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Full-scale modelling of an ozone reactor for drinking water treatment

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ABSTRACT

In 2003, the Flemish Water Supply Company (VMW) extended its drinking water production site in Kluizen (near Ghent, Belgium) with a combined ozonation and biological granular activated carbon (BGAC) filtration process. Due to this upgrade, biostability increased, less chlorination was needed and drinking water quality improved significantly. The aim of this study was to describe the full-scale reactor with a limited set of equations. In order to describe the ozonation process, a model including key processes such as ozone decomposition, organic carbon removal, disinfection and bromate formation was developed. Kinetics were implemented in WEST[®] and simulation results were compared to real data. The predicting performance was verified with a goodness-of-fit test and key parameters were determined through a local sensitivity analysis. Parameters involving optical density (both rate constants and sto-ichiometric coefficients) strongly affect model output. Some parameters with respect to bromate and bacteria showed to be only, but to a large extent, sensitive to their associated concentrations. It was demonstrated that the model is able to describe the operation of the full-scale ozone reactor, however, further data collection for model validation is necessary.

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1. Introduction

In order to produce high quality drinking water from surface water resources, a combination of physical and chemical treatment steps is typically used. To achieve good bacteriological quality, chlorine is often used as oxidising agent. However, it is well established that chlorine can lead to many problems in the aquatic environment due to the formation of potential toxic organochlorine compounds such as tetrachloroethene, trichloroethene and halo-acetic acids [1]. Besides this, emerging pollutants and endocrine disruptors (EDCs) became important contaminants in water systems during the last few decades. In order to minimize this by-product formation and to remove harmful compounds, advanced oxidation processes (AOPs) have already proven to be effective technologies [2,3]. Besides the potential benefits in drinking water production, AOPs have a large potential for the treatment of different types of water and waste streams originating from waste water, such as domestic and industrial effluents, sludge and membrane con-

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centrates, swimming pool water and process water. Both low and high concentrated flows can be treated with AOP techniques [4,5]. Most techniques are based on the formation of hydroxyl radicals which are the strongest oxidators that can be used in water treatment systems. Hydroxyl radicals can be generated in water through different combinations of oxidants, like ozone and hydrogen peroxide, or by combining a single oxidant with UV radiation [6]. AOP techniques such as ozonation are either used for (1) the complete or partial oxidation of the organic contamination, (2) the oxidation of a specific contaminant or (3) the removal of pathogens. However, it should be noted, that ozone also produces disinfection by-products (DBP) such as bromate which is formed out of bromide and considered to be a potential human carcinogen [7]. Ozone is used for the removal of MIB (2-methylisoborneol) in drinking water production installations as this component is responsible for odour and taste problems [8]. The stage 1 and stage 2 rules, promulgated by US EPA (Environmental Protection Agency), defined a MCL (maximum contaminant level) of $10 \,\mu g l^{-1}$ bromate for systems using ozone [9]. In Europe, a similar regulatory level is applied since 1998 (98/83/EC). If the incoming water contains high bromide levels, bromate formation can be restricted by, e.g. adding hydrogen peroxide (H₂O₂) to the ozonation process, [2]. In this case, hydroxyl radicals become important players.

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Despite the many advantages and added value of the AOP techniques, there still exist several bottlenecks and research questions concerning these techniques. First, scale-up of lab-scale research reactors to full-scale industrial reactors is often failing. Second, further research on process control and optimization is necessary. Third, the removal of organic components and micro-organisms is not yet completely unravelled. Modelling of AOP processes offers an elegant and cost-effective tool to tackle these research questions.

Many different attempts have already been made to describe ozone decomposition with or without the presence of organic compounds [10,11]. Two general accepted deterministic models for ozone decomposition in "pure water" have been developed in the early 1980s, both based on the first model of Weiss [12]. The model of Staehelin, Hoigné and Bühler, known as the SHB model, was experimentally developed at acidic to neutral pHs [13,14], while Tomiyasu, Fukutomi and Gordon (TFG) developed their model at high pH values [15]. A comparison of both, together with some simulation results can be found in Ref. [16]. On the other hand, numerous empirical and semi-empirical studies describing ozone decomposition [17,18], reactions with organic compounds [16,18], by-product formation [19] and disinfection [20] were conducted in the last decades. Lovato et al. [17] extended the SHB model with an empirical approach by relating one of the 18 kinetic constants to the solution pH. van der Helm described ozone decomposition and organic compound removal by using UV absorbance at 254 nm (UVA₂₅₄) as surrogate for the NOM concentration [18]. Sohn et al. [19] developed an empirical relation to predict bromate formation related to several operational and water quality parameters.

