**beditor**: A Computational Workflow for Designing Libraries of Guide RNAs for CRISPR-Mediated Base Editing

Rohan Dandage,*†‡§, Philippe C. Després,*‡§, Nozomu Yachie,**†‡§ and Christian R. Landry*†‡§

*Département de Biochimie, Microbiologie et Bio-informatique and **Département de Biologie, Faculté de Sciences et Génie, Université Laval, Québec, Québec G1V 0A6, Canada, †PROTEO, The Québec Research Network on Protein Function, Structure and Engineering, ‡Centre de Recherche en Données Massives (CRDM), and §§Research Center for Advanced Science and Technology, University of Tokyo, 153-8904, Japan, ††Department of Biological Sciences, Graduate School of Science, University of Tokyo, 113-0033, Japan, ‡‡Institute for Advanced Biosciences, Keio University, Tsuruoka, 997-0035, Japan, and §§Graduate School of Media and Governance, Keio University, Fujisawa, 252-0882, Japan

ORCID IDs: 0000-0002-6421-2067 (R.D.); 0000-0003-1407-0643 (P.C.D.); 0000-0003-1582-6027 (N.Y.); 0000-0003-3028-6866 (C.R.L.)

ABSTRACT CRISPR-mediated base editors have opened unique avenues for scar-free genome-wide mutagenesis. Here, we describe a comprehensive computational workflow called beditor that can be broadly adapted for designing guide RNA libraries with a range of CRISPR-mediated base editors, Protospacer Adjacent Motif (PAM) recognition sequences, and genomes of many species. Additionally, to assist users in selecting the best sets of guide RNAs for their experiments, a priori estimates of editing efficiency, called beditor scores, are calculated. These beditor scores are intended to select guide RNAs that conform to requirements for optimal base editing: the editable base falls within maximum activity window of the CRISPR-mediated base editor and produces nonconfounding mutational effects with minimal predicted off-target effects. We demonstrate the utility of the software by designing guide RNAs for base editing to model or correct thousands of clinically important human disease mutations.

KEYWORDS CRISPR; base editing; gene editing; genome-wide mutagenesis; gRNA design

CRISPR-mediated base editors (BEs) are engineered by fusing a DNA-modifying protein with a nuclease-defective Cas9 (dCas9) protein, allowing scar-free targeted mutagenesis (Nishida et al. 2016; Gaudelli et al. 2017; Hess et al. 2017; Kim 2018; Rees and Liu 2018). Currently, two major types of BEs are available: cytosine base editors (CBEs) that enable the conversion of cytosine to uracil by catalysis and then to thymine through replication or repair (Hess et al. 2016; Komor et al. 2016; Ma et al. 2016; Nishida et al. 2016), for example BE3 (Komor et al. 2016), and Target-AID (C•G to T•A) (Nishida et al. 2016). Similarly, adenine base editors (ABEs) enable conversion of adenine to inosine by catalysis and then to guanine through replication or repair (A•T to G•C) (Cox et al. 2017; Gaudelli et al. 2017; Zafra et al. 2018), for example ABE7.10 (Gaudelli et al. 2017). Currently BEs enable many codon level substitutions and thus amino acid substitutions (Supplemental Material, Figure S1). Owing to the unique capability of scar-free mutagenesis, BEs have recently found numerous applications in both model and nonmodel organisms (Shimatani et al. 2017; Kim 2018; Qin et al. 2018; Tang and Liu 2018; Zafra et al. 2018) and have substantial promise in therapeutic applications (Rossidis et al. 2018; Villiger et al. 2018).

For designing guide RNAs (gRNA) in BE-mediated mutagenesis experiments, several specific requirements for the optimal activities of DNA modifying and Protospacer Adjacent Motif (PAM) sequences need to be taken into consideration. The complexity of this task especially increases when designing gRNA libraries against large sets of targets located across a given genome. Because the design of gRNA sequences is a primary requirement for the success of the BE-mediated...
mutagenesis experiment, the methods and strategies involved in designing gRNA libraries are arguably one of the most important factors in such experiments. Currently available gRNA designing tools are however either specifically focused on nonsense mutations (Billon et al. 2017) or allow limited customization (BioRxiv: https://doi.org/10.1101/373944) (see Table S1), leaving a major roadblock in the applications of the BEs in genome editing. The continuous discoveries of new BEs and the expansion of their existing editing capabilities (Nishimasu et al. 2017) demand a complementary development in computational methods that would allow designing gRNAs using new and improved BEs. Moreover, considering the prospective applications across nonmodel organisms, compatibility with diverse genomes is essential. Overall, therefore, a comprehensive computational framework to design gRNA libraries can potentially fuel the progress in CRISPR-base editing technology and its diverse applications in genome editing.

