

One-to-N sequence matches in coronavirus genomes

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```
#### General parameters for the analysis ####

reloadImage <- TRUE

#### Sequence collection ####

## Supported:
collections <- c(
  "around-CoV-2",
  "selected",
  "around-CoV-2-plus-GISAID",
  "selected-plus-GISAID"
)

# collection <- "around-CoV-2" # 14 genomes
# collection <- "around-CoV-2-plus-GISAID" # 16 genomes
# collection <- "selected" # ~60 genomes
collection <- "selected-plus-GISAID" # ~60 genomes

## Note about GIDAID sequences.
##
## A few genomes were not available in NCBI Genbank at the time of
## this analysis, and had to be downloaded from GISAID. These sequences
## can however not be redistributed, they should thus be downloaded
## manually to reproduce the full trees.

## Exclude incomplete genomes (i.e. those containing a lof of Ns) to avoid biases in the distance compu
excludeIncomplete <- TRUE
```

```

#### Define directories and files ####
dir <- list(main = '..')
dir$R <- file.path(dir$main, "scripts/R")

#### Create output directory for sequences ####
dir$seqdata <- file.path(dir$main, "data")
dir.create(dir$seqdata, showWarnings = FALSE, recursive = TRUE)

## Instantiate a list for output files
outfiles <- vector()

## Memory image
dir$images <- file.path(dir$main, "memory_images")
dir.create(dir$images, recursive = TRUE, showWarnings = FALSE)
outfiles["Memory image"] <- file.path(dir$images, "one-to-n_matches.Rdata")

## Input files
infiles <- list()

## Output tables
# di$output <- file.path(dir$main, "")
# dir$tables <-

## Load custom functions
source(file.path(dir$R, "align_n_to_one.R"))
source(file.path(dir$R, "plot_pip_profiles.R"))

## A unequivocal pattern to identify the reference genome in the sequence names of the input file
refPattern <- "HuCoV2_WH01_2019"

## Exclude some genomes with a lot of Ns, because they bias the PIP profiles and alignments and trees
excludePatterns <- c("PnMP789", "PnGu-P2S_2019")

#### Features of interest in the reference genome ####

features <- list()

## Specific features

## S1: pre-cleavage part of the spike protein
features[["S1"]] <- c(start = 21599, end = 23617)
## S2: post-cleavage part of the spike protein
features[["S2"]] <- c(start = 23618, end = 25381)
## Receptor Binding Domain (RBD)
features[["RBD"]] <- c(start = 22517, end = 23185)
## Potential Pangolin origin after Xiao (https://doi.org/10.1101/2020.02.17.951335)
features[["Recomb-Xiao"]] <- c(start = 22871, end = 23092)
## Recombinant region 1 seen on the PIP profiles
features[["Recomb-reg-1"]] <- c(start = 21500, end = 22800)
## Recombinant region 2 seen on the PIP profiles
features[["Recomb-reg-2"]] <- c(start = 22800, end = 24000)

```

```

## Recombinant region 3 seen on the PIP profiles
features[['Recomb-reg-3']] <- c(start = 27800, end = 28350)

## Annotated coding sequences
features[['CDS-S']] <- c(start = 21563, end = 25384) ## Spike gene
features[['CDS-ORF3a']] <- c(start = 25393, end = 26220)
features[['CDS-E']] <- c(start = 26245, end = 26472)
features[['CDS-M']] <- c(start = 26523, end = 27191)
features[['CDS-ORF6']] <- c(start = 27202, end = 27387)
features[['CDS-ORF7a']] <- c(start = 27394, end = 27759)
features[['CDS-ORF8']] <- c(start = 27894, end = 28259)
features[['CDS-N']] <- c(start = 28274, end = 29533)
features[['CDS-ORF10']] <- c(start = 29558, end = 29674)
features[['CDS-ORF1ab']] <- c(start = 266, end = 21555)

## All the sequences after the bif ORF coding for the polyprotein 1ab
features[['After-ORF1ab']] <- c(start = 21556, end = 29899)

## Report the parameters
message("\tReference genomes: ", refPattern)

## Genome dir and files
if (length(grep(pattern = "GISAID", x = collection)) > 0) {
  useGISAID <- TRUE
  dir$genomes <- file.path(dir$main, "data", "GISAID_genomes")
  # collections <- paste0(collections, "-plus-GISAID")
  # collection <- paste0(collection, "-plus-GISAID")
} else {
  dir$genomes <- file.path(dir$main, "data", "genomes")
}

## Define the input genome
infile$genomes <- file.path(
  dir$genomes,
  paste0("genomes_", collection, ".fasta"))

## Genome sequences
if (!file.exists(infile$genomes)) {
  stop("Genome sequence file is missing", "\n", infile$genomes)
}

#### Load genome sequences ####
genomes <- readDNASTringSet(filepath = infile$genome, format = "fasta")

## Shorten sequence names by suppressing the fasta comment (after the space)
names(genomes) <- sub(pattern = " .*", replacement = "", x = names(genomes), perl = TRUE)

## Exclude genomes
if (excludeIncomplete) {
  excludePattern = paste0("(", paste(collapse = ")|(", excludePatterns), ")")
  excludedGenomeNames <- grep(pattern = excludePattern, x = names(genomes),

```

```

        value = TRUE, invert = FALSE)
filteredGenomeIndices <- grep(pattern = excludePattern, x = names(genomes),
                             value = FALSE, invert = TRUE)
message("\tExcluded ", length(excludedGenomeNames), " genomes: ", paste(collapse = ", ", excludedGenomeNames))
message("\tRemaining genomes: ", length(filteredGenomeIndices))
genomes <- genomes[filteredGenomeIndices]
# names(genomes)
}

## Report the number of genomes
genomeNames <- names(genomes)
nbGenomes <- length(genomeNames)
message("\tLoaded ", nbGenomes, " genomes from file ", infiles$genomes)
# View(genomes)

#### Define reference and query genomes ####
refGenomeName <- grep(pattern = refPattern, x = names(genomes),
                     ignore.case = TRUE, value = TRUE)
if (is.null(refGenomeName)) {
  stop("Could not identify reference genome with pattern ", refPattern)
}
message("\tReference genome name: ", refGenomeName)

## Compute some statistics about genome sizes
genomeStat <- data.frame(
  n = 1:length(genomes),
  row.names = names(genomes),
  status = rep("Query", length.out = length(genomeNames))
)
genomeStat[,"status"] <- as.vector(genomeStat[,"status"])
genomeStat[refGenomeName,"status"] <- "Reference"
g <- 1
for (g in genomeNames) {
  genomeStat[g, "length"] <- length(genomes[[g]])
}

#### Define the color associated to each sequence ####

## Color palette per species
speciesPalette <- list(
  Human = "#880000",
  Bat = "#888888",
  Pangolin = "#448800",
  Camel = "#BB8800",
  Pig = "#FFBBBB",
  Civet = "#00BBFF"
)

## Species prefix in the tip labels
speciesPrefix <- c("Hu" = "Human",
                  "Bt" = "Bat",
                  "Pn" = "Pangolin",
                  "Cm" = "Camel",

```

```

        "Pi" = "Pig",
        "Cv" = "Civet")

## Strain-specific colors
strainColor <- c(
  "HuCoV2_WH01_2019" = "red",
  "HuSARS-Frankfurt-1_2003" = "#0044BB",
  "PnGu1_2019" = "#00BB00",
  "BtRaTG13_" = "#FF6600",
  "BtYu-RmYN" = "#FFBB22",
  "BtZXC21" = "black",
  "BtZC45" = "black")

## Identify species per tip
for (prefix in names(speciesPrefix)) {
  genomeStat[grepl(pattern = paste0("^", prefix), x = row.names(genomeStat), perl = TRUE), "species"] <-
}

## Assign a color to each species
genomeStat$color <- "grey" # default
genomeStat$color <- speciesPalette[as.vector(genomeStat$species)]

for (strain in names(strainColor)) {
  genomeStat[grepl(pattern = paste0("^", strain),
    x = row.names(genomeStat), perl = TRUE), "color"] <- strainColor[strain]
}

## Assign specific color to some nodes

## Define a color for each genome
genomeColors <- (unlist(genomeStat$color))
names(genomeColors) <- row.names(genomeStat)

```

Parameters

