

Nozzle: a report generation toolkit for data analysis pipelines

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ABSTRACT

Summary: We have developed Nozzle, an R package that provides an Application Programming Interface to generate HTML reports with dynamic user interface elements. Nozzle was designed to facilitate summarization and rapid browsing of complex results in data analysis pipelines where multiple analyses are performed frequently on big datasets. The package can be applied to any project where user-friendly reports need to be created.

Availability: The R package is available on CRAN at <http://cran.r-project.org/package=Nozzle.R1>. Examples and additional materials are available at <http://gdac.broadinstitute.org/nozzle>. The source code is also available at <http://www.github.com/parklab/Nozzle>.

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1 INTRODUCTION

Owing to the increasing size and complexity of biological data, a considerable portion of bioinformatics analyses are implemented as (semi-)automated pipelines. The main task of these pipelines is to control the flow of large datasets through a series of analysis modules, which are often independent applications. These analyses typically need to be performed repeatedly over time as data are accumulated gradually. GenePattern (Reich *et al.*, 2006), Galaxy (Goecks *et al.*, 2010) and Taverna (Hull *et al.*, 2006) are popular workflow management systems used to implement such pipelines for high-throughput analysis of genomics data. Another common approach is to use shell scripting to tie together different tools into an analysis pipeline. Although pipelines greatly reduce the effort required to apply different algorithms to large datasets, they often result in a multitude of figures, lists and tables at varying levels of detail. This presents significant challenges for the biologists who must sift through a large amount of information to identify potentially interesting findings. In the use case we describe, each run of the pipeline results in more than 11 000 figures and tables. In some cases, this problem can be addressed by feeding these results into a database with a dynamic query interface. Although databases are capable of supporting powerful exploration tools and interactive visualizations, their development and maintenance require a significant investment of resources and their data model is less flexible than

reports. This is particularly problematic when the tools in the pipeline are replaced as requirements change. Alternatively, the results can be summarized in reports. Several R libraries exist that can be used to generate reports for analysis pipelines. R2HTML (Lecoutre, 2003) and hwriter (<http://cran.r-project.org/package=hwriter>) are low-level libraries for writing HTML files. Sweave (Leisch, 2002), knitr (Xie, 2012) and related tools for reproducible research can be used to generate PDF or HTML documents from within R scripts. However, none of these tools provide support in generating reports with dynamic user interface components for the presentation of extensive and complex analysis results. To address these limitations, we have developed the Nozzle R package, which supports pipeline developers in creating comprehensive and user-friendly HTML reports to describe the results of analysis pipelines.

2 NOZZLE REPORTS

Nozzle reports are generated bottom-up (Fig. 1): in Phase 1 report elements are generated, e.g. a table and a paragraph of text, in Phase 2 they are assembled into larger structures, e.g. the table and the paragraph are added to a titled section, which is added to the report, and in Phase 3 the report is rendered into HTML.

The guiding principle for the design of the Nozzle package is to enable report authors—usually pipeline developers—to focus on the content rather than on the layout or generation of the report. We achieve this through the high-level R Application Programming Interface (API) that enables authors to create report elements such as figures, tables, paragraphs of text, bibliographic or web references, lists, sections and subsections using regular R commands without knowledge of the technologies used for the presentation of the final report.

Nozzle reports provide a rich user interface (Fig. 2). All figures support dynamic switching between a thumbnail view and a detail view. They also have a caption and can be linked to a high-resolution or PDF version of the figure. All tables are sortable and support automatic trimming of floating point numbers to a user-defined number of significant digits. Tables also include a caption and can be linked to external files that contain additional information. Text and the content of table cells can be marked up semantically using the Nozzle-specific ‘result’ style as well as common styles such as ‘parameter’, ‘link’ or simply ‘emphasis’. The ‘result’ markup indicates that the corresponding text represents an analysis result (Fig. 2b). When a result is flagged as significant, Nozzle uses this information to guide readers to

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Phase 1: create report elements

```
r <- newCustomReport( "My Report" );
s <- newSection( "My Section" );
ss1 <- newSection( "My Subsection 1" );
ss2 <- newSection( "My Subsection 2" );
t <- newTable( iris[45:55,], "Iris data." );
p <- newParagraph( "Some sample text." );
```

Phase 2: assemble report structure bottom-up

```
ss1 <- addTo( ss1, t ); # parent, child_1, ..., child_n
ss2 <- addTo( ss2, p );
s <- addTo( s, ss1, ss2 );
r <- addTo( r, s );
```

Phase 3: render report to file

```
writeReport( r, filename="my_report" ); # w/o extension
```

Fig. 1. Sample R script to create a basic Nozzle report that illustrates the three phases of the bottom-up approach. See Supplementary Figure S1 for the HTML report

