Using Various Nonlinear Response Surfaces for Mathematical Description of the Type of Combined Toxicity

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Abstract
The article considers the problem of characterizing the type of combined action produced by a mixture of toxic substances with the help of nonlinear response functions. Most attention is given to second-order models: the linear model with a cross-term and the quadratic model. General propositions are formulated based on the data on combined toxicity patterns previously obtained by the Ekaterinburg nanotoxicology team in animal experiments and analyzed with the help of the linear model with a cross-term. It is shown now that the quadratic model features these general characteristics in full measure, but interpretation of combined toxicity types based on isobolograms obtained by the quadratic model is more difficult. This suggests that where both models ensure a comparable quality of combined toxicity type identification, it would be enough to use the linear model with a cross-term.

Keywords
combined toxicity, response surface methodology, isobolograms, nonlinearity

Introduction
Today, the response surface methodology (RSM) is one of the most important general methods used in the analysis of combined effects produced by mixtures of bioactive substances, including toxic ones.¹-⁷ This method enables the potentialities of effective experimental design to be used for approximating a response function. Constructing such approximation requires choosing an analytical model whose parameters would be determined by fitting to experimental data.

The simplest way to approximate a response function is to construct a linear regression in relation to the doses of the toxic substances involved (hereinafter y will denote the response variable Y depending on the doses x₁ of the toxicants X₁, X₂, . . . , Xₙ)

\[ y = b₀ + \sum_{i=1}^{n} b_i x_i. \]

This model, however, is not adequate for identifying combined action types other than additivity. To allow for more complex types of combined toxicity, this model should include nonlinear terms or use explicitly nonlinear dependence models of a different analytical structure. At the same time, a model that is too complicated could overlook more essential patterns of change in the response with the dose.

The need for a trade-off between model complexity and interpretability has led to the recommendation that the model of choice for describing the nonlinear dependence of a response on influencing factors in the RSM should be one of the following two.¹-³

The first one is the linear model with cross-terms, or the main effects model with interaction, given by the equation

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\[ y = b_0 + \sum_{i=1}^{n} b_i x_i + \sum_{i \neq j} b_{ij} x_i x_j + \ldots + b_{12..n} x_1 x_2 \ldots x_n. \] (1)

In this model, any possible nonlinearity of the response is allowed for by the introduction of cross-terms, that is, the products of the first-order independent variables.

The second model is different from model (1) in that it has higher order independent variables and is called an nth order polynomial model. The right-hand part of the equation that defines this model contains an nth order general polynomial in the variables \( x_1, x_2, \ldots, x_n \)

\[ y = b_0 + \sum_{i=1}^{n} b_i x_i + \sum_{i \neq j} b_{ij} x_i x_j + \ldots + \sum_{i} b_i x_i^2. \] (2)

Both of these models are rather simple and, at the same time, sufficiently flexible tools for modeling nonlinear dose–response functions. In the general case, they may be used as a starting point for constructing the theoretical model of a response function. Note that each of these models includes as a special case the linear model, which, as noted above, expresses the additive nature of combined action.

Although these models may be defined for any number of influencing factors, it is practically possible to use them for describing the combined toxicity of 2 or 3 variables only. Indeed, functions with a large number of variables will contain, as a rule, products of variables with opposite signs. This would make it difficult to use such models for identifying combined action types. In such studies, researchers therefore tend to use models for 2 variables given by

\[ y = b_0 + b_1 x_1 + b_2 x_2 + b_{12} x_1 x_2 \] (3)

for the main effects model with interaction, and

\[ y = b_0 + b_1 x_1 + b_2 x_2 + b_{12} x_1 x_2 + b_{11} x_1^2 + b_{22} x_2^2 \] (4)

for the quadratic model (which may also be referred to as a “quadratic model with a cross-term”).

