Manual of the MATLAB pipeline for the analysis of genomic time course data from the *Gene Expression Omnibus* (Version 1.43)

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1 The Pipeline for Dynamic Gene Expression Data

The Pipeline for Dynamic Gene Expression Data, also referred to as Pipeline4DGEData, is the implementation of the pipeline proposed by Carey et al. (2017). It is written in MATLAB (R2016a). The tool is designed to analyze time course data from the Gene Expression Omnibus (GEO) in order to discover gene regulatory networks from high-throughput data.

Section 2 provides a very brief summary of the theoretical concepts involved in the usage of Pipeline4DGEData. However, the reader is strongly encouraged to refer to the GEO documentation and the work of Carey et al. (2017) for full details. Section 3 provides installation instructions and Section 4 illustrates the usage of Pipeline4DGEData. The example presented in the latter should be used as basis for other analyses to be performed by the reader.

2 Theoretical basis and terminology

2.1 The Gene Expression Omnibus (GEO)

The Gene Expression Omnibus (GEO)¹ is a public repository administered by the National Center for Biotechnology Information (NCBI)² for the storage of high throughput data. The GEO stores microarray data in MIAME compliant formats and makes it publicly available. The data is organized into series, samples and platforms.

2.2 GEO Series

A data series is a set of related samples. It may also provide information of the investigation associated to the contained samples. The GEO identifies each series by an accession number with the prefix 'GSE'. For example, series GSE59015.

¹https://www.ncbi.nlm.nih.gov/geo/

²https://www.ncbi.nlm.nih.gov

2.3 GEO Samples

The data obtained from an experimental sample. It also describes how the experiment was conducted and how the data was handled or manipulated. The GEO identifies each sample by an accession number with the prefix 'GSM'. For example, samples GSM1424445 and GSM1424453, which are associated to series GSE59015.

2.4 GEO Platforms

A description of the microarray used to collect a sample. The GEO identifies each platform by an accession number with the prefix 'GPL'. For example, platform GPL17880, which is associated to the samples in series GSE59015.

2.5 Experimental conditions

An experimental condition is the set of time-independent characteristics or parameters of a sample and/or the circumstances in which it was taken. Usually the main characteristic differentiating a condition from a related one is a stimulus (or lack of), which normally consists of a treatment of viral infection. Depending on the study, other characteristics may be present. These may include, the subjects gender or cell type used, for instance. The experimental condition excludes the time point, if any.

In series GSE59015 there are two experimental conditions, namely Wildtype and D10, each one comprising eight samples.

2.6 Study

A study is a set of samples that refer to a common domain of experimental conditions. A common domain of experimental conditions may consist of, for instance, the presence of a stimulus (treatment or viral infection) and the absence of it. In most cases, the series and study coincide. In these cases, all the samples contained in the series belong to the same study. However, it is possible that the GEO stores series that encompass samples from varying domains. In these cases, the samples contained in the series are grouped into different studies.

2.7 Experimental macroconditions

2.8 Replicates

Replicates can be biological or technical.³

2.9 Genes' between-replicate noise

Genes may exhibit variation across replicates from the same condition. This variation, referred to as the *gene's between-replicate noise* and denoted by δ , is measured for each gene as follows. Let $x_{i,t}^s$ be the expression level of replicate s's i-th gene at time t and let S be the total number of replicates. Let $\sigma_{i,t}$ be the standard deviation of the expression level of the i-th gene at time t across the S replicates. The gene i's between-replicate noise, denoted by δ_i , 4 is given by

$$\delta_i = \frac{1}{T} \sum_{t \in t_1}^{t_T} \sigma_{i,t}. \tag{1}$$

Section 9 introduces the MATLAB function provided for the calculation of the genes' between-replicate noise from a group of replicates.

3 Installation of Pipeline4DGEData

The system requirements of Pipeline4DGEData are listed below.

• MATLAB R2016a or newer.