We developed a comprehensive computational workflow called beditor (Figure 1A) that can design gRNA libraries with any requirements of DNA-modifying enzymes. These include the range of nucleotides where maximum catalytic activity of BE occurs, henceforth simply referred to as “activity window,” and PAM recognition sequence. beditor is directly compatible with >125 genomes hosted in the Ensembl genome database (Zerbino et al. 2018) and any annotated custom genomes. Additionally, the beditor workflow provides a priori estimates called beditor scores for each gRNA that accounts for specific editing requirements of BEs, gRNA binding at off-target sites, and the number and types of off-target effects (Figure 1B). Such estimates will inform researchers in optimizing throughput of their BE-mediated mutagenesis experiments. With its Graphical User Interface (GUI), command line interface, and open source Application programming interface (API), the beditor workflow has broad applicability for the genome editing community, and its open source implementation will allow for continuous enhancements in the future.

Materials and Methods

Implementation

The beditor workflow is implemented as an open source python 3.6 package hosted at https://pypi.org/project/beditor. The source code of beditor can be accessed at https://www.github.com/rraadd88/beditor. The documentation of the software with API is available at https://www.github.com/rraadd88/beditor/README.md. The beditor workflow depends on other open source softwares such as PyEnsembl (PyEnsembl 2018), BEDTools (Quinlan and Hall 2010), BWA (Li and Durbin 2009), and SAMtools (Li et al. 2009) at various steps of the analysis. User-provided mutation information is first checked for validity with PyEnsembl (https://github.com/openvax/pyensembl). Genomic sequences flanking the mutation sites are fetched using BEDTools (Quinlan and Hall 2010). The designed gRNAs are aligned with the reference genome using BWA (Li and Durbin 2009), and alignments are processed using SAMtools (Li et al. 2009) for evaluation of off-target effects using the beditor scoring system. Visualization of alignments of guide RNAs with genomic DNA are created using the DnaFeaturesViewer package (https://github.com/Edinburgh-Genome-Foundry/DnaFeaturesViewer).

beditor scoring system

Alignment of the designed gRNAs (with PAM sequence) with the provided reference genome is carried out using BWA, as used in Li and Durbin (2009), allowing for a maximum of two mismatches per alignment (Concordet and Haeussler 2018). The beditor score is evaluated as follows:

\[
P_i = \begin{cases} 
P_{\min} & \text{if mismatch is near PAM,} \\
P_{\max} & \text{if mismatch is distant from PAM,} 
\end{cases} \quad \text{if } \text{mismatch is distant from PAM,} \quad \text{if } \text{mismatch is near PAM,} \\
\end{equation}

\[
P_a = \prod_{i=1}^{M_{\max}} P_i \quad \text{if } \text{mismatch is distant from PAM,} \quad \text{if } \text{mismatch is near PAM,} \\
\end{equation}

\[
G_a = \begin{cases} 
G_g & \text{if genic,} \\
G_{ig} & \text{if intergenic.} 
\end{cases} \quad \text{if } \text{mismatch is distant from PAM,} \quad \text{if } \text{mismatch is near PAM,} \\
\end{equation}

\[
B = \left( \sum_{i=1}^{n} P_i * G_a \right)^* A. \quad \text{if } \text{mismatch is distant from PAM,} \quad \text{if } \text{mismatch is near PAM,} \\
\end{equation}

