Discovering a motif discovering appproach

LCG BEII 2019

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Goal of the exercise

The goal of this exercise is to get an intuition of a motif discovery approach relying on the detection of over-represented oligonucleotides.

Our approach will be pragmatic.

We retrieved the upstream non-coding sequences of the genes involved in methionine biosynthesis and sulfur assimilation, and counted the occurrences of each hexanucleotide.

We also computed

- the relative frequencies (occurrences of each oligo / sum of all oligo occurrences) in the sequence of interest (the promoters of methionine-associated genes)
- the relative frequencies of ach hexanucleotide in the whole set of yeast promoters.

We would like to know if some 6nt are over-represented in promoters of methionine-associated genes relative to the occurrences that would be expected from a random selection of yeast promoters.

Create a workspace for this practical

• In your home directory, create a work directory for this practical (for example ~/LCG_BEII/practical_motif_discovery,

```
workdir <- "~/LCG_BEII/practical_motif_discovery"
dir.create(workdir, showWarnings = FALSE, recursive = TRUE)
setwd(workdir)</pre>
```

Loading the data table

1. Download the oligonucleotide count table. Scerevisiae MET-genes oligos-6nt-2str-noov occ-freq.tsv

```
oligo.url <- "http://jvanheld.github.io/LCG_BEII/practicals/motif_discovery/data/Scerevisiae_MET-genes_oligo.file <- basename(oligo.url) ## Suppress the URL path and keep only the file name for local storag download.file(oligo.url, destfile = oligo.file)
```

- 2. In **R**, open a new script or R markdown file.
- 3. Load the data table, print the 5 top rows and the 5 bottom rows.

```
oligo.table <- read.delim(oligo.file, header = 1, row.names = 1)
# View(oligo.table)</pre>
```

head(oligo.table, n = 5)

```
obs_freq exp_freq occ exp_occ aaaaaa|ttttt 0.004592808 0.004896299 41 43.71 aaaaac|gtttt 0.001120197 0.001998518 10 17.84 aaaaag|ctttt 0.003696651 0.003604251 33 32.18 aaaaat|atttt 0.004032710 0.004160627 36 37.14 aaaaca|tgttt 0.001344237 0.001932479 12 17.25
```

tail(oligo.table, n = 5)

```
    obs_freq
    exp_freq
    occ
    exp_occ

    ttccaa|ttggaa
    0.0008961577
    0.0008428396
    8
    7.52

    ttcgaa|ttcgaa
    0.0001120197
    0.0003224542
    1
    2.88

    ttgaaa|tttcaa
    0.0019043352
    0.0019087053
    17
    17.04

    ttgcaa|ttgcaa
    0.0001120197
    0.0004030214
    1
    3.60

    tttaaa|tttaaa
    0.0005600986
    0.0009379354
    5
    8.37
```

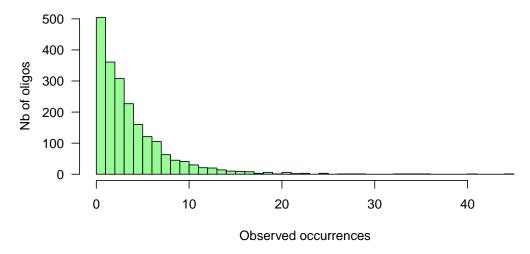
Exploring observed and expected counts

4. Draw an histogram of the observed occurrences and evaluate the spread of counts.

```
x <- oligo.table$occ
range(x)</pre>
```

[1] 0 45

Distribution of oligonucelotide occurrences



5. Draw a scatter plot comparing the observed and expected occurrences for each hexanucleotide.

Observed vs expected occurrences

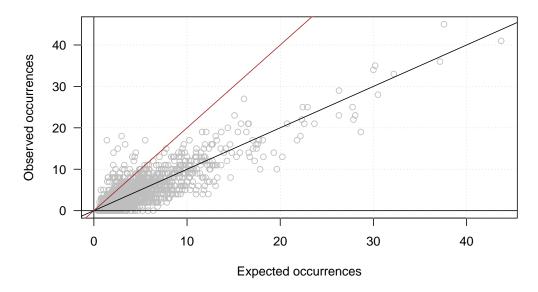


Figure 1: **Scatter plot of observed versus expected occurrences.** The black diagonal corresponds to the null hypothesis, the brown line denotes an arbitrary threshold on fold-change > 2.