The sources of Pipeline4DGEData are available on its official GitHub repository. The latest version is always available for download from here as a compressed folder. Once downloaded, this folder must be decompressed. By default, the decompressed folder is named Pipeline4DGEData-master/. In

³Biological Replicates vs Technical Replicates.

⁴Another possibility is to define δ_i in terms of the coefficient of variation instead of the standard deviation. In other words, the between-replicate noise could be defined as $\delta_i = \frac{1}{T} \sum_{t \in t_1}^{t_T} \frac{\sigma_{i,t}}{\mu_{i,t}}$, where $\mu_{i,t}$ is the mean of the expression level of the *i*-th gene at time t across the S replicates. However, this measure is unreliable given that expression levels are interval scale measures and the coefficient of variation is reliable only in ratio scale measures.

this manual it will be assumed that this folder has been renamed Pipeline4DGEData/and so it will be referred to as such throughout this document. This folder is the home directory of Pipeline4DGEData.⁵ User-defined input for new analyses to be run must be provided in subfolder Input (details in Section 5) and the program returns the results in subfolder Output (details in Section 6).

Pipeline4DGEData uses the Functional Data Analysis for MATLAB library, which is available from its official site online. The library can be downloaded from this direct link as a compressed folder. Once downloaded, the decompressed folder (named fdaM/) must be copied to Pipeline4DGEData/. That is to say, all the sources of the library must be in path Pipeline4DGEData/fdaM/. After this, attached file mean_grouped.m must be copied into Pipeline4DGEData/fdaM/@fd/.

Pipeline4DGEData also requires the SBEToolbox (A Matlab Toolbox for Biological Network Analysis) library (Konganti et al., 2013) (Version 1.3.3 or later). This library is available from its official GitHub repository. It can be downloaded from this direct link as a compressed folder. Once downloaded, the decompressed folder (named SBEToolbox-1.3.3/) must be copied to Pipeline4DGEData/. That is to say, all the sources of the library must be in path Pipeline4DGEData/SBEToolbox-1.3.3/.

Once all the above is done, Pipeline4DGEData will be ready for use. Details in Section 4.

4 Usage of Pipeline4DGEData

This section shows how to use Pipeline4DGEData for the analysis of a set of experimental conditions by illustrating its usage on GEO series GSE59015. Application of Pipeline4DGEData to any other GEO series should be analogous. It is assumed here that has been installed as described in Section 3.

GEO series GSE59015 refers to a study about the resistance of *Plasmodium falciparum* (i.e., the malaria parasite) to antimalarial drugs for the design of new medications. Two experimental conditions (Section 2.5 provides an introduction to experimental conditions) are considered: Wildtype and D10. The latter refers to *P. falciparum* cells that have been exposed to a medication and the former refers to a cells that haven't been exposed to any stimulus and thus serves as the control group. The list of samples is shown in Figure 1.

 $^{^5\}mathrm{Care}$ should be taken in order not to confuse Pipeline4DGEData (the tool) and Pipeline4DGEData/ (the home folder of the tool).

```
GSM1424445 Wildtype 0 hours post invasion
GSM1424446 Wildtype 6 hours post invasion
GSM1424447 Wildtype 12 hours post invasion
GSM1424448 Wildtype 18 hours post invasion
GSM1424449 Wildtype 24 hours post invasion
GSM1424450 Wildtype 30 hours post invasion
GSM1424451 Wildtype 36 hours post invasion
GSM1424452 Wildtype 42 hours post invasion
GSM1424453 D10 ΔIDH/ΔKDH 0 hours post invasion
GSM1424454 D10 ΔIDH/ΔKDH 6 hours post invasion
GSM1424455 D10 ΔIDH/ΔKDH 12 hours post invasion
GSM1424456 D10 ΔIDH/ΔKDH 18 hours post invasion
GSM1424457 D10 ΔIDH/ΔKDH 24 hours post invasion
GSM1424458 D10 ΔIDH/ΔKDH 30 hours post invasion
GSM1424459 D10 ΔIDH/ΔKDH 36 hours post invasion
GSM1424460 D10 ΔIDH/ΔKDH 42 hours post invasion
```

Figure 1: Samples associated to series GSE59015.