```

## Define a list of parameters
parameters <- list()
parameters$collection <- collection
parameters$refGenomeName <- refGenomeName
parameters$nbGenomes <- nbGenomes
parameters$genomeDir <- dir$genomes
parameters$genomeFile <- infiles$genomes
kable(t(as.data.frame.list(parameters)), caption = "Parameters of the analysis",
  col.names = "Parameter")

```

Table 1: Parameters of the analysis

	Parameter
collection	selected-plus-GISAID
refGenomeName	HuCoV2_WH01_2019

	Parameter
nbGenomes	38
genomeDir	../data/GISAID_genomes
genomeFile	../data/GISAID_genomes/genomes_selected-plus-GISAID.fasta

Genome statistics

```
kable(genomeStat, caption = "Reference and query genomes")
```

Table 2: Reference and query genomes

	n	status	length	species	color
BtBM48-31	1	Query	29276	Bat	#888888
BtBtKY72	2	Query	29274	Bat	#888888
BtCp-Yun_2011	3	Query	29452	Bat	#888888
BtGX2013	4	Query	29161	Bat	#888888
BtHKU3-12	5	Query	29704	Bat	#888888
BtHKU5	6	Query	30482	Bat	#888888
BtHKU9-1	7	Query	29114	Bat	#888888
BtJL2012	8	Query	29037	Bat	#888888
BtLYRa11	9	Query	29805	Bat	#888888
BtRaTG13_2013_Yunnan	10	Query	29855	Bat	#FF6600
Btrec-SARSG_2008	11	Query	29750	Bat	#888888
BtRm1/2004	12	Query	29749	Bat	#888888
BtRp-Shaanxi2011	13	Query	29484	Bat	#888888
BtRp3-2004	14	Query	29736	Bat	#888888
BtRs_672-2006	15	Query	29059	Bat	#888888
BtRs4874	16	Query	30311	Bat	#888888
BtSC2018	17	Query	29648	Bat	#888888
BtYN2013	18	Query	29142	Bat	#888888
BtYN2018B	19	Query	30256	Bat	#888888
BtYN2018C	20	Query	29689	Bat	#888888
BtYNLF_31C	21	Query	29723	Bat	#888888
BtZC45	22	Query	29802	Bat	black
BtZXC21	23	Query	29732	Bat	black
CmMERS	24	Query	29851	Camel	#BB8800
Cv007-2004	25	Query	29540	Civet	#00BBFF
Hu229E	26	Query	27317	Human	#880000
HuCoV2_WH01_2019	27	Reference	29899	Human	red
HuMERS_172-06_2015	28	Query	30068	Human	#880000
HuNL63	29	Query	27553	Human	#880000
HuOC43	30	Query	30741	Human	#880000
HuSARS-Frankfurt-1_2003	31	Query	29727	Human	#0044BB
HuTGEV	32	Query	28586	Human	#880000
PiPRCV	33	Query	27765	Pig	#FFBBBB
PiSADS	34	Query	27163	Pig	#FFBBBB
PnGX-P1E_2017	35	Query	29801	Pangolin	#448800
PnGX-P2V_2018	36	Query	29795	Pangolin	#448800
BtYu-RmYN02_2019	37	Query	29671	Bat	#FFBB22
PnGu1_2019	38	Query	29825	Pangolin	#00BB00

The collection selected-plus-GISAID contains 38 virus genome sequences.

One-to-N alignemnts of selected features

We perform a global pairwise alignment (Needle-Waterman algorithm) between each feature of the reference (HuCoV2_WH01_2019) and each one of the query genomes.

```
##### One-to-N alignmemnt of user-specified genomic features #####
# featureName <- "CDS-S" # for the test
featureName <- "S1" # for the test
# featureName <- "Recomb-reg-3" # for the test
# featureName <- "RBD" # for the test
# for (collection in (collections)) {
if (reloadImage) {
  load(outfiles["Memory image"])
  reloadImage <- TRUE
} else {
  allFeatureAlignments <- list()
}
for (featureName in names(features)) {
  featureLimits <- features[[featureName]]
  featureStart <- featureLimits[["start"]]
  featureEnd <- featureLimits[["end"]]
  featureLength <-
    featureEnd - featureStart + 1

  message("Searching matches for feature ", featureName,
    " (", featureLimits[1], "-", featureLimits[2], ")",
    " in collection ", collection)

  ##### N-to-1 alignments of spike-coding sequences #####
  featurePrefix <- paste0(
    featureName,
    "_", collection)

  dir[[featureName]] <- file.path(dir$seqdata, featureName)
  dir.create(dir[[featureName]], showWarnings = FALSE, recursive = TRUE)
  outfiles[featureName] <- file.path(
    dir[[featureName]], paste0(featurePrefix, ".fasta"))
  message("\tOutput directory: ", dir[[featureName]] )

  ## Get sequence of the feature from the reference genome
  refSeq <- subseq(genomes[refGenomeName],
    start = featureLimits[1],
    end = featureLimits[2])

  ## Match the reference feature with all the genomes
  if (reloadImage) {
    featureAlignmentsNto1 <- allFeatureAlignments[[featureName]]
  } else {
    featureAlignmentsNto1 <- alignNtoOne(
      refSequence = refSeq,
      querySequences = genomes,
      type = "global-local",
```

```

        outfile = outfiles[featureName])

    allFeatureAlignments[[featureName]] <- featureAlignmentsNto1
}

genomeOrder <- order(featureAlignmentsNto1$stat$score, decreasing = TRUE)

if (featureLength > 20000) {
    windowSize <- 800
} else if (featureLength > 5000) {
    windowSize <- 500
} else if (featureLength > 3000) {
    windowSize <- 200
} else {
    windowSize <- max(100, 10 * round(featureLength / 100))
}

## PIP profile of one-to-N alignment
plotPIPprofiles(
    alignments = featureAlignmentsNto1$alignments[genomeOrder],
    reversePlot = TRUE,
    windowSize = windowSize,
    main = paste0(featureName, " (", featureStart, ":", featureEnd, ", ", featureLength, " bp)",
    # colors = NULL,
    colors = genomeColors,
    legendMargin = 0.25,
    legendCorner = "topright", lwd = 1,
    legendCex = 0.5, ylim = c(30, 100))

kable(featureAlignmentsNto1$stats[genomeOrder,],
    caption = paste0(
        "One-to-N alignment of feature ",
        featureName))
}

# }

```

Output files

```
kable(t(as.data.frame(dir)), col.names = "Dir", caption = "Directories")
```

Table 3: Directories

	Dir
main	..
R	../scripts/R
seqdata	../data
images	../memory_images
genomes	../data/GISAID_genomes
S1	../data/S1
S2	../data/S2

	Dir
RBD	../data/RBD
Recomb.Xiao	../data/Recomb-Xiao
Recomb.reg.1	../data/Recomb-reg-1
Recomb.reg.2	../data/Recomb-reg-2
Recomb.reg.3	../data/Recomb-reg-3
CDS.S	../data/CDS-S
CDS.ORF3a	../data/CDS-ORF3a
CDS.E	../data/CDS-E
CDS.M	../data/CDS-M
CDS.ORF6	../data/CDS-ORF6
CDS.ORF7a	../data/CDS-ORF7a
CDS.ORF8	../data/CDS-ORF8
CDS.N	../data/CDS-N
CDS.ORF10	../data/CDS-ORF10
CDS.ORF1ab	../data/CDS-ORF1ab
After.ORF1ab	../data/After-ORF1ab

```

outfileTable <- data.frame(path = as.vector(outfiles))
outfileTable$basename <- basename(as.vector(outfileTable$path))
outfileTable$dir <- dirname(as.vector(outfileTable$path))
outfileTable$link <- paste0(
  "[", outfileTable$basename, "]"(", outfileTable$path, ")")
)