For an alignment between a gRNA sequence and the genome, \(P_i\) is a penalty assigned to a nucleotide in the gRNA sequence based on the position of a mismatch in the aligned sequence relative to the PAM. If the mismatch is near the PAM sequence, a minimal penalty \(P_{\min}\) is assigned. Conversely, if the mismatch is far from the PAM, a maximum penalty \(P_{\max}\) is assigned. The relative values of such penalties were determined by fitting a third-degree polynomial equation to the mismatch tolerance data from Doench et al. (2016) (Figure S2). This way, penalties increase non-linearly from \(P_{\min}\) to \(P_{\max}\) as the distance of nucleotide \(i\) from PAM sequence increases. Individual penalties assigned for all the mismatched nucleotides in a gRNA (total \(M_{\max}\)) are then multiplied to estimate a penalty score for a given alignment called \(P_a\) (Equation 2). In cases of gRNAs with lengths other than 20, the fitted equation is used to interpolate penalty scores. In case of 5’ PAMs, the order of the vector-containing position-wise penalty scores is reversed. \(G_a\) is a penalty defined by whether the off-target alignment lies within a genic or an intergenic region (Equation 3). \(A\) is a penalty based on whether the editable base lies within the activity window of BE (Equation 4).

Note that due to the lack of large-scale BE editing data, in the current version of beditor, penalties are set based on the importance of each requirement (Table S2). Such penalties would be informed from empirical data in the future developments.

The overall beditor score \(B\) for a gRNA is determined by multiplying penalties assigned per alignment (\(P_a\) and \(G_a\)) for all alignments \(n\) with a penalty assigned to the gRNA (A)
Among many base editor and PAM sequence-specific requirements, nucleotide windows for maximum activity are considered while designing the gRNAs. Finally, potential off-target effects are estimated. (B) A scoring system specifically designed for a priori evaluation of mutagenesis potential of gRNAs. Penalties are assigned based on (1) the total number of off-target alignments of gRNAs to the reference genome, (2) positions of the mismatches in the off-target alignments relative to the PAM, and (3) genomic locations of off-target alignments, and lastly, (4) whether the editable base lies inside the activity window of the BE. Using all the above penalties, a final score is calculated for each gRNA sequence – the beditor score.

As a demonstrative analysis using custom BEs and PAM recognition sequences (Figure 2 and Table S3), 1000 nucleotide and amino acid mutations were randomly assigned across the genome of *Saccharomyces cerevisiae* (https://github.com/rraadd88/test_beditor). Such sets of mutations create uniform data sets ideal for testing features of beditor. The input mutation data were created for both mutation formats (either amino acid or nucleotide) and modes of mutagenesis (“model” or “correct”). The command “beditor–cfg params.yml” was executed. Here, params.yml contains input parameters of the analysis (Table S4) in a user-friendly YAML format. For this analysis, input parameter “host” was set to “Saccharomyces_cerevisiae” and parameter “genomeassembly” was set to “R64-1-1.”

Summary statistics on the editability are included as Table S6, and gRNA libraries are included as Data S2.

**Case study analysis**

For the case study analysis, a set of clinically associated human mutations were obtained from the Ensembl database in GVF format (ftp://ftp.ensembl.org/pub/release-93/variation/gvf/homo_sapiens/homo_sapiens_clinically_associated.gvf.gz, date modified: 08/06/2018, 16:13:00). From genomic coordinates of SNPs, inputs for nucleotide mutations (reference and mutated nucleotide) and amino acid mutations (transcript ids, amino acid position, reference residue, and mutated residue) were identified using PyEnsembl (PyEnsembl 2018). The mutation information was provided to the beditor workflow as a tab-separated file. The input parameters used in this analysis are the same as Table S4 except for the host name, “homo_sapiens,” and genome assembly version, “GrCh38.” Summary statistics on the editability are included as Table S7, and gRNA libraries are included as Data S3.

**Data availability**

The authors affirm that all data necessary for confirming the conclusions of this article are represented fully within the article and its tables and figures.

Processed data from this study i.e., gRNA libraries designed for demonstrative analysis 1, 2 and the case study analysis are provided as Data S1–S3, respectively. This data set has been deposited using the GSA Figshare portal.

The beditor software is available at https://github.com/rraadd88/beditor under the GNU General Public License.
(GPLv3). The database of gRNAs designed through the beditor workflow can be accessed at http://rradd88.github.io/soft/beditor.

The data set analyzed in the study, i.e., set of clinically associated human mutations, was obtained from the Ensembl database in GVF format (ftp://ftp.ensembl.org/pub/release-93/variation/gvf/homo_sapiens/homo_sapiens_clinically-associated.gvf.gz, date modified: 08/06/2018, 16:13:00). Supplemental material available at Figshare: https://doi.org/10.25386/genetics.7448234.