The following checklist provides the information that is required for running a pipeline analysis. This information should be compiled before starting Pipeline4DGEData.

- 1. The accession number of the GEO series.
 - In this example, this number is GSE59015.
- 2. The experimental conditions on which the analysis is to be performed and their corresponding GEO samples. And identifying name should be chosen in advance for each condition.
 - In this example, the conditions' names and their samples are as follows.
 - Condition Wildtype comprising samples GSM1424445 through GSM1424452, as shown in Figure 1.
 - Condition D10 comprising samples GSM1424453 through GSM1424460, as shown in Figure 1.
- 3. The condition files for the conditions on which the analysis is to be performed (Section 5.1 provides an introduction to condition files). These

files must be located in folder Input and named as specified in Section 5.

- In this example, the files for the two conditions must be as shown in Figure 3.
- 4. The accession number of the GEO platform associated to the GEO series.
 - In this example, this number is GPL17880.
- 5. The index of the column in the GPL record where the gene names are stored.
 - In this example, this index is 7, as shown in Figure 2.

Once the above is ready, the analysis can be started by running pipeline from the MATLAB console, as shown in Figure 4. The program will then prompt the user to enter the index of the column in the GPL record where the gene names are located (Item 5 in the checklist above). In this example, this index is 7 and must be entered as shown in Figure 5. After this the program will proceed to read and match all the probe ids to the corresponding gene names, informing the user that the process can take some time and notifying the progress as shown in Figure 6. When finished, the program will prompt the user to specify the gene ID type used in the dataset by selecting from a predefined list displayed on the MATLAB console. In this example, the type is AGILENT_ID and thus 5 must be entered, as shown in Figure 7. After this, the program will start the analysis for ALL the condition files present in folder Input. In this example, the program starts with condition D10 as shown in Figure 8. When the analysis has been completed for all the experimental conditions, the program will confirm that the process has been finished successfully, as shown in Figure 9. The results are output to folder Output and organized as explained in Section 6.

Data table										
ID	COL	ROW	NAME	SPOT_ID	CONTROL_TYPE	ORF	CHROMOSOMAL_LOCATION	DESCRIPTION		
1 !	96	164	GE_BrightCorner	GE_BrightCorner	pos					
2 !	96	162	DarkCorner	DarkCorner	pos					
3 !	96	160	DarkCorner	DarkCorner	pos					
4	96	158	PF14_0264_v7.1_P6/7	PF14_0264_v7.1_P6/7	FALSE	PF14_0264	unmapped	protein kinase, pr		
5	96	156	PF13_0192_v7.1_P2/2	PF13_0192_v7.1_P2/2	FALSE	PF13_0192	unmapped	conserved Plasmo		
6	96	154	PFF0150C_v7.1_P2/2	PFF0150C_v7.1_P2/2	FALSE	PFF0150c	unmapped	conserved Plasmo		
7 !	96	152	MAL13P1.133_v7.1_P5/16	MAL13P1.133_v7.1_P5/16	FALSE	MAL13P1.133	unmapped	conserved Plasmo		
8 !	96	150	PFF0630C-A_v7.1_P1/1	PFF0630C-A_v7.1_P1/1	FALSE	PFF0630c-a	unmapped	conserved Plasmo		
9 !	96	148	MAL13P1.264_v7.1_P1/2	MAL13P1.264_v7.1_P1/2	FALSE	MAL13P1.264	unmapped	conserved Plasmo		
10	96	146	PF11_0465_v7.1_P2/3	PF11_0465_v7.1_P2/3	FALSE	PF11_0465	unmapped	dynamin-like prof		
11 !	96	144	PF14_0264_v7.1_P4/7	PF14_0264_v7.1_P4/7	FALSE	PF14_0264	unmapped	protein kinase, pr		
12	96	142	PFF0525W_v7.1_P1/1	PFF0525W_v7.1_P1/1	FALSE	PFF0525w	unmapped	conserved Plasmo		
13	96	140	PFE0360C_v7.1_P2/3	PFE0360C_v7.1_P2/3	FALSE	PFE0360c	unmapped	conserved Plasmo		
14	96	138	PFF0765C_v7.1_P3/4	PFF0765C_v7.1_P3/4	FALSE	PFF0765c	unmapped	conserved Plasmo		
15	96	136	PF13_0155_v7.1_P6/8	PF13_0155_v7.1_P6/8	FALSE	PF13_0155	unmapped	conserved Plasmo		
16	96	134	PFA0160C_v7.1_P2/2	PFA0160C_v7.1_P2/2	FALSE	PFA0160c	unmapped	nucleoside transp		
17	96	132	PFC0875W_v7.1_P6/9	PFC0875W_v7.1_P6/9	FALSE	PFC0875w	unmapped	ABC transporter,		
18	96	130	PF14_0684_v7.1_P2/2	PF14_0684_v7.1_P2/2	FALSE	PF14_0684	unmapped	conserved Plasmo		
10	06	178	DF11 N288 v71 D1/2	DF11 0288 v71 D1/2	FALSE	DF11 0288	unmanned	concerned Discon		