kable(outfileTable[, c("dir", "link")],
  col.names = c("dir", "file"),
  caption = "Output files")

```

Table 4: Output files

dir	file
../memory_images	one-to-n_matches.Rdata
../data/S1	S1_selected-plus-GISAID.fasta
../data/S2	S2_selected-plus-GISAID.fasta
../data/RBD	RBD_selected-plus-GISAID.fasta
../data/Recomb-Xiao	Recomb-Xiao_selected-plus-GISAID.fasta
../data/Recomb-reg-1	Recomb-reg-1_selected-plus-GISAID.fasta
../data/Recomb-reg-2	Recomb-reg-2_selected-plus-GISAID.fasta
../data/Recomb-reg-3	Recomb-reg-3_selected-plus-GISAID.fasta
../data/CDS-S	CDS-S_selected-plus-GISAID.fasta
../data/CDS-ORF3a	CDS-ORF3a_selected-plus-GISAID.fasta
../data/CDS-E	CDS-E_selected-plus-GISAID.fasta
../data/CDS-M	CDS-M_selected-plus-GISAID.fasta
../data/CDS-ORF6	CDS-ORF6_selected-plus-GISAID.fasta
../data/CDS-ORF7a	CDS-ORF7a_selected-plus-GISAID.fasta
../data/CDS-ORF8	CDS-ORF8_selected-plus-GISAID.fasta
../data/CDS-N	CDS-N_selected-plus-GISAID.fasta
../data/CDS-ORF10	CDS-ORF10_selected-plus-GISAID.fasta