Results

Design of beditor workflow

The beditor workflow contains sequential steps that lead from input target sequences and other input parameters to the designed gRNA libraries as output (Figure 1A). The user provides information about the desired set of mutations as an input, and a library of gRNAs is generated with the corresponding a priori estimates called beditor scores to help users in selecting the best-performing gRNAs. beditor can also be used to execute only a subset of the analysis steps by changing the input parameters or providing inputs for intermediate steps. The standard input of beditor depends on the format of mutations, i.e., nucleotide or amino acid. To carry out nucleotide level mutations, the users need to provide genome coordinates and the desired nucleotide after mutagenesis. For carrying out amino acid level mutations, the users provide Ensembl stable transcript ids, the position of the targeted residues, and the corresponding mutated residue. Users can also provide inputs to limit the amino acid substitutions to a custom substitution matrix and specify whether only

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**Figure 2** Demonstrative analysis of gRNAs designed with custom base editors and PAM recognition sequences. To demonstrate the utility of beditor in utilizing custom base editors and PAM recognition sequences, sets of 1000 randomly assigned mutations in S. cerevisiae (see Supplemental methods) were analyzed in two mutation formats (nucleotide or amino acid) and two modes of mutagenesis (“model” and “correct”) (each combination is shown in parts A–D). In each heatmap, number of gRNAs designed by each combination of a base editor (in columns) and PAM recognition sequence (in rows) is shown.
nonsynonymous or synonymous substitutions should be carried out. In addition to creating mutations on a wild-type background (“model” mode), the \textit{beditor} workflow also provides an option to design guides that would remove alternative SNPs and to mutate to the reference or wild-type alleles (“correct” mode). The program can be accessed via GUI (Figure S3), command line, or API. In addition to the gRNAs designed to carry out provided mutations, the \textit{beditor} workflow can also design control gRNAs that are important in the large-scale mutagenesis experiments. The positive control gRNAs designed by \textit{beditor} generate nonsense mutations in the coding region of interest and negative control gRNAs that lack an editable nucleotide in the editing window of the BE, thus expected to have null effect.

\textbf{Customizability for broad utility}

The \textit{beditor} workflow utilizes a PyEnsembl python API (PyEnsembl 2018) to fetch and work with the genomes of over 125 species and their various assemblies from the Ensembl genome database (Zerbino \textit{et al.} 2018), providing a broad utility for researchers across a wide spectrum of fields. \textit{beditor} is also compatible with any custom user-made annotated genome. The ability to carry out parallel processing allows for the design of large gRNA sequence libraries using minimal computational resources (Figure S4). The users can incorporate BEs with varied editing properties and even novel BEs as per requirements. Similarly, they can incorporate any custom PAM sequences (e.g., Table S3) in addition to the experimentally validated PAMs already incorporated in the current version of \textit{beditor} (listed in Figure 3).
Table S8). Additionally, the location of the PAM with respect to the gRNA (upstream or downstream) and the provided length of gRNA are taken into consideration in the analysis. Lastly, beditor scores allow users to select the best set of gRNAs for mutagenesis experiments.