Figure 2: Table of platform GPL17880 showing that the gene names are stored in the 7-th column (titled 'ORF').

```
0 hours,GSM1424453
6 hours,GSM1424454
12 hours,GSM1424455
18 hours,GSM1424456
24 hours,GSM1424457
30 hours,GSM1424458
42 hours,GSM1424459
```

(a) Condition file for condition Wildtype.

```
0 hours,GSM1424445
6 hours,GSM1424446
12 hours,GSM1424447
18 hours,GSM1424448
24 hours,GSM1424449
30 hours,GSM1424450
36 hours,GSM1424451
42 hours,GSM1424451
```

(b) Condition file for condition D10.

Figure 3: Figure 3a shows the contents of the condition file for condition Wildtype and Figure 3b shows the contents of the condition file for condition D10.

Command Window

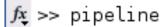


Figure 4

```
Command Window

Loading dataset. This can take some time, please wait...

At Check platform GPL17880's record on GEO and enter the index of the column where the gene IDs/names are located: 7
```

Figure 5

```
Loading dataset. This can take some time, please wait...

Check platform GPL17880's record on GEO and enter the index of the column where the gene IDs/names are located: 7

Loading gene IDs/names. This can take some time, please wait...

Read 1000 gene IDs/names out of 14783.
Read 2000 gene IDs/names out of 14783.
Read 3000 gene IDs/names out of 14783.
Read 4000 gene IDs/names out of 14783.
Read 6000 gene IDs/names out of 14783.
Read 6000 gene IDs/names out of 14783.
Read 7000 gene IDs/names out of 14783.
Read 8000 gene IDs/names out of 14783.
Read 9000 gene IDs/names out of 14783.
Read 10000 gene IDs/names out of 14783.
Read 10000 gene IDs/names out of 14783.
Read 11000 gene IDs/names out of 14783.
Read 12000 gene IDs/names out of 14783.
Read 12000 gene IDs/names out of 14783.
Read 12000 gene IDs/names out of 14783.
Read 14000 gene IDs/names out of 14783.
Read 14000 gene IDs/names out of 14783.
Read 14000 gene IDs/names out of 14783.
Read 14783 gene IDs/names out of 14783.
Read 14783 gene IDs/names out of 14783.
```

