.0310	-0.3219
217	А	28	3	300	0.3431	-0.4523	0.7954
218	В	42	3	1000	0.8597	-0.3239	1.1840

^a (Study)

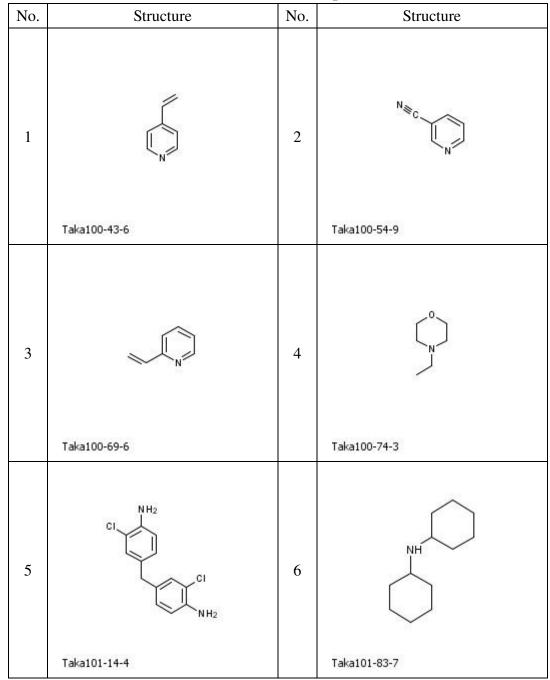
A: 28-day repeated oral dose toxicity study

B: Repeated dose and reproductive/developmental toxicity study

- C: One-generation reproduction toxicity
- D: Simple oral administration reproductive toxicity study

^b (Institution)

- 1: Safety Research Institute for Chemical Compounds Co., Ltd.
- 2: Research Institute for Animal Science in Biochemistry and Toxicology
- 3: Food and Drug Safety center, Hatano Research Institute, Japan
- 4: Mitsubishi Chemical Medience Corporation
- 5: Public Interest Incorporated Foundation BioSafety Research Center
- 6: Bozo Research Center
- 7: Panapharm Laboratories Co., Ltd.
- 8: Nihon Bioresearch Inc.
- 9: Drug Safety Testing Center Co., Ltd.
- 10: NOTOS B.V. 5231 DD' sHertogenbosch
- 11: Japan Bioassay Research Center
- 12: National Institute of Health Science



III. Chemical Structures of the 218 Chemical Compounds