In this contribution a simplified model is presented for the simulation of the full-scale ozone reactor of the Flemish Water Supply Company (VMW) in Kluizen (Belgium). The aim of this study is to describe this process with a limited set of equations, to determine the key parameters and to perform different scenario analysis. The results will be used as a starting point for further model development, especially in terms of extending the mechanisms of organic compound removal and biodegradability enhancement. As such, the model might contribute to answering the research questions stated above.

The $60,000 \text{ m}^3 \text{ day}^{-1}$ water treatment plant (WTP) which is the subject of this study is fed with raw water captured from lowlands. Until 2003, the treatment concept consisted of micro-sieving, enhanced coagulation followed by sludge blanket clarification, oxidation with chlorine, sand filtration and granular activated carbon filtration (GAC). A final disinfection with chlorine was applied [21,22]. The intensive chlorination resulted in high trihalomethanes (THM) levels and prevented biofilm growth on the activated carbon granules. Due to the absence of biological activity and limited contact time in the GAC (15 min), NOM removal was mainly by adsorption, which is limited to the first 10,000 bed volumes (BV, a unit that expresses the volume of water that already passed through the filter as a multiple of the volume of the filter bed). In 2000 the earthy-musty taste and odour compound MIB appeared in the feed water with concentrations above the odour threshold value of 10 ng l⁻¹, caused by algal growth in the reservoirs. This, along with a lack of biostability of the water due to high TOC levels, forced the drinking water company to search for an effective oxidizing technique. An ozone production and mixing unit was introduced, together with biological granular activated carbon (BGAC) filters. Ozone was implemented for both disinfection and oxidation. Due to the excellent disinfection capacities, a first chlorination step could be omitted. On the other hand, ozone was also implemented to enhance biodegradability in favour of the biofilm present in the BGAC to remove MIB and NOM. Only a final chlorination step remained [21,22].



Fig. 1. Schematic overview of the ozone production and mixing unit.

2. Materials and methods

2.1. Ozone reactor

Ozone is produced from oxygen with two Wedeco EFFIZON® ozone generation units, each with a production capacity of $4000 \text{ g} \text{ h}^{-1}$ (180 g ozone per Nm³ oxygen/ozone mixture). The ozone generating elements consist of discrete borosilicate glass tubes with a diameter less than 11 mm. A part of the main water stream is pumped up and pressured up with a high pressure booster pump. The water is brought into a venture injector where ozonated gas is introduced into the water. The side stream is then re-introduced to the main water pipe prior to a static mixer. The normally applied ozone dose in the water is $2.5 \text{ mg} \text{ l}^{-1}$. A schematic overview of the gas transfer process is given in Fig. 1. A sampling point is located after the static mixer (before the activated carbon filters). After the mixing is completed, the water proceeds to the BGAC filters. Gas flow as well as ozone concentrations in the gas are continuously monitored in order to evaluate the ozone transfer efficiency. An off-gas ozone destruction system converts any residual or non-dissolved ozone to oxygen so that the ozone concentration in the treated off-gas is lower than 0.1 ppm [21,22].

2.2. Biological granular activated carbon (BGAC) filters

The BGAC consists of ten pressure filters with a diameter of 6 m and operational pressure of 1.5–2 bar. GAC is operated as a twostage filtration, the first filter stage operates between 25,000 and 50,000 BV. After 50,000 BV (two years), the carbon is reactivated and the filter is switched to the second stage position, which operates between 0 and 25,000 BV. A contact time of 6 min is obtained in the filters, above the carbon bed. Non-dissolved gas is collected in an upward tee above each GAC filter and led away to the ozone destruction system [21,22].