6. Compute the ratio of observed / expected occurrences, and draw a scatter plot with this ratio (Y) as a function of the expected occurrences (X).

6. Compute the log-ratio of observed / expected occurrences, and draw a scatter plot with this log-ratio (Y) as a function of the expected occurrences (X).

(obs/exp) ratio

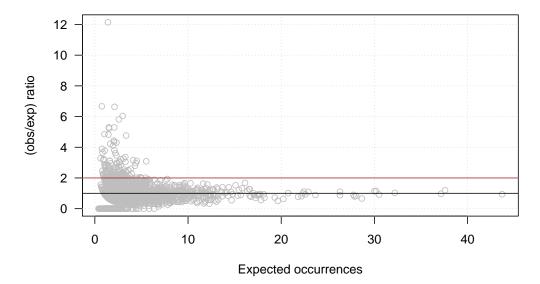


Figure 2: **Scatter plot of observed versus expected occurrences.** The black diagonal corresponds to the null hypothesis, the brown line denotes an arbitrary threshold on fold-change > 2.

$$lr = log(x/\langle X \rangle)$$

7. Compute the log-likelihood ratio (llr), defined below, and draw a scatter plot with this llr as a function of the expected occurrences.

$$llr = f \cdot log(x/ < X >)$$

Log-ratio

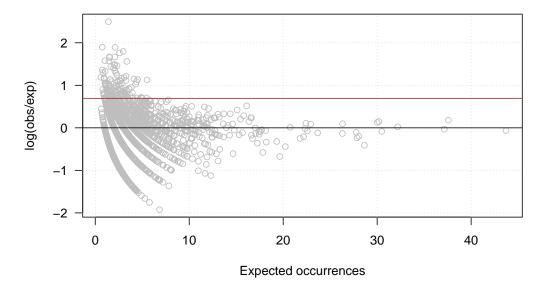


Figure 3: **Scatter plot of observed versus expected occurrences.** The black diagonal corresponds to the null hypothesis, the brown line denotes an arbitrary threshold on fold-change > 2.

```
abline(h = 0, col = "black")
# abline(h = log(2), col = "brown")
```

Computing over-representation significance

8. Draw a binomial distribution with parameters n = 8000, p = 0.0001.

Log-likelihood ratio

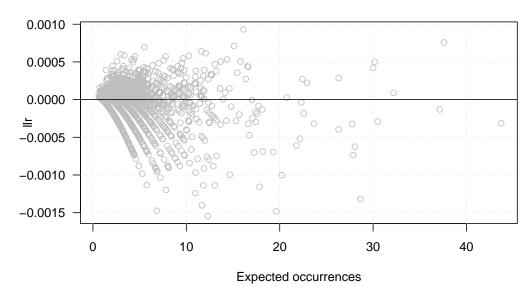
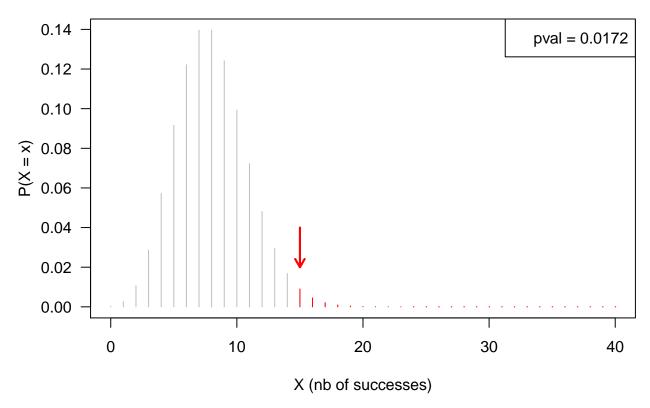


Figure 4: **Scatter plot of log-likelihood ratio (llr) versus expected occurrences.** The black line corresponds to the null hypothesis, the brown line denotes an arbitrary threshold on fold-change > 2.