../data/CDS-ORF1ab	CDS-ORF1ab_selected-plus-GISAID.fasta
../data/After-ORF1ab	After-ORF1ab_selected-plus-GISAID.fasta

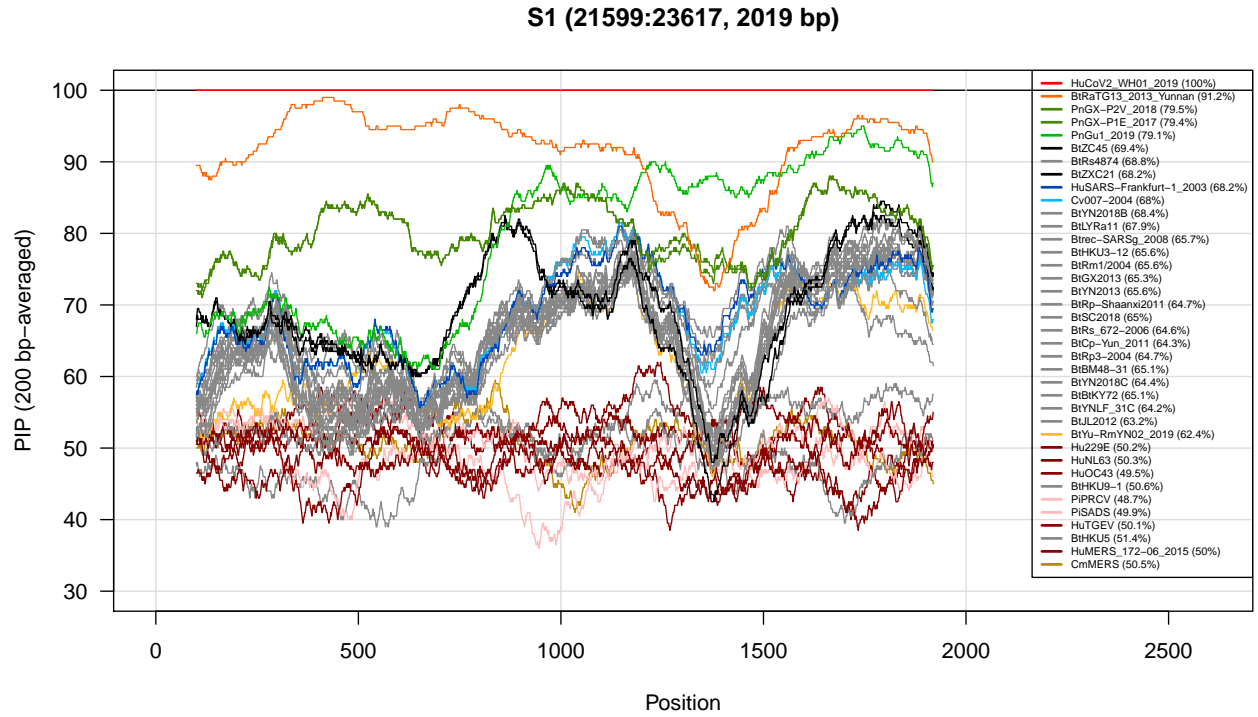


Figure 1: Feature-specific Percent Identical Positions (PIP) profiles.

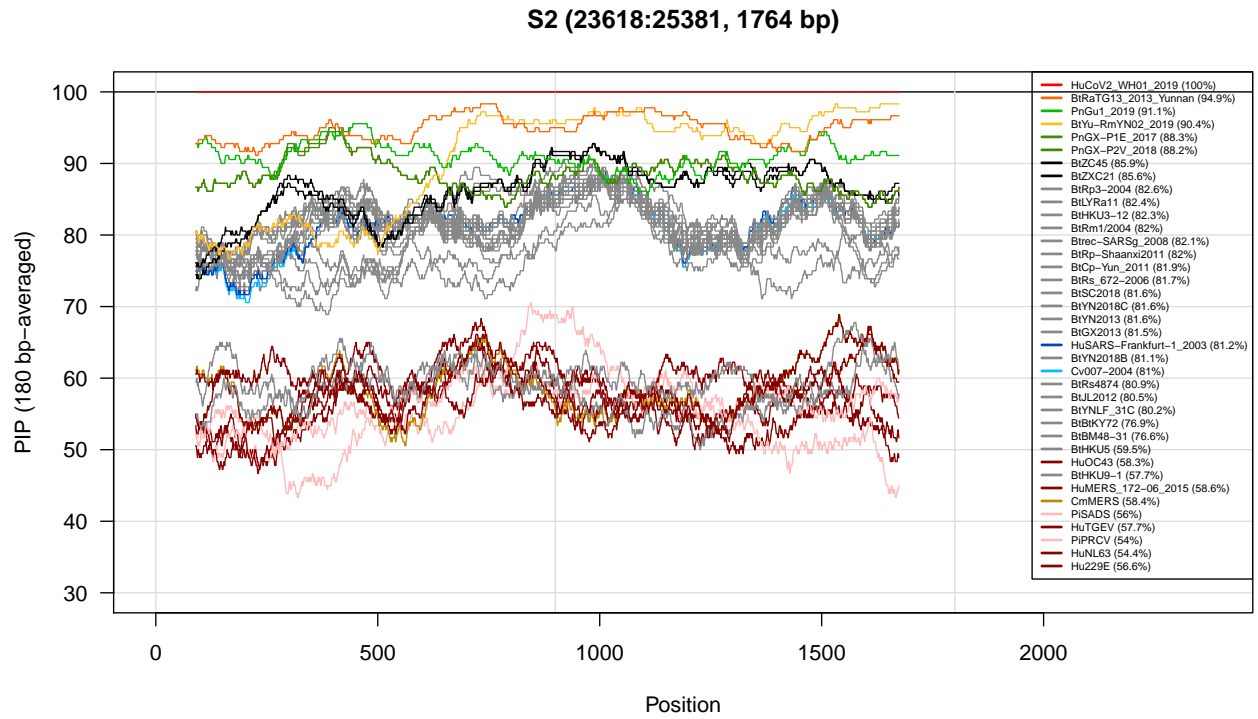


Figure 2: Feature-specific Percent Identical Positions (PIP) profiles.

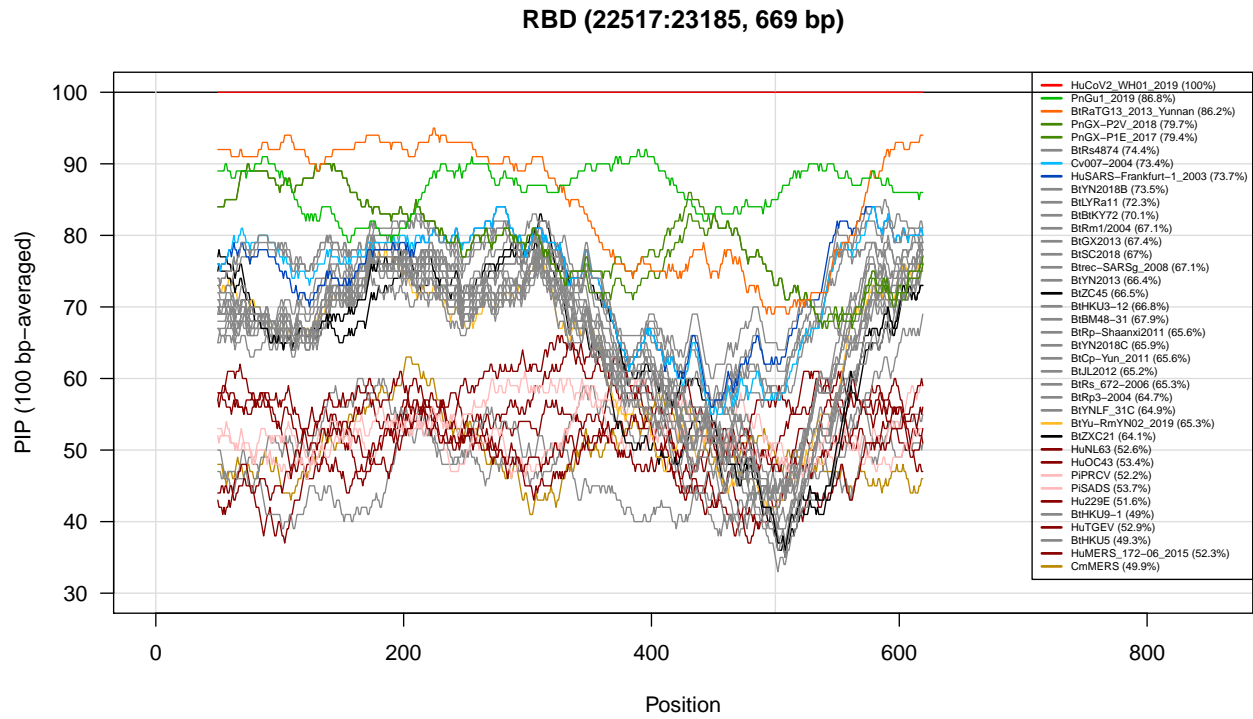


Figure 3: Feature-specific Percent Identical Positions (PIP) profiles.

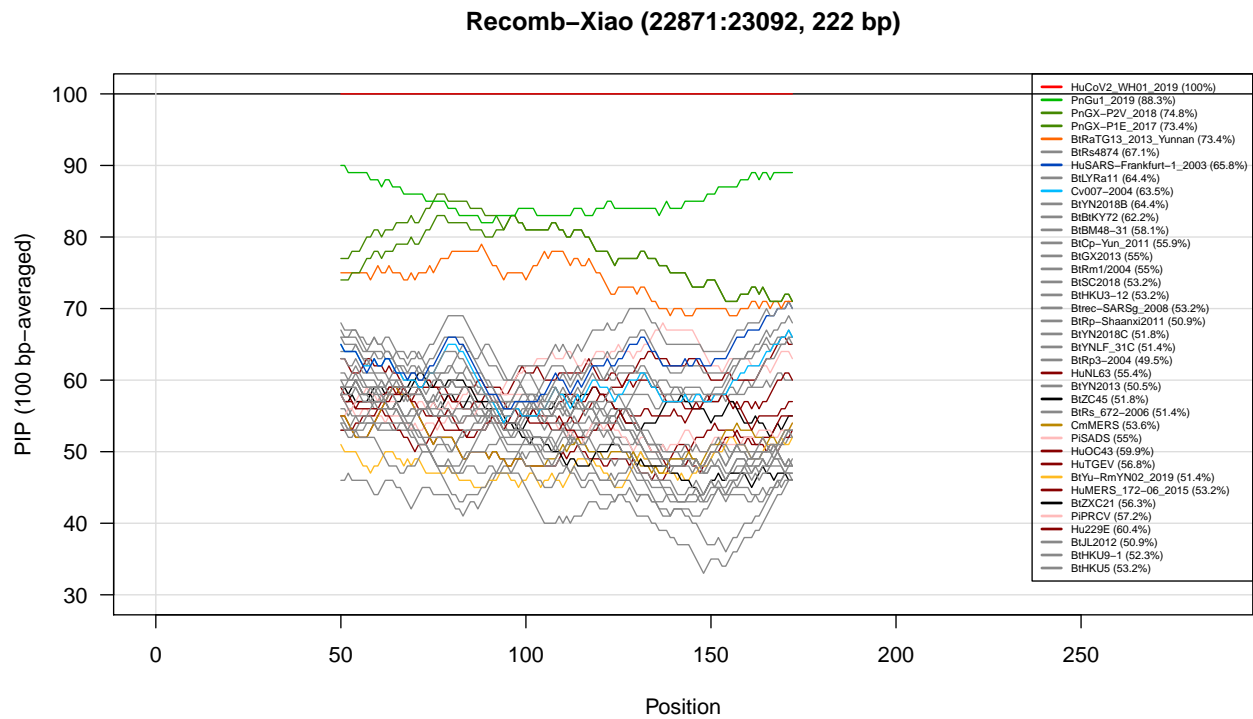