Figure 6

```
1: AFFYMETRIX_3PRIME_IVT_ID
2: AFFYMETRIX_EXON_GENE_ID
3: AFFYMETRIX_SNP_ID
    4: AGILENT_CHIP_ID
5: AGILENT_ID
    5: AGILENI_ID
6: AGILENI_OLIGO_ID
7: ENSEMBL_GENE_ID
8: ENSEMBL_TRANSCRIPT_ID
9: ENTREZ_GENE_ID
    10: FLYBASE_GENE_ID
    11: FLYBASE_TRANSCRIPT_ID
12: GENBANK_ACCESSION
13: GENOMIC_GI_ACCESSION
    14: GENPEPT ACCESSION
    15: ILLUMINA_ID
    16: IPI_ID
    17: MGI_ID
18: PFAM ID
    19: PIR_ID
    20: PROTEIN_GI_ACCESSION
21: REFSEQ_GENOMIC
22: REFSEQ_MRNA
    23: REFSEQ_PROTEIN
    24: REFSEQ_RNA
    25: RGD_ID
    26: SGD_ID
    27: TAIR_ID
28: UCSC_GENE_ID
    29: UNIGENE
    30: UNIPROT_ACCESSION
31: UNIPROT ID
    32: UNIREF100_ID
    33: WORMBASE_GENE_ID
34: WORMPEP_ID
    35: ZFIN_ID
Æ Enter the type of gene ID used in the study associated to GSE59015 (e.g., enter 9 for ENSEMBL_GENE_ID): 5
```

Figure 7

```
Enter the type of gene ID used in the study associated to GSE59015 (e.g., enter 9 for ENSEMBL_GENE_ID): 5

The analysis of condition "D10" is starting.
```

Command Window

Figure 8

```
The analysis of condition "D10" is starting.

The analysis of condition "D10" has been completed.

Results have been output to folder /home/jramirez/Pipeline4DGEData/Results/GSE59015/Conditions/D10/

Loading dataset. This can take some time, please wait...

The analysis of condition "Wildtype" is starting.

The analysis of condition "Wildtype" has been completed.

Results have been output to folder /home/jramirez/Pipeline4DGEData/Results/GSE59015/Conditions/Wildtype/

The analysis is complete for all the subjects/conditions.

All results from dataset GSE59015 have been output to folder /home/jramirez/Pipeline4DGEData/Output/GSE59015/Conditions/

A consolidated report on all conditions from dataset GSE59015 can be found in /home/jramirez/Pipeline4DGEData/Output/GSE59015/paper.pdf

A >> |
```

Figure 9

5 Input files

5.1 Condition files

Condition files are the input files for the pipeline analysis of one or more experimental conditions (See Section 2.5 regarding experimental conditions). The information of each experimental condition must be provided in a text file. One file for each condition. The condition files must be stored in folder Input. Each condition file must be named with the following format: [GEO SERIES NUMBER] _-_[NAME OF EXPERIMENTAL CONDITION] _-_[NUMBER OF TOP DRGs FOR CLUSTERING].txt.

All the samples associated with the experimental condition must be provided along with the time points using the format described as follows. Each line of the file must consist of the time point, including the time unit (e.g., 2 hours), followed by a comma (,) and then followed by the accession number of one sample. For example, the file for condition "D10" of *GEO* series GSE59015 must be named GSE59015_-_D10_-_3000.csv and the contents must be as follows.

0 hours,GSM1424453 6 hours,GSM1424454 12 hours,GSM1424455 18 hours,GSM1424456 24 hours,GSM1424457 30 hours,GSM1424458 36 hours,GSM1424459 42 hours,GSM1424460

The condition file can group different replicates that represent the same experimental condition. In this case the pipeline will run one single analysis with the average of all the replicates specified in the file. For example, the file for all the replicates in *GEO* series GSE41067 must be named GSE41067_-ALL_-10000.csv and the contents must be as follows.

```
0 hours, GSM1008154, GSM1008162, GSM1008170
```