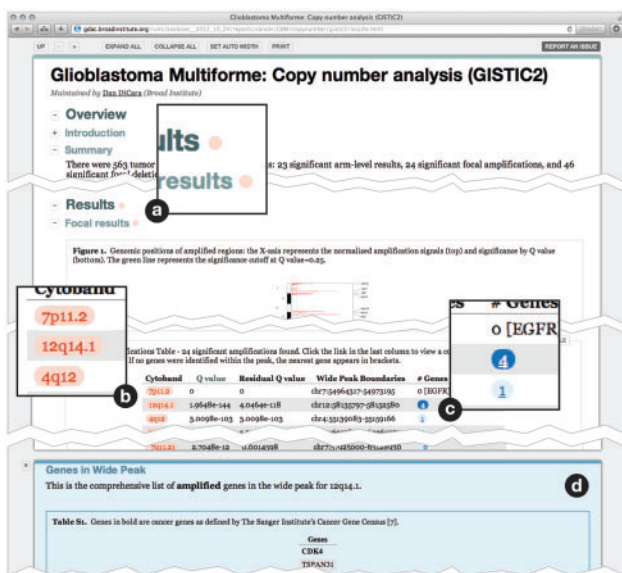


Fig. 2. A sample Nozzle report. (a) Red markers indicate statistically significant—as defined by the report author—results in this section. (b) Red boxes indicate significant results. (c) Underlined results have associated supplementary information. Clicking opens the (d) Supplementary Information panel

sections containing significant results by highlighting the corresponding section (Fig. 2a). This is particularly useful in comprehensive reports with many sections or in situations when readers must frequently review large numbers of reports and would like to focus first on significant findings.

Results can also be linked to Supplementary Information (Fig. 2c and d), which may contain any report elements including sections, figures and tables. These are shown on demand using a split-screen approach, allowing the readers to view main results and Supplementary Information side by side. This is a powerful

tool for creating reports that focus on the key findings while still providing access to more detailed information.

3 R API

R was chosen over other programming languages for the Nozzle API owing to its large user base and high popularity among practitioners in the computational biology and bioinformatics communities. Many biological data analysis pipelines contain components implemented in R, making integration of the Nozzle library straightforward.

The Nozzle R API was intentionally kept simple with only four key classes of methods: *constructor* (e.g. *newFigure*, *newSection*) and *formatter* methods (e.g. *asResult*, *asParameter*) to create and format content during the first phase, *assembly* methods (e.g. *addTo*, *addToResults*) to combine elements during the second phase and finally a single *writeReport* method to render the assembled report in the third phase. Additionally, Nozzle provides a set of advanced features that give developers more control over the content and structure of reports. For example, the API contains several setter/getter methods to modify parameters of the report, e.g. maintainer or copyright information, logos or Google Analytics tracking identifiers. Report authors may also overwrite the default report styles such as fonts and colors by providing a Cascading Style Sheets (CSS) file. Furthermore, developers can define the visibility of report elements using three privacy levels (private, group and public) and exclude pertinent sections in the final report by providing a corresponding visibility flag when rendering to HTML. This feature allows developers to easily censor sensitive information in public reports.

4 IMPLEMENTATION

Nozzle works with R 2.10 or later. Internally, reports are represented as a tree of report elements and implemented as nested R list objects. These lists are rendered into HTML files that include a set of JavaScript functions to support dynamic features of the user interface. For this purpose, JQuery (<http://www.jquery.com>) and the JQuery Table Sorter plugin (<http://www.tablesorter.com>) are embedded, making Nozzle reports independent of external library files. Likewise, default CSS definitions are embedded in the HTML and used for layout and styling of the reports. The reports are compatible with Firefox 4+, Chrome 12+, Safari 5+, Opera 11+ and Internet Explorer 9+.

5 USE CASE: THE CANCER GENOME ATLAS

We have deployed Nozzle in the context of the Firehose pipeline management system developed at the Broad Institute (<http://gdac.broadinstitute.org>). Firehose is used for comprehensive automated and reproducible analyses of the data generated by The Cancer Genome Atlas (TCGA; <http://cancergenome.nih.gov>). In December 2012, the analysis workflow comprised approximately 35 different modules. They cover a wide range of analyses on different data types, including clustering of messenger RNA, microRNA and methylation data, copy number analysis with GISTIC 2.0 (Mermel *et al.*, 2011), mutation analysis, correlation analyses between clinical and various molecular data,

as well as pathway analyses. In this project, a team of 10 software developers and computational biologists from four institutions (Broad Institute, Dana-Farber Cancer Institute, Harvard Medical School and Institute for Systems Biology) used Nozzle to implement reports for individual pipelines. The complete analysis workflow is currently run once per month for each of 27 disease cohorts studied in TCGA, generating close to 500 reports. These reports are publicly available on the website of the TCGA Genome Data Analysis Center (GDAC) at the Broad Institute (see above URL) and archived by the TCGA Data Coordination Center. Between 1 February 2012 and 31 December 2012, more than 2700 viewers have accessed more than 17 000 reports in approximately 6000 visits.

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