**Selecting the best-performing gRNAs**

We defined a novel beditor scoring system that can be used to select the best-performing gRNAs from a designed gRNA library. Due to the lack of large-scale, genome-wide base editing data, we relied on a few general rules that are applicable to all BEs. These rules pertain to the requirements for optimal mutagenesis. With the penalties assigned to each requirement, the beditor score of the optimally performing guide RNA would tend to be higher while that of poorly performing guide RNA would be lower (see Materials and Methods). In the future, we wish to determine penalty scores from empirical data. The current version of beditor scoring system assesses four of the general requirements for optimal mutagenesis. Based on the conformity of the gRNA to the requirements, four penalty scores are assigned (Figure 1B, see Materials and Methods). (1) While general rules of the BE-mediated editing is an active field of research, we utilize the basic rule which is common between all base editors: the editable base should lie within the maximum activity window of the BE. Thereby, the gRNAs are penalized if the editable base does not lie within the maximum activity window of BE. Next, utilizing the alignments of gRNAs to the genome, potential off-target sites are identified. (2) A penalty score is assigned based on the gRNA binding at off-target. It is evaluated by capturing a general trend of mismatch tolerance along the length of a gRNA (Doench et al. 2016) (see Materials and Methods). Given the lack of large-scale, base editing data, we used empirical data from “conventional” CRISPR-Cas9–based genetic screens, assuming that the basic principles of gRNA recruitment and binding would be conserved between the two variants of CRISPR-based mutagenesis technologies. (3) Additionally, from alignments of the gRNAs, a penalty is assigned based on the location of the off-target site (genic or intergenic regions). Off-target editing at intergenic regions is less likely to confound the mutational effects compared to functionally important genic regions. Accordingly, penalties are assigned. (4) Lastly, to account for the number of off-target sites, the three penalties are multiplied together to evaluate a beditor score per gRNA. Effectively, the optimal gRNAs have a beditor score of 1, while a lower beditor score indicates incompatibility with BE requirements, significant off-targets, or off-target effects that confound mutational effects. Finally, to filter out the gRNAs containing putative RNA polymerase III transcriptional terminators (Gao et al. 2018), the length of the poly-T stretch per individual gRNA is indicated in the output of beditor.

**Demonstrative analysis 1: customizability with respect to base editing strategies**

We demonstrate that the users can incorporate and use custom BEs and PAM recognition sequences by designing gRNA libraries against sets of 1000 randomly assigned mutations (Supplemental methods) with 6 BEs (among which 4 are not yet developed and are hence called “hypothetical”) and 16 PAM recognition sequences. With all the combinations of BE specific requirements and all combinations of two mutation formats (nucleotide and amino acid) and two modes of mutagenesis (“model” and “correct”), gRNA libraries were designed (Figure 2 and Data S1).

**Demonstrative analysis 2: ability to work with different species**

To show beditor’s ability to work with genomes of different species, we designed gRNA libraries for 10 representative species.
Ensembl genomes, for sets of 1000 randomly assigned mutations (Supplemental methods). The genomes of all the species were directly fetched from the Ensembl genome database and gRNA libraries were designed for both mutation formats (nucleotide and amino acid) and two modes of mutagenesis ("model" and "correct"). Through the designed gRNA libraries (Data S2), reasonable editability (Table S6) was achieved for all the species.

**Case study: designing a gRNA library against a set of clinically relevant SNPs**

To demonstrate the utility of our computational workflow, we designed a library of gRNAs against a set of clinically relevant SNPs in the human genome composed of 61,083 nucleotide level and 81,819 amino acid level mutations (see Supplemental methods). This analysis was carried out with two different BEs: Target-AID and ABE, and two PAM sequences: NGG (Jinek et al. 2012; Hsu et al. 2013) and NG (Hu et al. 2018; Nishimasu et al. 2018) and in “model” and “correct” mode. Note that the purpose of this case study analysis is to provide a demonstration of the functionalities of the beditor workflow. For instance, ABE base editor is not known to function in association with NG PAM. Yet, as shown in demonstrative analysis 1, the users may try any combinations of custom BEs and PAM sequences. The output libraries of gRNA sequences (Data S3) target ~25% of the total mutations provided as input (Table S7). The resulting gRNA libraries were composed of gRNAs designed with each of the input BEs and PAM sequence which targeted both the strands (Figure 3, A and B). On average, ~1.6 guides were designed for each mutation. Summary visualizations of gRNA libraries (Figure 3C), as well as visualizations of alignments of gRNAs with the target sequence (Figure S5), were generated. Additionally, the percentage of substitutions that can be edited with the designed guides (% editability) is represented as substitution maps (Figure 4) to indicate the proportion of input mutations that can be edited with the input BEs and PAMs. The gRNAs designed for this case study analysis, as well as with all the combinations experimentally validated pairs of BEs and PAM sequences (Table S8), are provided as a database with a web interface at http://rraadd88.github.io/soft/beditor.