Figure 4: Feature-specific Percent Identical Positions (PIP) profiles.

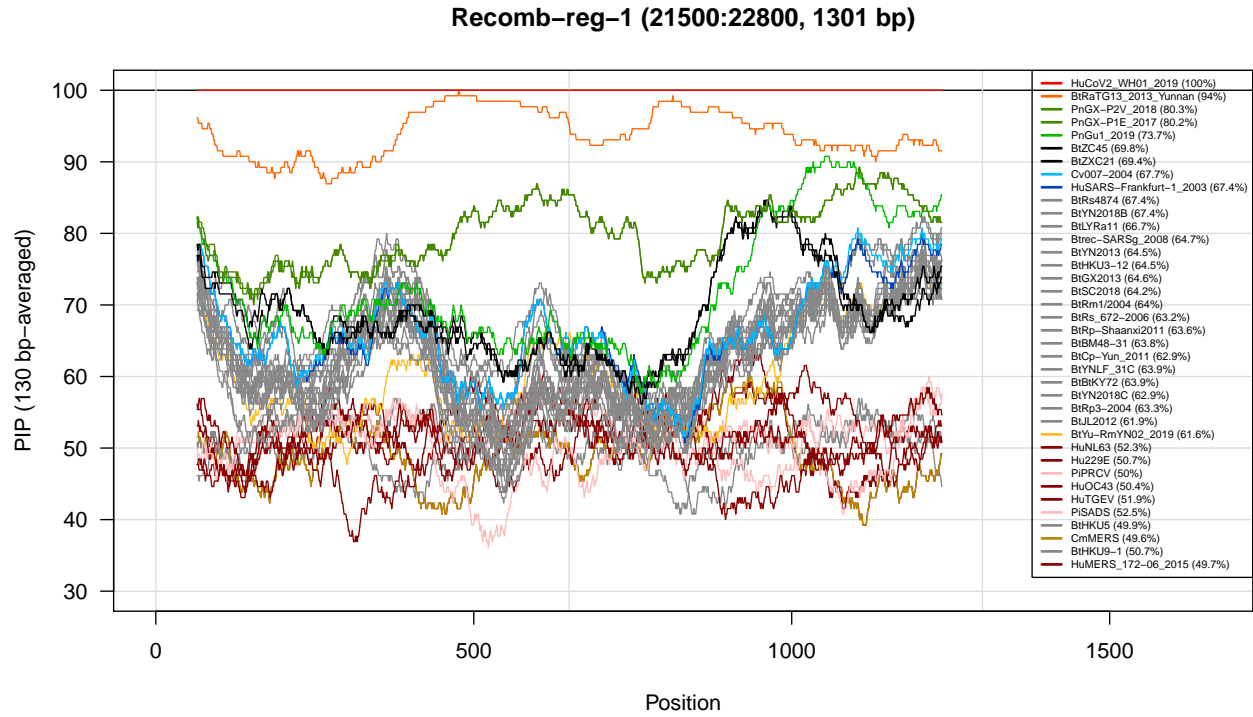


Figure 5: Feature-specific Percent Identical Positions (PIP) profiles.

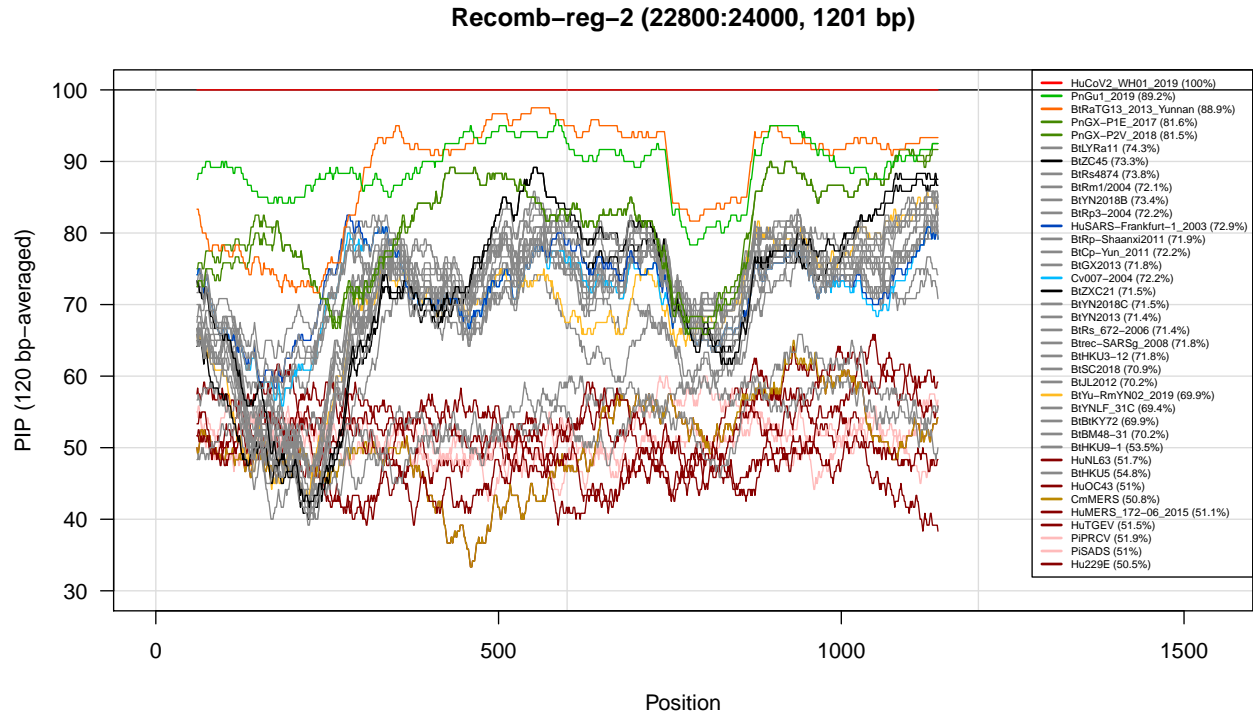


Figure 6: Feature-specific Percent Identical Positions (PIP) profiles.

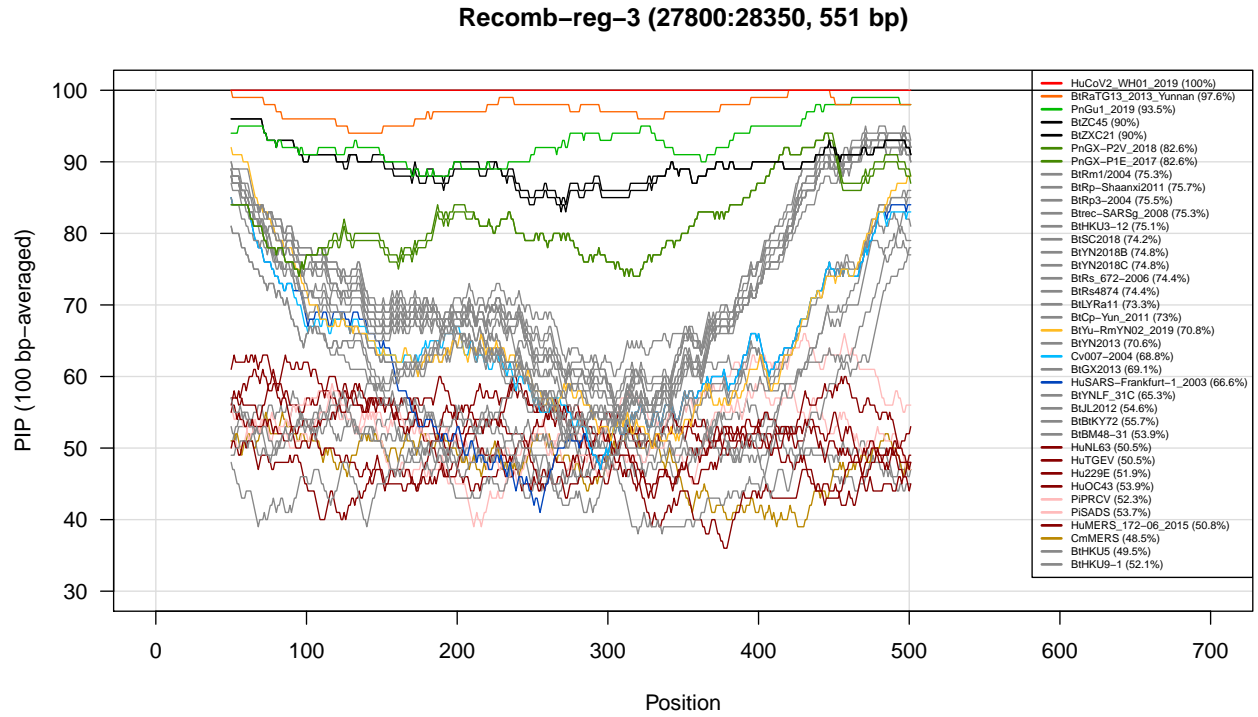


Figure 7: Feature-specific Percent Identical Positions (PIP) profiles.

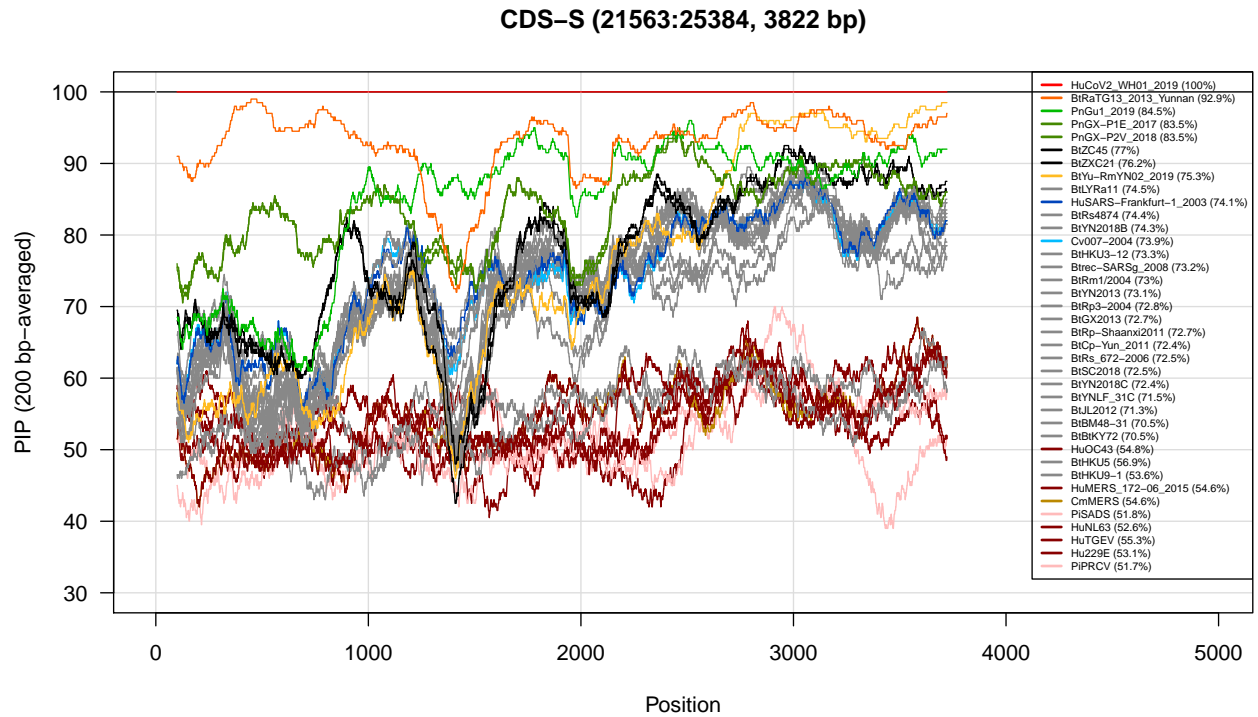


Figure 8: Feature-specific Percent Identical Positions (PIP) profiles.