2.3. Modelling approach

The ozone reactor of the Flemish Water Supply Company was implemented in the modelling and simulation platform WEST[®] (MostforWater, Belgium) as two continuous stirred tank reactors (CSTRs) in series. In the first tank with a volume of 10.68 m³ the ozone is introduced. This tank represents the part of the main water pipe before and with the static mixer. A second reactor represents the water on top of the activated carbon filters where a second reac-



Fig. 2. Implementation of the ozone reactor in the simulation platform WEST[®].

tion phase takes place. The simulation configuration as used in the software program is given in Fig. 2.

Four ozone reactions were implemented in WEST[®] according to [7,11,20,23]:

- Ozone decomposition.
- Reaction of optical density (OD) with ozone (organic carbon oxidation).
- Disinfection.
- Bromide oxidation (bromate formation).

Ozone reactions in water can be classified as either direct or indirect [10,23]. E.g., the disinfection reaction rate can be described by the following equation, where molecular ozone as well as hydroxyl radicals contribute to the oxidizing capacity of the system:

$$\rho = (k_{O_3} + k_{OH}R_c)[O_3][X_{\text{bact}}] \text{ with } R_c = \frac{[HO^\bullet]}{[O_3]} \approx 10^{-8} \text{ to } 10^{-6}$$
(1)

with ρ : reaction rate, in this example the inactivation rate of bacteria (CFU m⁻³ s⁻¹); k_{O_3} : second order rate constant for the direct reaction of molecular ozone with a specific compound, in this example micro-organisms (m³ g⁻¹ s⁻¹); k_{OH} : second order rate constant for the indirect ozone reaction pathway of hydroxyl radicals with micro-organisms (m³ g⁻¹ s⁻¹); k_c : the ratio of the concentrations of hydroxyl radicals and ozone; [O₃]: the concentration of ozone in solution (g m⁻³); [X_{bact}]: the density of viable micro-organisms, in this model expressed as colony-forming units per liter (CFU m⁻³).

When R_c is assumed to be constant and very little, both direct and indirect reactions can be lumped into one:

$$\rho = k_{O_3}[O_3][X_{\text{bact}}] \tag{2}$$

However, if the indirect mechanism plays an important role in the oxidation of some calculated species, predictive capabilities of the model will deteriorate because process efficiency in that case highly depends on other process conditions such as scavenger concentrations that are not included in the model. Probably radical reactions indeed occur in the waterworks because MIB was removed after ozonation, while rate constants with regard to the direct and indirect pathway are <10 and $3 \times 10^9 \,\text{M}^{-1} \,\text{s}^{-1}$, respectively [23].

Direct ozone decomposition is modelled assuming that ozone follows a first order decay with a rate constant of $0.000485 \, \text{s}^{-1}$ [11]. UV absorbance (optical density) at 254 nm was used as a surrogate for the amount of organic material that reacts with ozone [11,24]. However, it has to be highlighted that this parameter repre-

sents a part of the organic pollution concentration as it specifically gives a measure of the amount of aromatic and unsaturated compounds in water [25]. On the other hand, this parameter offers great opportunities for modelling and control as it can be determined on-line and consequently a huge amount of real-time and accurate data are available. Accordingly, OD might be a useful parameter in model-based control of WTPs.

Model calibration was performed with historical data of the year 2008 over a period of 300 days. For $k_{\rm OD}$, which is the rate constant for the reaction of UVA₂₅₄ with ozone, an initial value in the range of 0.1 m³ g⁻¹ s⁻¹ was used [11], while after calibration through parameter estimation a value of 0.0135 m³ g⁻¹ s⁻¹ was found. The stoichiometry or yield (Y) of the reaction is represented by $Y_{O_3/OD}$, with a numerical value of 0.22 [11]. This implies that one unit of OD (m⁻¹) consumes 0.22 g of ozone.

Disinfection kinetics were adapted from the classic Chick–Watson model [20] (and references therein):

$$\frac{d[X_{\text{bact}}]}{dt} = -k_{\text{X}}[O_3][X_{\text{bact}}]$$
(3)

For k_X , the inactivation rate constant for a particular microorganism (m³ g⁻¹ s⁻¹), a value of 0.6 m³ g⁻¹ s⁻¹ was applied after calibration, while the initial value was 1.72 m³ g⁻¹ s⁻¹ [11]. The stoichiometry of the reaction is represented by $Y_{O_3/X}$, with a numerical value of 1.29×10^{-14} [20].