Since in the design of experiment theory, levels of factors are coded with integer numbers, using, for instance, the values 1 (controls), 0 (mean dose, if any), and +1 (maximum experimental dose of the toxic substance), then Equations 3 and 4 may be simplified when applied to certain types of experiment. For example, for an experiment with one 2-level factor and one 3-level factor, model (4) takes the form (assuming that the 3-level factor is coded with variable \( x_3 \))

\[ y = b_0 + b_1 x_1 + b_2 x_2 + b_{12} x_1 x_2 + b_{11} x_1^2. \]

For a \( 3^2 \) full factorial experiment, the quadratic model (4) is applicable in its full form (with both variables squared). For an experiment with two 2-level factors, Equation 4 is reduced to Equation 3. In general, model (3) is the only possible nonlinear model for a \( 2^2 \) experiment.\(^8,9\)

One can obtain from Equations 3 or 4 approximations to single-factor dose–response functions by fixing the values of one of the variables. For instance, if the value of one of the factors has been fixed (eg, \( x_2 = x_3^2 \)), Equation 3 for the dependence of the response \( y \) on the doses of the factor \( x_1 \) provides a linear equality (in relation to the dependence on \( x_1 \))

\[ y(x_1, x_3^2) = b_0 + b_2 x_3^2 + x_1 (b_1 + b_{12} x_2^2). \]

Doing the same for model (4) results in the quadratic dependence

\[ y(x_1, x_2^2) = b_0 + b_2 x_2^2 + b_{22} (x_2^2)^2 + x_1 (b_1 + b_{12} x_2^2) + b_{11} x_1^2. \]

Thus, in contrast to model (3), the presence of squared independent variables in model (4) leads to nonlinear approximations to single-factor dose–response functions.

Moreover, the presence of quadratic terms in model (4) can render this model essentially different from model (3) and give rise to additional difficulties for meaningful interpretation of combined toxicity analysis results obtained by means of this model. However, linear, quadratic, and more complex functions are broadly used in toxicological studies for approximating single-factor or multiple-factor dependence of the response on the doses of one or several toxicants.\(^10-17\) A promising approach to modeling of toxicological data is based on threshold models.\(^18\) Generally speaking, a single-factor dose–response dependence may be described by different models, and, as is stressed by the same authors,\(^18\) model selection is not a crucial issue. However, a multiple-factor dose–response function requires approximation by a function of the form (4) or a more complex one.

The uses of model (3) for 2 factors or the linear model with cross-terms (1) with \( n \) factors are sufficiently well presented in the literature. For instance, it is often used as the main model in the design of experiment for a mixture setting.\(^10-21\) Moreover, in experimental design employed for studying mixtures, the general polynomial model for \( n \) factors (2) is reduced to the canonical model (1) by virtue of the equality \( x_1 + x_2 + \ldots + x_n = 1 \), where \( x_1, \ldots, x_n \) in this case represent the fractional proportion of the components in the mixture.

In a series of publications,\(^22-36\) model (3) was used by us for analyzing the combined toxicities of various substances. It was shown that this model enables one to describe both the unidirectional action types (additivity, subadditivity, superadditivity) and oppositely directed actions of toxic agents in a certain region of experimental dose combinations. Model (3) was also used for studying combinations of other toxic substances.\(^37\) The properties of model (3) which are relevant to combined toxicity description were generally addressed in the studies.\(^8,9\)

The basic method for representing combined actions of toxic substances with the help of the response surface equation involves constructing response surfaces and sectioning them with constant-effect planes to obtain relevant lines. In toxicology, these lines (for 2 factors) or surfaces (for 3 and more factors) are called isoboles. The method of isoboles has been used for exploring joint action of toxic combinations in numerous studies carried out by other researchers as well.\(^3,37-51\)

The shape of an isobole makes it possible to visually represent and distinguish such unidirectional combined toxicity
types as additivity, subadditivity, and superadditivity (see Figure 1). The isobole for the additive action presents a straight line with regions of sub- and superadditive action on its sides.\textsuperscript{5,39,41-44} The distinctive feature of these lines is the presence of isoeffective doses for each toxicant, that is, each isobole crosses each of the coordinate axes at some point (with a positive coordinate). For the case of 2 variables, this means the presence of the points \((D_1, 0)\) and \((0, D_2)\), \(D_1 > 0\), \(D_2 > 0\) through which the isobole passes (see Figure 1).

However, even the application of such relatively simple functions as (3) and (4) often results in isoboles of a more complex shape. Below, we consider relevant examples. For the response function (3), the interpretation of such isoboles was considered in detail in the publications.\textsuperscript{8,9,22,24,29} For model (4), this issue remains largely debatable.