- 1 hours, GSM1008155, GSM1008163, GSM1008171
- 2 hours, GSM1008156, GSM1008164, GSM1008172
- 4 hours, GSM1008157, GSM1008165, GSM1008173
- 6 hours, GSM1008158, GSM1008166, GSM1008174
- 8 hours, GSM 1008159, GSM 1008167, GSM 1008175
- 10 hours, GSM1008160, GSM1008168, GSM1008176
- 12 hours, GSM1008161, GSM1008169, GSM1008177

The condition files can be written manually, following strictly the format described above. Optionally, this task can be carried out more easily by using script create_input_files.m. In order to do this, create_input_files.m must be run and the instructions provided thereafter by the program must be followed. More details in Section 7.

5.2 Macrocondition files

The conditions comprising the macrocondition must be provided in a text file that must be located in folder Input. This file must be named using the following format: [GEO SERIES NUMBER]_-_[NAME OF MACRO CONDITION].txt. For example, the following are the contents of a macrocondition of five H3N1 subjects from GSE52428.

H3N1_001 H3N1_002 H3N1_003 H3N1_004 H3N1_005

This file can be constructed manually, or with BASH script prepare_input.sh. This option requires running the script with the following syntax.

- ./prepare_input.sh [GEO SERIES] [NAME OF MACRO CONDITION].txt Example
- ./prepare_input.sh GSE52428 H1N1.txt

The above will read ALL the conditions whose analyses have been completed for the GEO series indicated and write the file with the format described earlier.

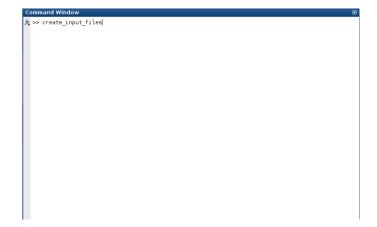


Figure 10: Calling create_input_files to create condition files.

6 Output of results

Pipeline4DGEData outputs all of its results to subfolder Output. Each pipeline analysis is carried out on an experimental condition and results from several conditions are grouped in folder by the accession number of the associated GEO series.

7 Guided creation of condition files

The (experimental) condition files for the pipeline analysis can be created manually with the format described in Section 5.1 on page 12 or with the help of script create_input_files.m. In order to do the latter, the command create_input_files must be executed on the MATLAB console from the pipeline directory.

Prior to running the pipeline analysis for the two conditions (Subsection 5.1 provides an introduction to conditions files), the corresponding condition files must be created. In order to do this, MATLAB must be started <u>inside</u> folder Pipeline4DGEData/ and in the console the command create_input_files must be called, as shown in Figure 10 and then pressing Enter.

The program will then request the accession number of the GEO series on which the analysis is to be performed. This information must be entered as single quotes ('), as shown in Figure 12. In this example, the accession number entered is GSE59015.

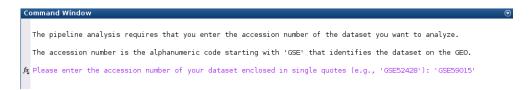


Figure 11: Entering the accession number for the creation of the condition files.

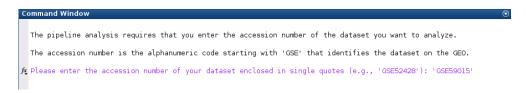


Figure 12: Entering the accession number for the creation of the condition files.

After hitting Enter, the program will read the series data from the GEO and will request the index of the column where the gene names are located in the platform record.

8 compare.m

compare.m: input is a macrocondition file. See Subsection 5.2.

9 measure_fit_of_replicates.m

This is a script that calculates the between-replicate noise for all genes across a number of replicates, as shown in Subsection 2.9. Its input is a macrocondition file, as defined in Subsection 5.2. Results will be output in folder Pipeline_HOME/Output/[GSE Series]/Comparison_of_replicates/[Macrocondition]/.