Bromide oxidation was incorporated as bromate is an important by-product of ozonation in bromide containing waters. Bromide was assumed to be directly oxidized to bromate, although in reality ozone first oxidizes bromide to form hypobromous acid and hypobromite. The latter is further oxidized to bromite which finally forms bromate [23,26,27]. As such, one rate constant for bromate formation was determined after calibration and the stoichiometric coefficients from this process were derived from a reaction where 1 mole of bromate is formed out of 1 mole of both ozone and bromide. The rate constant was found to be 0,00043 m³ g⁻¹ s⁻¹.

The kinetics and stoichiometric coefficients used in the model are presented in Table 1 as a Petersen matrix. This matrix presentation offers a clear overview of the chemical reaction mechanisms included in the model. Reaction rates are indicated in the right column. Matrix elements are stoichiometry parameters.

The temperature and pH influence was not accounted for in this study as these parameters remain almost constant during the waterworks daily operation. Although, for example the inactivation constant for bacteria, k_X , is temperature dependent [20]. With respect to pH, this parameter has a slight effect on the ozone decomposition rate when lower than 7, but at higher values, the rate

Table 1

Petersen matrix representing the model used in this study.

Process	$O_3 (g m^{-3})$	$OD(m^{-1}m^{-3})$	X_{bact} (CFU m ⁻³)	$Br^{-}(g m^{-3})$	$BrO_3 (g m^{-3})$	Reaction rate
Ozone decomposition Reaction of optical density with ozone Disinfection Bromide oxidation	-1^{a} $-Y_{O_{3}/OD}^{b}$ $-Y_{O_{3}/X}^{c}$ -1	-1	-1	-1.66	2.66	$k_{O_3}[O_3]$ $k_{OD}[O_3][OD]$ $k_X[O_3][X_{bact}]$ $k_{Br}[O_3][Br^-]$
a [11.00]						

^a [11,23].

^b [11]. ^c [20]. increases significantly [10]. For instance, von Gunten and Hoigné showed that the half-life of ozone is 10 times higher at pH 10 than that at pH 11 [27]. These researchers also reported that less bromate is formed when lowering the pH. Further, no gas transfer equations were included in the model. A dissolved ozone concentration in the influent of the first tank was defined at the beginning of each simulation run. Finally, the activated carbon present in the second compartment could have a catalytic effect on ozone decomposition. However, this was not considered.

2.4. Data interpretation

The goodness-of-fit between experimental and simulated values was quantified by calculating Theil's inequality coefficient (TIC) [28], which is expressed as follows:

$$TIC = \frac{\sqrt{\sum_{i} (y_{i} - y_{m,i})^{2}}}{\sqrt{\sum_{i} y_{i}^{2}} + \sqrt{\sum_{i} y_{m,i}^{2}}}$$
(4)

where y_i represents the simulated data points; $y_{m,i}$ representing the measured data points.

A value of the TIC lower than 0.3 indicates a good agreement with measured data [29].

A sensitivity analysis was performed in WEST[®] to determine the most important model parameters (those parameters that have a major influence on the model output). The relative sensitivity function (RSF) was adopted to evaluate the sensitivity of the model output (concentration of ozone, OD, CFU and bromate) to a change of model parameters (rate constants k_{O_3} , k_{OD} , k_X and k_{Br} and the stoichiomtric coefficients $Y_{O_3/OD}$ and $Y_{O_3/X}$).

The RSF was calculated out of the sensitivity function (SF) by the finite forward difference method with a perturbation factor of 0.1% [30]. This means that SFs were calculated by raising the nominal parameter value with 0.1% as shown in following equation:

$$\frac{\partial y}{\partial \theta_{j_+}} = \frac{y(t, \theta_j + \xi \theta_j) - y(t, \theta_j)}{\xi \theta_j}$$
(5)

where $y(t, \theta_j)$ represents the output variable; θ_j represents the nominal parameter value; ξ is the perturbation factor.