When analyzing experimental data with the help of the model function (3), we discovered\textsuperscript{22-36} some general principles, which were observed in all these experiments and which we believe to be representative of some important general patterns in combined toxicity, namely:

I. The combined toxicity type depends on the effect by which it is assessed.
II. The combined toxicity type depends on the choice of effect level.
III. The combined toxicity type depends on the dose ratio of the toxic agents in the combination.
IV. Identification of the combined toxicity type in a certain region of dose combinations may reveal an opposite action of the toxicants.

Thus, it would be incorrect to draw any conclusion about the type of combined action of toxicants without specifying the effect, its level, and the dose combination involved. Only where the isobologram displays lines of the same type over the entire experimental range of doses could the type of combined toxicity demonstrated by the toxicants in relation to a given effect be stated unambiguously without specifying the level of this effect and the region of dose combinations. However, it is much more frequent to encounter cases where the combined toxicity type depends on both. Below, we show that these statements are true of the quadratic model (4) as well.

**The Geometric Meaning of the Response Surfaces (3) and (4)**

In considering models (3) and (4), it is useful to represent them as equations describing a certain surface in the space of the variables \((x_1, x_2, y)\). Because both Equations 3 and 4 are second-order equations, they define second-order surfaces.\textsuperscript{52,53} If we disregard the degenerate cases of these surfaces, then Equation 3 will describe only one surface, that is, a hyperbolic paraboloid all sections of which with horizontal planes are hyperbolas with perpendicular asymptotes (Figure 2A). A more general model (4) defines 2 possible response surfaces—a general hyperbolic paraboloid with a random arrangement of the asymptotes created by its horizontal sections, or an elliptic paraboloid\textsuperscript{52,53} (Figure 2B).

Although they are very dissimilar, each of these surfaces can represent both the variants of unidirectional combined action (additivity, superadditivity, and subadditivity) and actions in opposite directions. At the same time, even where both model (3) and (4) define a hyperbolic paraboloid for some index, the resulting isoboles may be essentially different due to a more complex form of the response surface in model (4).

The distinctive features of models (3) and (4) may be seen if we consider the sections of these surfaces in Figure 3A and B, respectively, in the region containing their centerpoints (Figure 3). As can be seen from these figures, the presence of 2 asymptotes in models (3) and (4) in a case of a hyperbolic paraboloid results in an explicit division of the region of dose combinations into subregions, each of which features curves of a certain type. Changing from one region into another in an uninterrupted manner is impossible because the level lines cannot cross the asymptotes. However, the isobole for a given response level may consist of 2 lines located in different regions of dose combinations thus forming a composite curve.

If surface (4) presents an elliptic paraboloid, its sections considered in the region containing the center form closed ellipses. Interpretation of these ellipses in the context of combined toxicity type identification presents a considerable difficulty.
Thus, in general where we come across such complicated curves as full hyperbolas for the model of hyperbolic paraboloid (3) or (4) or full ellipses for the model of elliptic paraboloid (4), it is necessary to divide the region of dose combinations into subregions to enable an unambiguous interpretation of the combined action type to be made in each of them. This prominent feature of complex isoboles will be illustrated by the experiments that we have carried out so far.

Examples of Isoboles for the Quadratic Model (4)

The examples presented in this section are based on the previously published experiments briefly described below.

Experiment (E1)

The results of this experiment were reported in the studies by Katsnelson et al., Minigalieva et al., and Katsnelson et al. Stable suspensions of NiO and/or Mn3O4 nanoparticles (NPs) were administered to rats by intraperitoneal (IP) injection at a dose of 0.50 mg or 0.25 mg three times a week up to 18 injections, either separately or in different combinations. Thus, we obtained 9 groups of rats: (1) controls, and rats given (2) 0.50 mg NiO-NPs, or (3) 0.25 mg NiO-NPs, or (4) 0.50 mg Mn3O4-NPs or (5) 0.25 mg Mn3O4-NPs, or (6) 0.50 mg NiO-NPs + 0.50 mg Mn3O4-NPs, or (7) 0.25 mg NiO-NPs + 0.25 mg Mn3O4-NPs, or (8) 0.50 mg NiO-NPs + 0.25 mg Mn3O4-NPs, or (9) 0.25 mg NiO-NPs + 0.50 mg Mn3O4-NPs.
Experiment (E2)
The results of this experiment were reported in the study by Katsnelson et al. Sodium fluoride solution was administered to rats by IP injection at a dose equivalent to 0.1 median lethal dose (LD₅₀) 3 times a week up to a total of 18 injections. Two-thirds of these rats and of the sham-injected ones were exposed to the whole body impact of a 25-mT static magnetic field (SMF) for 2 or 4 hours a day (note 1), 5 times a week. Thus we had 6 groups of rats: (1) controls (IP injections of normal saline); (2) rats administered sodium fluoride IP injections; (3) rats exposed to the SMF for 4 hours per day; (4) rats exposed to the SMF for 2 hours per day; (5) rats exposed to the SMF for 4 hours per day and fluoride IP injections; (6) rats exposed to the SMF for 2 hours per day and fluoride IP injections.