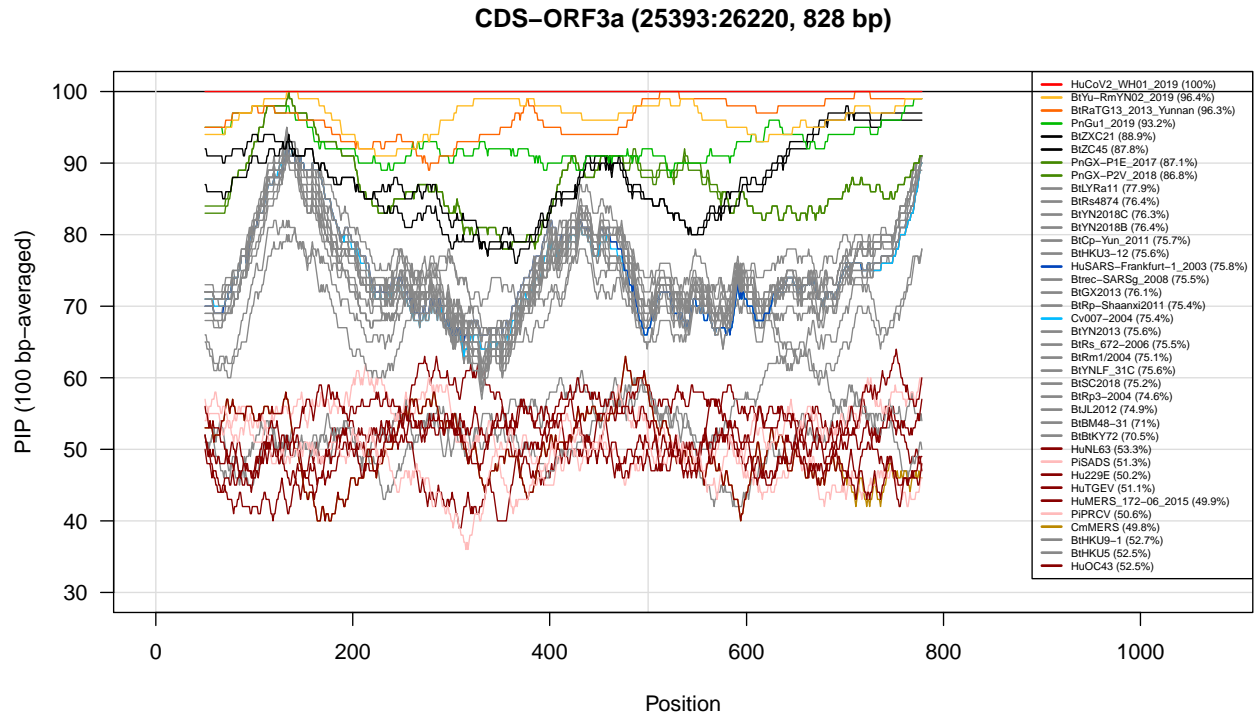


Figure 9: Feature-specific Percent Identical Positions (PIP) profiles.

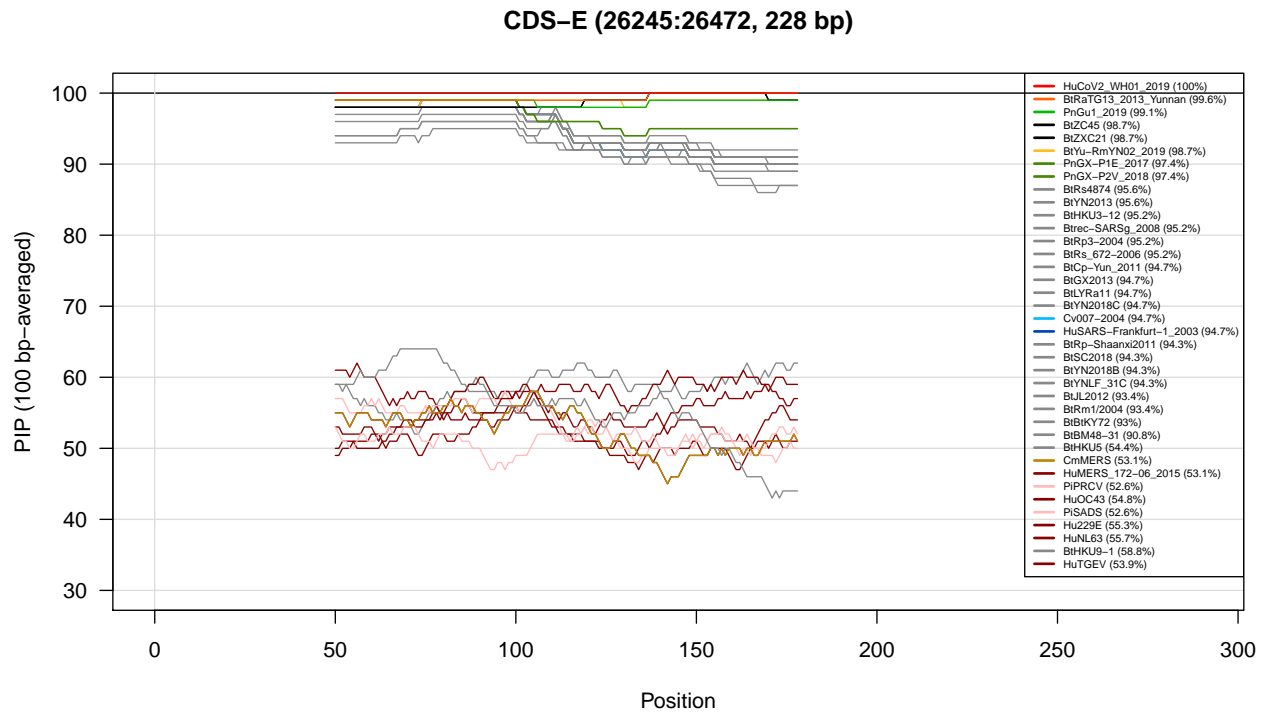


Figure 10: Feature-specific Percent Identical Positions (PIP) profiles.

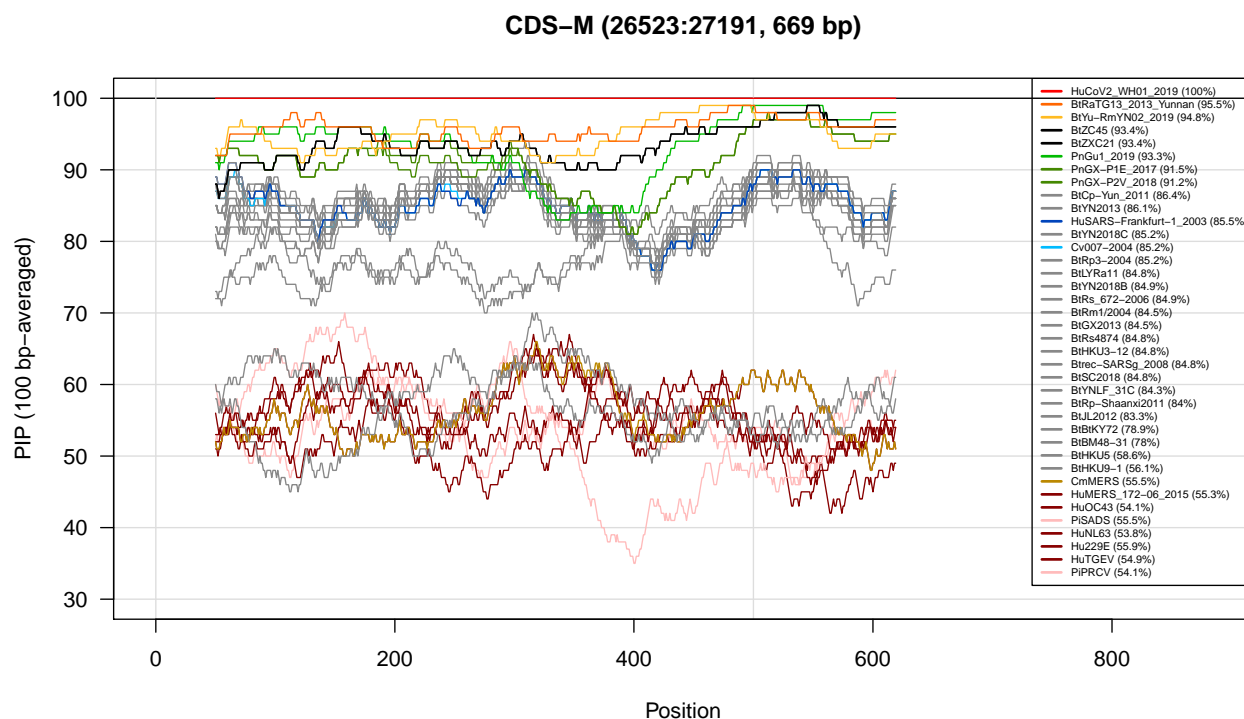


Figure 11: Feature-specific Percent Identical Positions (PIP) profiles.

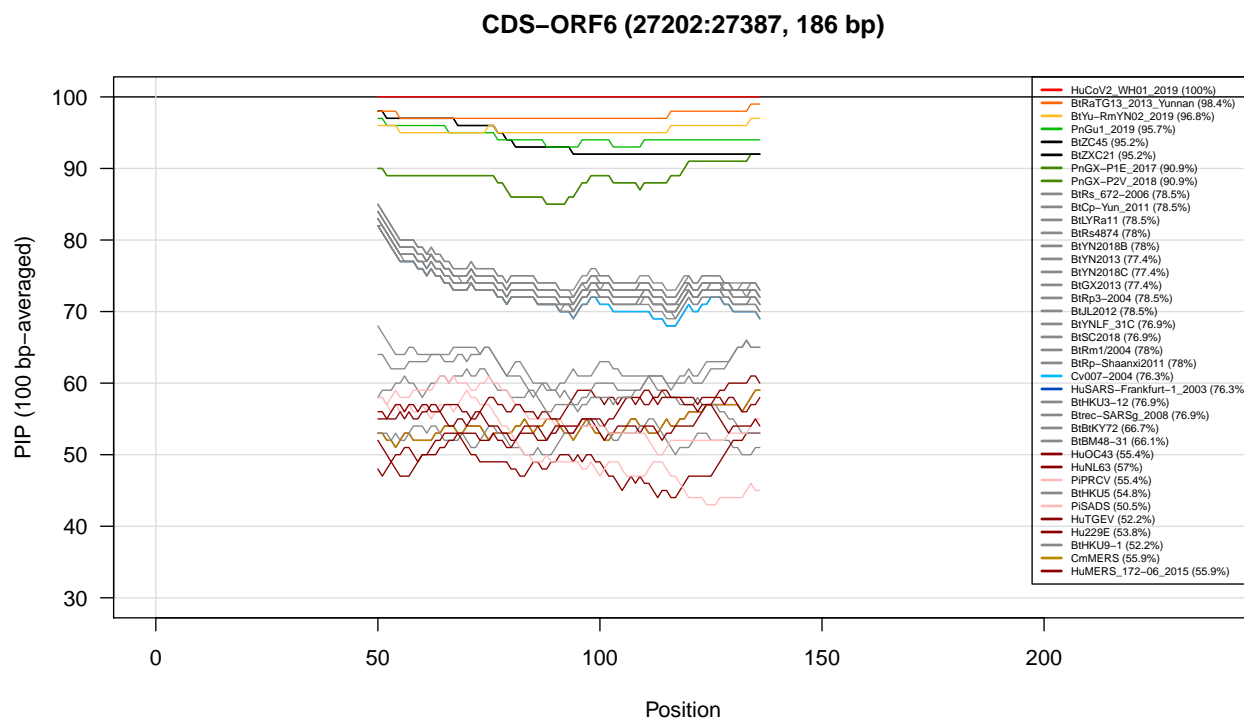


Figure 12: Feature-specific Percent Identical Positions (PIP) profiles.

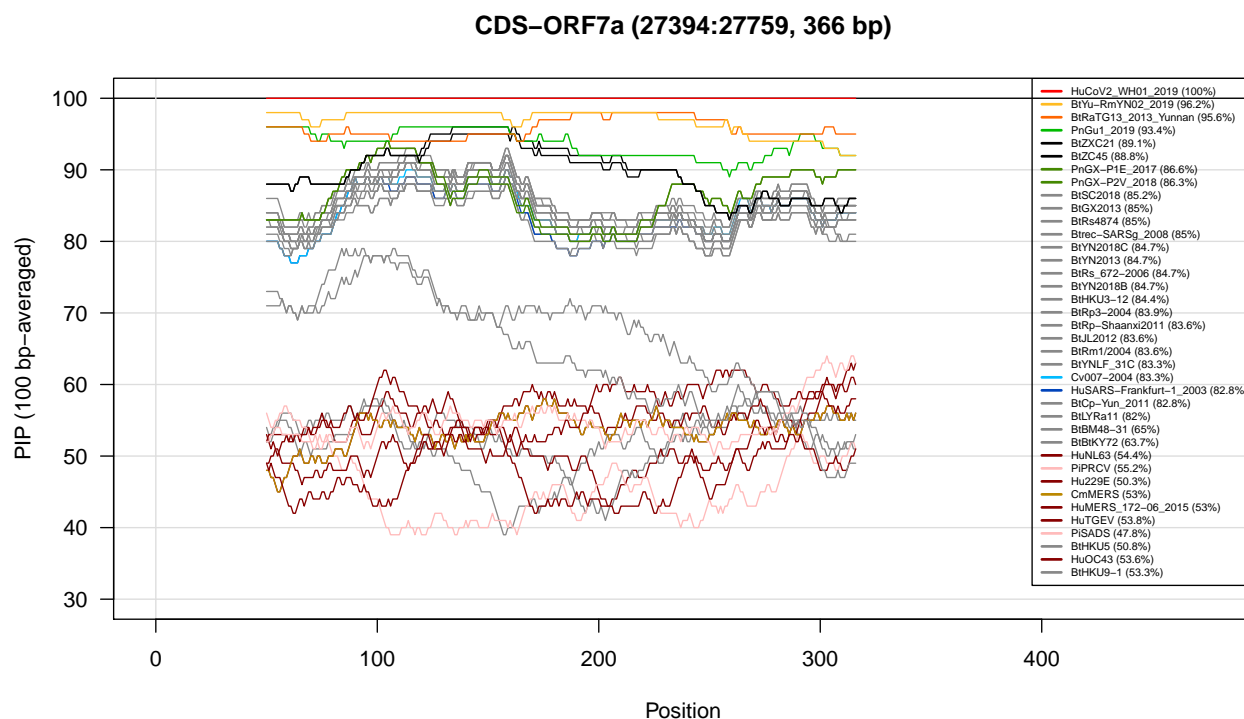