RSF was calculated as follows:

$$RSF = \frac{SF \times \theta}{y(t,\theta)}$$
(6)

A RSF less than 0.25 indicates that the parameter is not influential. Parameters are moderately influential when 0.25 < RSF < 1. When 1 < RSF < 2 and RSF > 2, the parameter seems to be very and extremely influential, respectively [31].

2.5. Analysis

Ozone was measured spectrophotometrically with the indigo reagent method at 600 nm [32]. Optical density was constantly measured with a process integrated UV spectrometer at 254 nm. Bromide, bromate and total CFU were analysed according to standard methods [33]. All analyses were performed by the Flemish Water Supply Company.

3. Results and discussion

3.1. Modelling results

Model calibration was performed with historical data of the year 2008 over a period of 300 days. The influent flow rate and optical density are represented in Fig. 3. The influent bacteria and bromide concentration were on average $183 \, \text{CFU} \, \text{I}^{-1}$ and $138 \, \mu g \, \text{I}^{-1}$,



Fig. 3. Influent flow rate and optical density.

respectively. The descending trend of the OD can be attributed to improved settling and flotation performance in the pre-treatment steps during that period.

As mentioned before, all samples were withdrawn before the activated carbon bed (after the static mixer). Consequently, all simulation results are representing the effluent of the first reactor in Fig. 2. Fig. 4 shows the measured and calculated optical density after the first reactor. A good agreement was obtained as the TIC for this parameter is calculated to be 0.044 (<0.3). The deviation of the applied rate constant for k_{OD} in comparison with literature reported values can be explained by differences in organic carbon content of the water.

On average, 25% of the OD was removed for both calculated and modelled values. Concerning the number of CFU, an average log removal of 1.2 was calculated, which is in accordance with the measured removal (1.1 log). Measured influent bacteria concentrations vary between 80 and 200 CFU l⁻¹, while those of the effluent are in the range of 10 CFU l⁻¹. Calculated and measured bacteria removals are compared in Fig. 5. As can be seen, calculated values agree well with experimental ones. This is confirmed with a calculated TIC of 0.084, although it has to be highlighted that more data points have to be collected in future studies.

Bromate was only measured once $(4 \ \mu g l^{-1})$. Therefore, the average calculated bromate concentration and the measured value are presented in Fig. 6. Based on the measured value, it can be seen that model predictions are realistic. In future work, the validation process will be repeated, especially for bromate formation. Measured and calculated values are far below the regulatory level of 10 $\mu g l^{-1}$.

3.2. Sensitivity analysis

The initial parameter values used in the sensitivity analysis are shown in previous paragraphs. It can be clearly noticed that parameters involving OD (k_{OD} and $Y_{O_3/OD}$) strongly affect model output



Fig. 4. Comparison of measured and calculated optical density.



Fig. 5. Comparison of measured and calculated logarithmic bacteria removals.



Fig. 6. Predicted and measured bromate concentrations in the effluent.

compared to others (Table 2). k_{OD} has a moderate effect on calculated ozone, bacteria and bromate concentrations. There is a smaller influence with respect to OD itself. The same conclusions can be made for $Y_{O_3/OD}$, although moderate effects are slightly higher. k_X and k_{Br} only influence their associated concentrations [X] and [BrO₃]. However, due to the very affecting character (RSF \approx 1), accurate estimation regarding these parameters will be necessary to obtain realistic predictions of bacteria and bromate levels. k_{O_3} and $Y_{O_3/X}$ do not exert an influence on simulation output. Bacteria (X) form part of OCS (ozone consuming substances), but they consume a negligible amount of ozone due to their extremely low concentrations [34]. This explains that $Y_{O_3/X}$ has no effect on the bacteria concentration as the ozone requirement for this reaction is met under normal operational conditions.

3.3. Scenario analysis

The effect of applied ozone concentration and flow rate on certain key variables was evaluated. The normally operational ozone

 Table 2

 RSF values indicating the degree influence of model parameters on output variables.