Experiment (E3)
The results of this experiment were reported in the study by Privalova et al. Barium chromate (BaCrO₄) and manganese dioxide (MnO₂) fine powders were suspended in normal saline and instilled intratracheally into rat lungs separately or in combination at a dose of 5.0 or 2.5 mg in 1 mL of the suspension, while the control rats received 1 mL of normal saline only. Thus, we obtained 9 groups of rats according to each dose combination of the chemicals involved. Some cytological characteristics of a bronchoalveolar lavage fluid cell population of rats were assessed 24 hours after the intratracheal instillation.

Combinations of Harmful Factors Acting in the Same Direction
Cases of unidirectional action are well known and have long since been considered a basis for describing the combined action in any experiment. Models (3) and (4) also enable one to describe just such cases. If the approximations (3) and (4) fit the experimental data well, the isoboles for these models will be qualitatively similar, though somewhat different in shape.

Consider the examples of unidirectional action isoboles in Figures 4. Figure 4A demonstrates additivity in the combined action of SMF and NaF. Note that in experiment (E2), one factor (SMF) had 3 levels, including zero, while the other (NaF) only 2. The quadratic model (4) in this case was therefore reduced to a model with one-squared variable

\[ y = b_0 + b_1x_1 + b_2x_2 + b_{12}x_1x_2 + b_{11}x_1^2, \]

where \( x_1 \) represents SMF.

The use of nonlinear approximations, such as (3) or (4), for identifying the type of combined toxicity makes it necessary to estimate the degree to which resulting isoboles differ from a straight line. Figure 4A demonstrates just such a case. For Equation 3, a conclusion concerning the departure of the isobole from linearity may be drawn from the estimate of statistical significance of coefficient \( b_{12} \) for the product \( x_1x_2 \). For approximation (4), a corresponding formula will be more complicated, as it will include also coefficients at the variables squared. However, in the case of Figure 4A, it is clear without any calculations that the isoboles should be regarded as straight lines. Thus, here, we deal with one and the same type of combined toxicity in all the range of experimental doses, namely, additivity.

We can describe in the same way the other 2 graphs in Figure 4. Similar examples can be provided for the other experiments of ours to which model (4) could be applied. Note in these cases that the type of combined action is the same over the entire region of dose combinations considered. In the majority of our experiments, however, the type varied depending on the response level or on the region of dose combinations.

The characteristic feature of unidirectional action isoboles is that they present graphs of decreasing functions. In other words, for maintaining the level of response with increasing the dose of one toxicant, it is necessary to reduce the dose of the
other, that is, a decrease in the dose of one substance is compensated by an increase in the dose of the other. This compensation may be linear as in Figure 4A, or nonlinear as in Figure 4B and C, but in each of these cases the constancy of the effect is ensured by a balanced change in the doses of the substances—an increase in the dose of one of them should be compensated with a decrease in the dose of the other. Consequently, from the standpoint of isobole shape, a unidirectional action manifests itself as an isobole, which is a decreasing function graph. More specifically, if \( Y(x_1, x_2) \) is a response function, the isobole corresponding to effect \( Y_0 \) is the graph of an implicit decreasing function defined by equality \( Y(x_1, x_2) = Y_0 \).

However, in our experiments, we often came across situations where we failed to describe the behavior of the isoboles over the entire range of dose combinations in one way as, for example, in Figure 4. Moreover, we came across isoboles that represented increasing functions and, hence, did not describe the unidirectional action of the agents involved.