This folder contains a file named [Macrocondition] _noise_per_gene_ALL_GENES.csv. Each line of this file corresponds to a probe in the GSE matrix and contains three values separated by commas (,). The first field is the probe ID, the second is the gene name, and the third is the gene's between-replicate noise. The probes are listed in this file in the same order they appear in the GSE matrix. A secondary file, named [Macrocondition] _noise_per_gene_ALL_GENES_SORTED_BY_NOISE.csv,

provides the same information with the probes listed by their betweenreplicate noise in decreasing order. That is to say, probes that appear on top are the ones whose expression is most consistent across the probes provided in the macrocondition file whereas those at the bottom are the ones exhibiting the most discrepancy in expression across these replicates.

10 Output of the pipeline

All results are output to folder Output. Results are grouped by the *GEO* series' accession number and by the name of the experimental condition, as illustrated in Figure 13.

11 Simulation of time course data

The robustness of the pipeline process can be validated by simulating gene expression data and verifying if the pipeline returns the same model the simulated data originate from. More specifically, expression data can be simulated from given a GRN and set of GRMs and given as input to the pipeline in order to verify that the algorithm approximates the original GRN and GRMs. In order to achieve this, two types of the gene expression curves need to be generated: those from DRGs and those from non-DRGs. First, the expression curves of the DRGs are simulated from the mean curves of the GRMs as follows. Let K be the matrix representing the expression of one gene response module composed of n genes over t time points, given by

$$K = \begin{pmatrix} X_{1,1} & X_{1,2} & \cdots & X_{1,t} \\ X_{2,1} & X_{2,2} & \cdots & X_{2,t} \\ \vdots & \vdots & \ddots & \vdots \\ X_{n,1} & X_{n,2} & \cdots & X_{n,t} \end{pmatrix}.$$

Also let M be the tuple of per-column means of K, given by

$$M = \begin{pmatrix} \mu_1 & \mu_2 & \cdots & \mu_t \end{pmatrix},$$

and let S be the per-column standard deviation of K, given by

$$S = \begin{pmatrix} \sigma_1 & \sigma_2 & \cdots & \sigma_t \end{pmatrix}$$
.

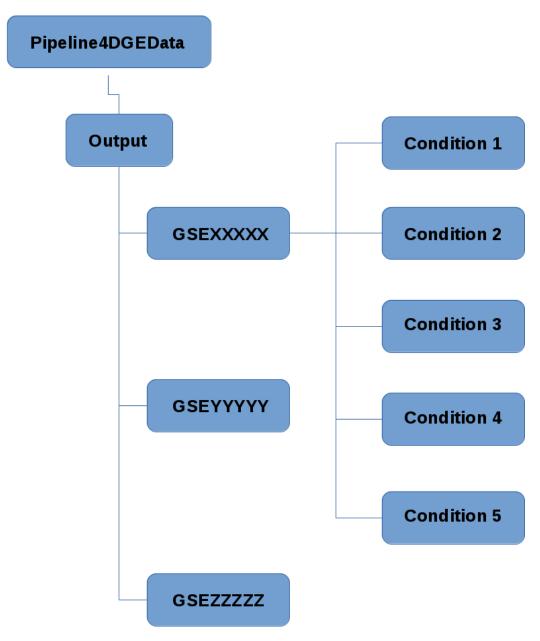


Figure 13: Folder hierarchy of results.

Then n simulated curves are obtained where the i-th element is a normal deviate of μ_i with standard deviation σ_i . This process is repeated for all the GRMs until obtaining the simulated expression curves of all the DRGs. Thereafter, the non-DRG curves are simulated as normal deviates of the pertime point means of all the DRG curves, with standard deviation 1.0. The simulated DRG and non-DRG expression curves can then be passed as arguments to the pipeline and see if the GRMs and GRN obtained approximate the originals.

The procedure described above is performed by script run_simulation.m, which reads the results of a previously-completed analysis from folder Output, constructs the simulated data and runs the pipeline in folder Output/[Condition]_simulated, where [Condition] is the name of the original experimental condition.

References

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