**Performance evaluation of beditor score**

From the case study analysis, beditor scores were evaluated for each gRNA sequence in the library. From the distribution of scores (Figure S6), the users may assign a threshold to filter out gRNAs with lower beditor scores. Collectively, by definition, the beditor scores are negatively correlated ($\rho = -0.94$) with the number of off-target alignments (Figure 5A), and the penalty assigned for each alignment based on distance of mismatches from the PAM sequence is positively correlated ($\rho = 0.65$) with the distance (Figure 5B). Note that the rank correlation is not perfect because of cases in which there were two mutations in the aligned sequence. Also, purely informed from features of alignments of gRNAs and the requirements of BEs, the beditor score recapitulates empirical activity values of gRNAs with a strong positive correlation ($\rho = 0.95$) determined in terms of Cutting Frequency Determination (CFD) score (Donech et al. 2016) (Figure 5C).

**Discussion**

CRISPR-mediated BEs have recently become a new paradigm in genome editing, owing to their unique ability to carry out precise mutations without the need for DNA breaks (Rees and Liu 2018). Consequently, a plethora of applications of BEs have emerged from all corners of the genome editing community, ranging from study of model and nonmodel organisms (Shimatani et al. 2017; Kim 2018; Qin et al. 2018; Tang and Liu 2018; Zafra et al. 2018) to therapeutics (Rossidis et al. 2018; Villiger et al. 2018). However, there has been a lack of robust and customizable software that can design gRNA libraries with any specific requirements of CRISPR-mediated base editing experiments. As presented here, the

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**Figure 5** Performance assessment of beditor scores from case study analysis of clinically relevant human SNPs. (A) Relationship between the number of genome-wide off-target alignments and beditor score per gRNA. The color of hexbins are scaled according to the number of gRNAs per bin. (B) Relationship between the distance of a mutation in off-target alignments and corresponding penalty assigned (P). The color of hexbins are scaled according to the number of off-target alignments per bin. (C) Relationship between the CFD score and beditor score for all the gRNAs carrying NGG PAM sequence. The color of hexbins are scaled according to the number of gRNAs per bin. $\rho$ is Spearman’s correlation coefficient.
novel computational workflow of beditor (Figure 1A) fills in this important gap by allowing comprehensive customizability in terms of requirements of BE, PAM sequence, and genome and thus increasing the applicability of the CRISPR-mediated base editing technology to a broader community of researchers.

We demonstrate the modularity of beditor workflow in terms of the type of BE, PAM, and genome by extensively testing them on synthetic sets of mutations. We also show that the workflow can be used to either create a mutation (“model” mode) or remove it (“correct” mode) with support for both nucleotide and amino acid format of mutations. Collectively, therefore, in terms of integrated customizability alone, beditor workflow provides a significant advance over other methods that provided only limited utilities (Table S1). In addition, we also introduce a novel method for a priori estimation of mutagenesis potential of gRNAs (Figure 1B) that utilizes features obtained from off-target alignments such as distance between mismatch and PAM recognition sequence. Such estimations would allow users to select a subset of designed gRNA library that would provide optimal mutagenesis in their experiments.

From the case study analysis of ~60,000 human clinically relevant SNPs, we show that the beditor workflow provides all-round gRNA design capabilities, scanning through combinations of multiple strategies (Figure 3, A and B). The validations of beditor score from this analysis revealed that rather simple penalty-based evaluations efficiently captured dependence on position of the mismatch in the alignment from the PAM sequence (Figure 5B) and dependence of the off-target effects on number of alignments per gRNA (Figure 5A). Also, the estimations were in strong correlation with empirical data on the off-target effects obtained from “conventional” CRISPR screens (Figure 5C). This is supported by the currently available data (Kim et al. 2017), which suggests that the Cas9-induced off-target effects could be the predictors of off-target effects of BEs (Rees and Liu 2018). However, recent studies have found that the off-target effects of BEs can vary between ABE and CBE (Lee et al. 2018; Jin et al. 2019; Zuo et al. 2019) and could even be largely unpredictable as in the case of CBEs containing APOBEC1 (Jin et al. 2019; Zuo et al. 2019). Therefore, in future, we wish to update the beditor scoring system further, as more information would emerge from large-scale base editing experiments. The future developments of beditor workflow would be carried out in an open source manner at https://github.com/rraadd88/beditor.

Together, considering the wide interest in scar-free and precise mutagenesis endowed by CRISPR-mediated base editing, moving ahead, novel and comprehensive design of gRNAs through beditor workflow is expected to be applicable to a broad community of researchers and possibly become an essential component of the CRISPR-mediated base editing technology itself.

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