**Combinations of Harmful Factors Acting in Opposite Directions**

The concept of oppositely directed actions of toxicants is insufficiently considered in toxicological studies. However, in the experiments of our team, this type of combined toxicity occurred rather frequently. Its distinctive feature is that for maintaining the level of effect while increasing the dose of one toxicant, it is necessary to increase the dose of the other as well. Consequently, for an action in opposite directions, the isobole-defining function \( Y(x_1, x_2) = Y_0 \) will be an increasing one.

Note that the majority of effects feature action in opposite directions only in a certain subregion of dose combinations although some effects may display it over the entire region of experimental dose combinations as well. Although oppositely directed action is a special type of combined toxicity, it could be considered together with subadditivity. Indeed, in subadditivity, the substances mutually attenuate each other, and so in order to achieve the same level of effect as in additivity, it is necessary to increase the dose of at least one of the toxicants. The oppositely directed action of the toxicants is even more antagonistic in this respect. The oppositely directed action may therefore be called “explicit antagonism” unlike subadditivity, which may be called “implicit (hidden) antagonism.” Besides, when one analyzes the type of combined action by the shape and slope of the isobole, the segments of subadditive and oppositely directed action are often observed to be connected on the same curve.

In our experiments, we saw various cases of oppositely directed action, which showed themselves as ascending curves of different shapes. Nevertheless, we do not think it would be reasonable to additionally subdivide such rising isobole curves into various types.

Figure 5 shows isoboles for an oppositely directed action of 2 substances. In Figure 5A, an oppositely directed action of fluoride and SMF is seen over the entire region of experimental dose combinations. In Figure 5B, an oppositely directed action of Mn3O4 and NiO NPs in one region of dose combinations combines with canonical cases of unidirectional action (subadditivity and superadditivity, respectively) in the region between the asymptotes of the hyperbolas. Note also that the subadditivity in the region of low and medium doses of Mn3O4 and NiO transforms into an opposite action at higher doses of either of the substances. A similar picture can be seen in Figure 5C, but in this case, higher doses of both toxicants (MnO2 and BaCrO2 microparticles) display a superadditive combined...
action. In both cases in Figure 5B and C, the entire region of experimental dose combinations is divided into subregions, in which the type of combined toxicity is maintained the same. In both cases, this subdivision is affected by the asymptotes of the set of hyperbolas forming the isobologram.

**Composite Isobolograms for the Quadratic Model (4)**

The division into subregions of one type of combined action depends essentially on the theoretical model used for representing the dependence of the response on the doses of the impacting factors. If several theoretical models are considered and each of these models fits experimental data well, the subregions of dose combinations with various types of combined toxicity predicted by such models will be qualitatively identical.

Let us consider some more examples of isobolograms in which the region of dose combinations is divided into subregions with certain types of combined action. If model (3) is used, the region is divided by the mutually perpendicular asymptotes of the hyperbola so that each subregion features only one type of combined action. For model (4), as we saw above, one of the possible response surfaces is also a hyperbolic paraboloid whose sections will be hyperbolas. The asymptotes present in the region of experimental dose combinations in this case also divide the region into segments, which should be analyzed separately. However, because the asymptotes of the hyperbolas in model (4) may intersect at any angle, the interpretation of the isoboles in this case will be more complicated. In particular, segments with different types of combined toxicity may occur together on the same line.

Figure 6 shows isobolograms for which the division of the region of dose combinations goes along the asymptotes of the hyperbolas resulting from the sectioning of surface (4) with constant-effect planes. In each subregion, the isoboles are arranged in about a similar pattern so that one such subregion displays combined toxicity of one certain type. For example, Figure 6A displays subadditivity at low levels of effect (the corresponding isoboles are located in the bottom left corner of Figure 6A). Proceeding into other regions of dose combinations (located between the asymptotes), we see isoboles of a completely different type. The interpretation of such isoboles depends on effect level. For example, for the isoboles that are close to the asymptotes of the hyperbolas (eg, for effect level 64,5), the combined toxicity is described as additive (at low doses of NiO), then changing into an oppositely directed action (at high doses of NiO). In general, it is obvious that such isobolograms cannot be unambiguously interpreted from the standpoint of combined toxicity type assessment.