8. Use the binomial distribution to compute the P-value of the observed occurrences.

$$P = T(X \ge x)$$

x <- oligo.table\$occ ## Nuumber of successes
n <- sum(x) ## Number of trials</pre>

P-value histogram

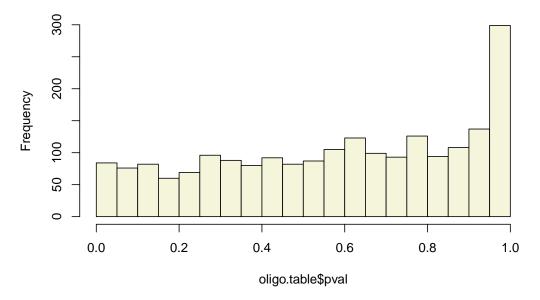


Figure 5: **Histogram of nominal p-values** for all the hexanucleotides grouped by pairs of reverse complements.

```
p <- oligo.table$exp_freq ## Success probability

p <- p / sum(p) # A correction for the fact that we discarded self-overlapping occurrences
# sum(p)

nbTests <- length(x) # Number of tests

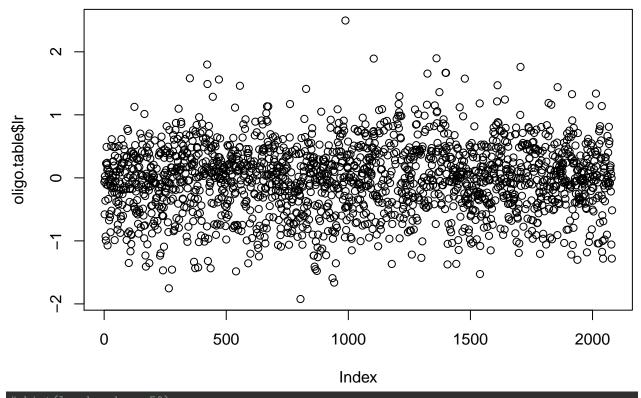
## Compute a P-value for each individual oligonucleotide
oligo.table$pval <- pbinom(q = x - 1, size = n, prob = p, lower.tail = FALSE)</pre>
```

9. Draw an histogram with the P-values of all hexanucleotides, with 20 bins.

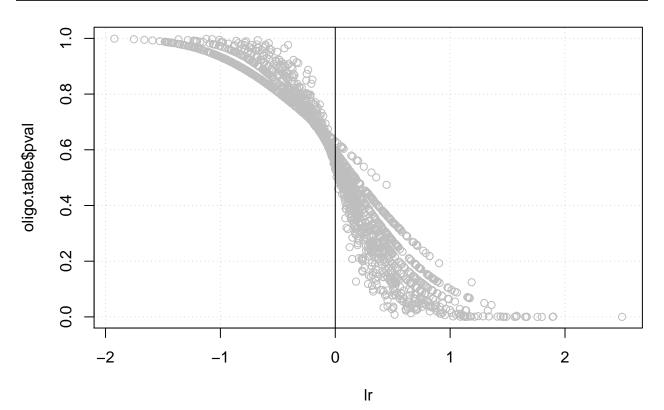
```
hist(oligo.table$pval, breaks = seq(0, 1, 0.05),
col = "beige", main = "P-value histogram")
```