Parameter	RSF	RSF						
	[O ₃]	[OD]	[X]	[BrO ₃]				
k ₀₃	0.07	0.02	0.07	0.07				
k _{OD}	0.30	0.18	0.28	0.30				
k _X	0	0	0.94	0				
$k_{\rm Br}$	0	0	0	0.98				
Y _{O3/OD}	0.4	0.11	0.38	0.4				
Y _{O3/X}	0	0	0	0				



Fig. 7. Scenario analysis, effect of flow rate on OD and bromate concentrations.

dose in the drinking water production centre is 2.5 mg l^{-1} . Scenarios were calculated for concentrations varying from 0 to 5 mg l^{-1} . Flow rate (and consequently hydraulic retention time) was varied within a range of 0 (batch reactor) to $1 \text{ m}^3 \text{ s}^{-1}$, while real influent flow rates are in the range of $0.4-0.5 \text{ m}^3 \text{ s}^{-1}$.

Effects of ozone dose on effluent OD and bromate formation are shown in Fig. 7. This figure reveals that a compromise has to be made regarding ozone dose and flow rate to comply with bromate levels without losing OD (and bacteria) removal goals. Fig. 7 also shows that the normally applied flow rates in this case are in a beneficial range. A significant amount of OD is removed and the bromate level stays well below the standard of $10 \,\mu g \, l^{-1}$. The influent OD concentration was $11.3 \, m^{-1}$.

As can be deduced from Fig. 8, the operational ozone concentration of 2.5 mg l^{-1} is well chosen. Again, sufficient OD is removed and the bromate guideline is met without problems.

3.4. Bromate formation

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Sohn et al. [19] stated that most of the models for predicting residual ozone and bromate formation take empirical functional forms because the complexity of natural organic matter restricts developing complete theoretically based chemical kinetic models. They developed a multiple regression model that was compared to experimental data from the VMW [21] (and references therein):

$$BrO_{3} = 1.55 \times 10^{-6} \times [TOC]^{-1.26} \times [pH]^{5.82} \times [O_{3}]^{1.57} \times [Br]^{0.73} \times t^{0.28} \times (1.035)^{T-20}$$
(7)

Fig. 9 shows this comparison. The model developed in this study was added to the graph. The regression model seems to show a better prediction of the experimental data, as also indicated when comparing the TIC values of both, the regression and the kinetic models with values of 0.17 and 0.21, respectively.



Fig. 8. Scenario analysis, effect of ozone dose on OD and bromate concentrations.



Fig. 9. comparison of predicting performance of a regression model and the kinetic model developed in this study.

However, both models agree well with reality (TIC < 0.3). The overestimation of bromate by the kinetic model can be caused by inaccurate parameter estimation with insufficient data as $k_{\rm Br}$ has proven to be very influential to simulated bromate concentrations (see Section 3.2). The assumptions that were made regarding the bromate mechanism, together with not included temperature effects can contribute to less prediction accuracy, together with the fact that only 1 data point was available for calibration. Regression models are able to describe experimental data very well and temperature or reactor correction factors can be easily added [19]. On the other hand, models based on well defined mechanisms can give more substantiated insight in processes and are easier to apply.

4. Conclusions

In this study, a simplified kinetic model describing ozone decomposition, organic carbon removal, disinfection and bromate formation during ozonation applied in drinking water production was developed. Calibration and simulation runs were based on historical data from a full-scale ozonation system at the Flemish Water Supply Company waterworks in Kluizen, Belgium. It was demonstrated that the developed model is able to predict excess ozone concentration, OD removal, bacteria inactivation and bromate formation, although further data collection and batch experiments will be necessary to further validate the model. A sensitivity analysis revealed that parameters involving optical density (both rate constants and stoichiometric coefficients) strongly affect model output. Some parameters with respect to bromate and bacteria showed to be only, but to a large extent, sensitive to their associated concentrations. OD seems to be a valuable parameter for the application of model-based control and optimization strategies as it can be determined on-line and consequently a huge amount of real-time and accurate data are available.

With drinking water standards becoming more stringent, models will become an important tool to assess drinking water plant performance [35]. This model will be used for further scenario analysis (particularly effect of ozone dosage on reactor performance) and will play an important role in the drinking water modelling studies in Flanders. The model will be used as starting point for a more detailed model which includes radical reactions (i.e. model of Staehelin, Hoigné and Bühler) [10,13,14] and a combination of AOP techniques (ozone, UV, H_2O_2) to guarantee satisfactory model predictions and to improve the applicability and optimization capacities.

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