**Conclusions**

Identifying the type of combined action in multiple factor toxicological studies presents a difficult problem. The reason is that the actual behavior of the response to a combined action of toxic agents may be ambiguous within the experimental region of dose combinations. Such behavior cannot be attributed to a particular model as its inherent feature. Rather, it is characteristic of the varied response of biological systems to a multifactorial impact in various dose combinations of the participating agents and the dependence of the combined action type on response level. Allowing for these circumstances leads to a more complex description of the type of combined toxicity, which, in particular, requires dividing the region of dose combinations into segments in which the type of combined action could be interpreted unambiguously. Such subdivision depends largely on the theoretical response model used, although it does not rule out a qualitatively similar picture of combined action as revealed by various models.

Moreover, the choice of a model for theoretical representation of the dependence of the response on the doses of the participating factors defines the shape of the line of constant response level (isobole) which may appear rather complicated...
even for a relatively simple response function model. For example, an isobole for one and the same level of response may appear consisting of the several segments located in different subregions of dose combinations with essentially different shapes. In other words, for the given level of response, the combined action is of different types depending on which region of dose combinations of the toxicants is considered.

Moreover, apart from the well-known classical types of unidirectional action isoboles representing additivity, subadditivity, and superadditivity, experiments may reveal isoboles whose shapes are considerably different from these types. Namely, the resulting complicated or composite isoboles may be explained by an oppositely directed action of the substances in some region of dose combinations. We define as oppositely directed combined action which requires increasing the doses of both toxicants for maintaining constant response. Such type of combined action can be called explicit antagonism. Depending on which effect is chosen for the analysis of combined toxicity, a given pair of substances may display various types of the latter including an oppositely directed one. However, most often an opposite action appears in only a certain subregion of the region of experimental dose combinations.

This variety of cases occur when one analyzes experimental data using even such simple models as the linear model with a cross-term (3) or the quadratic model (4). The basic difference of model (4) from model (3) consists in the presence of the squares of the independent variables in the former, the consequence of which is the possibility of a quadratic approximation for single-factor dose–response functions compared to linear functions for model (3). The latter may be applied as the basic one in a case where nonlinearity should be taken into account explicitly with no additional information on the shape of this nonlinearity being available. As is stated in “Experimental Design for Formulation”: “At the beginning of an investigation, and without prior subject-matter knowledge, one would have no idea about the functional relationship between the response and the mixture variables. What is often done at this point is to hypothesize a linear polynomial model, sometimes called a screening model.”

When constructing a model for experimental data, the researcher has a wide choice of possible analytical representations for the response function and so should choose one, representing a compromise between the complexity of approximation and the possibility of interpreting the findings, which is made possible with simple model functions. As such a compromise, the RSM theory recommends models (3) or (4). It is obvious that from the standpoint of combined toxicity type assessment, model (3) has the advantage that the shape of respective isoboles allows one to obtain a more simple and clear-cut interpretation of the combined action type. We maintain that in relation to the problem of combined toxicity type identification, the most important thing is to have basic qualitative characteristics, and these can be well obtained from the simpler model (3), provided it fits the available experimental data as well as model (4).

Thus, the analysis of toxicological experimental data for identifying the type of combined toxicity should start with the use of such relatively simple models as (3) and (4). There are standard tools for statistical verification of the quality of each of these models, which is a prerequisite for making an appropriate choice. For such a choice, it is useful to construct isobolograms for a given index (effect) in the region of experimental dose combinations. The shapes of these isoboles may be both classical (Figure 4) or more complicated (eg, Figures 5 and 6). A meaningful interpretation of complex isoboles is only possible by allowing for, firstly, the dependence of the type of combined toxicity on the chosen toxic effect; secondly, the chosen level of the effect; and thirdly, the ratio of the combined factors doses. It is essential for a correct interpretation of the isoboles to allow for the likelihood of an oppositely directed action of the toxic agents, which may be present together with any of the classical types of combined toxicity for the same level of effect. The quality of models (3) and (4) being comparable, preference should be given to model (3) because the interpretation of isoboles with this model is easier, and the qualitative picture of combined toxicity will be identical for both models.

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Note
1. The exposure unit time was used as a substitute for SMF dose.

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