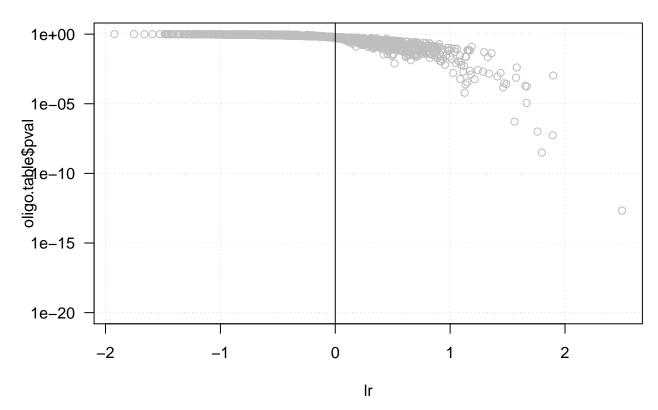
10. Draw a scatter plot with the P-value (Y) as a function of the log-ratio (X).

```
plot(oligo.table$lr)
```







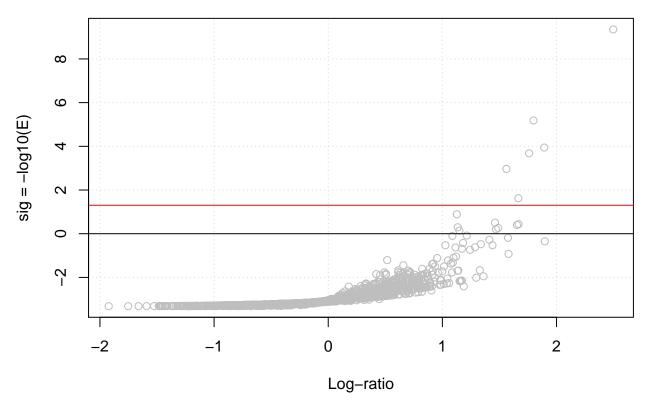


11. Compute the E-value, and the significance.

$$E = P \cdot N$$
$$sig = -log_{10}(E)$$

```
oligo.table$eval <- oligo.table$pval * nbTests
oligo.table$sig <- -log10(oligo.table$eval)
```

12. Draw a **Volcano plot**, with the significance as a function of the log-ratio.

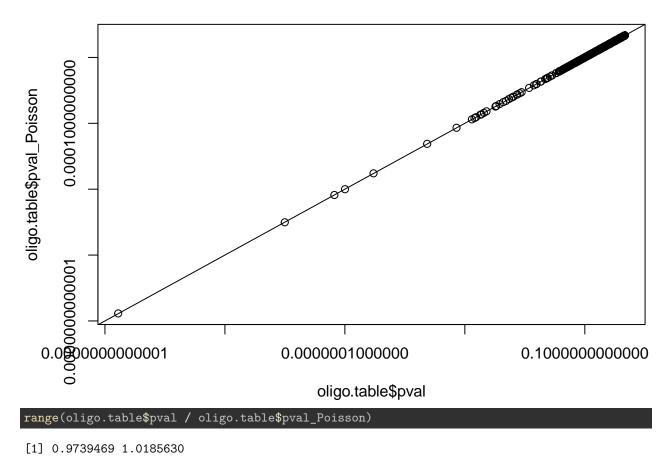


13. Compute the P-value using the Poisson distribution as approximation of the binomial. Are we in suitable conditions for this approximation? Draw a plot comparing the P-values obtained by the binomial and Poisson distributions.

```
lambda <- oligo.table$exp_occ * sum(oligo.table$occ) / sum(oligo.table$exp_occ)

oligo.table$pval_Poisson <- ppois(
    q = oligo.table$occ - 1, lambda = lambda, lower.tail = FALSE)

plot(oligo.table$pval, oligo.table$pval_Poisson,
    log = "xy")
abline(a = 0, b = 1)</pre>
```



[1] 0.9739469 1.0185630