Figure 13: Feature-specific Percent Identical Positions (PIP) profiles.

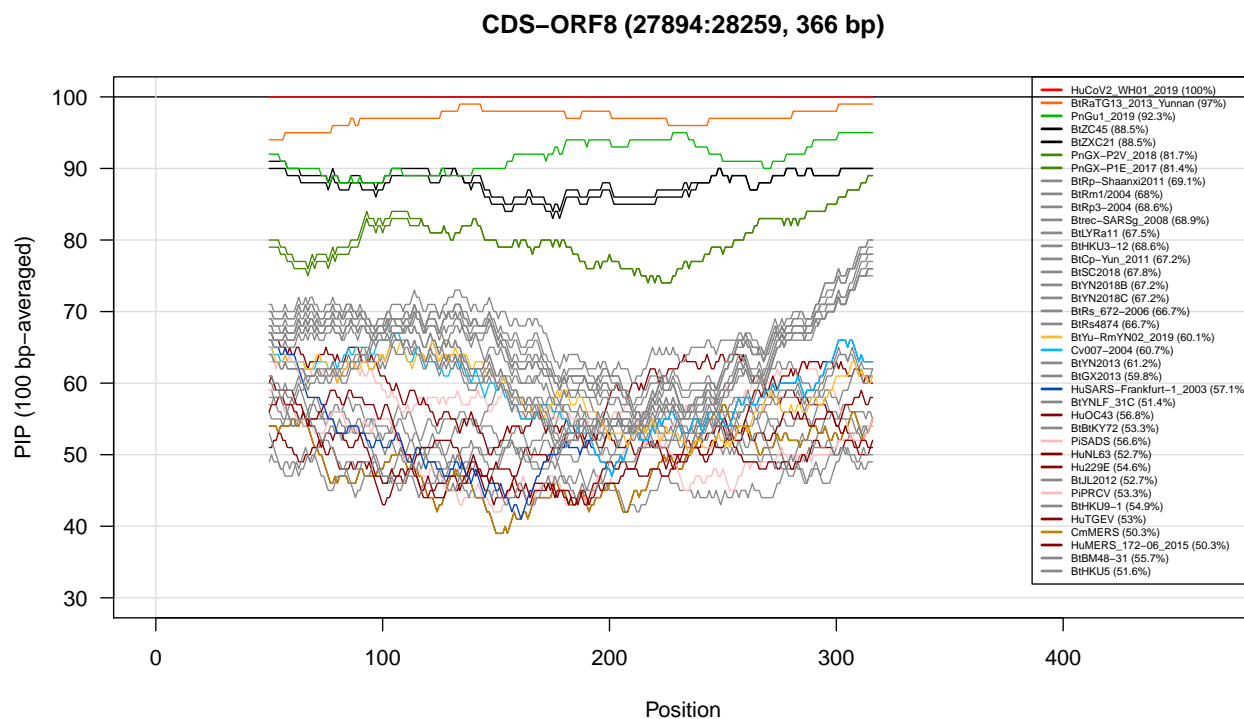


Figure 14: Feature-specific Percent Identical Positions (PIP) profiles.

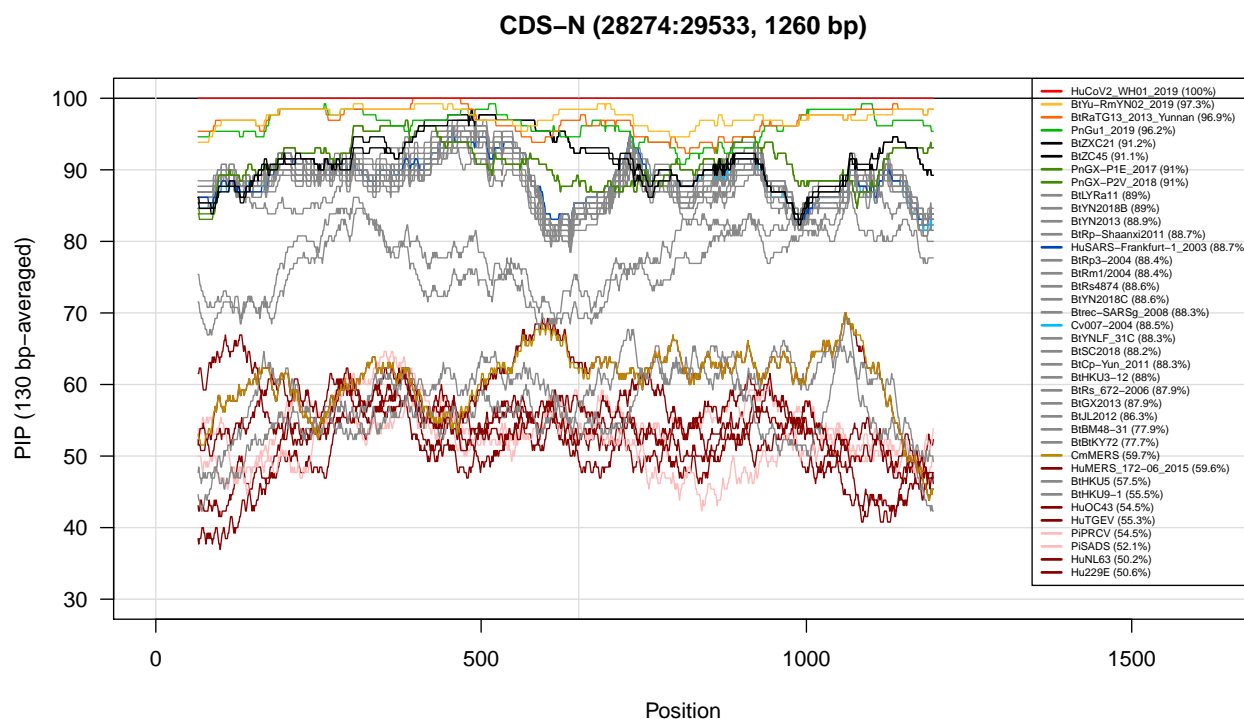


Figure 15: Feature-specific Percent Identical Positions (PIP) profiles.

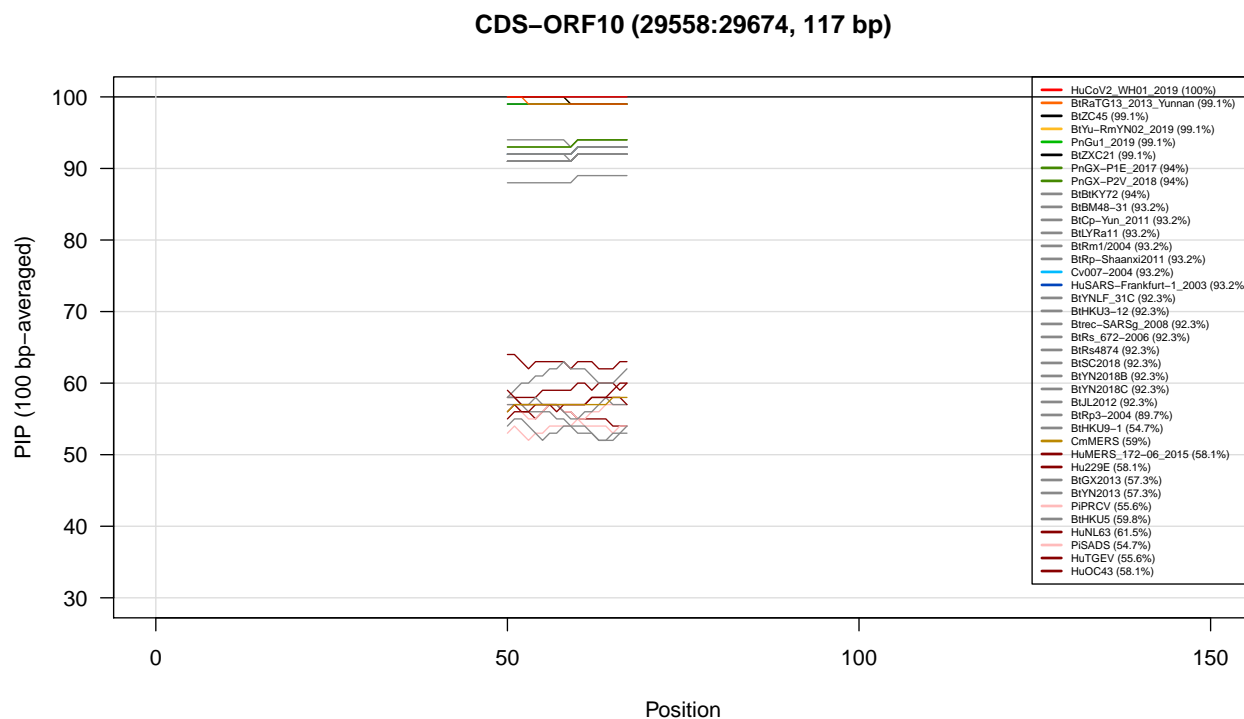


Figure 16: Feature-specific Percent Identical Positions (PIP) profiles.

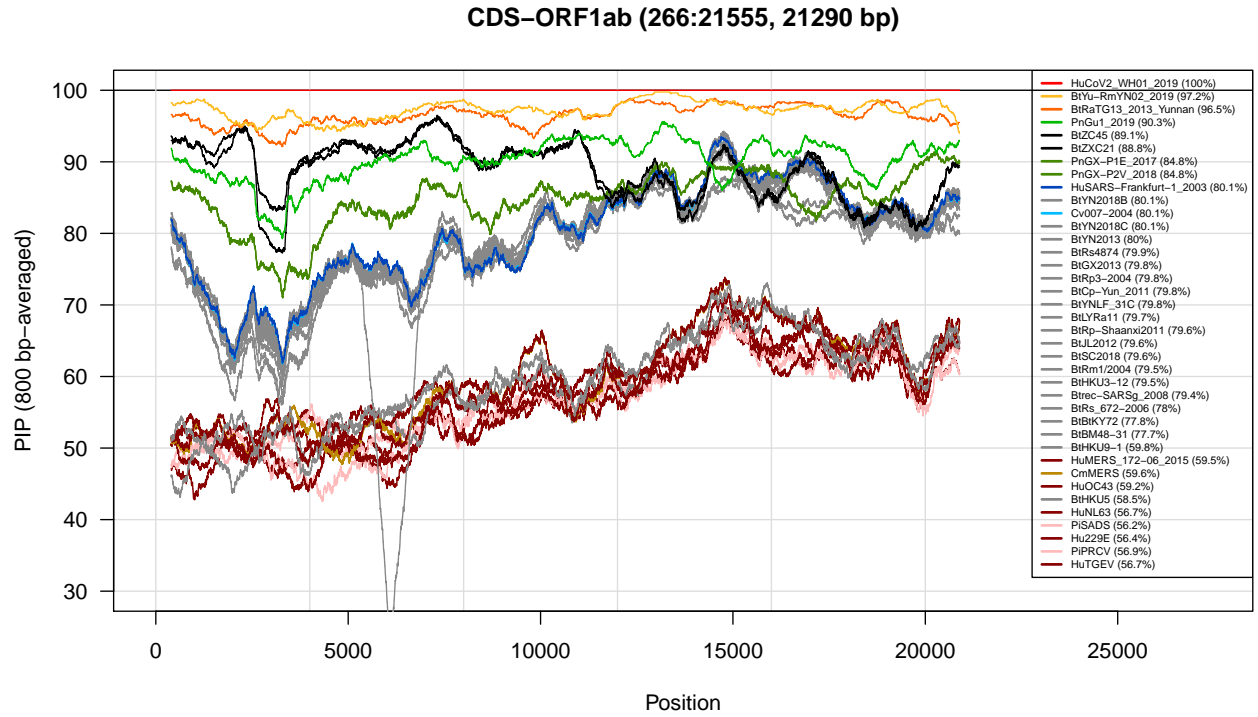


Figure 17: Feature-specific Percent Identical Positions (PIP) profiles.

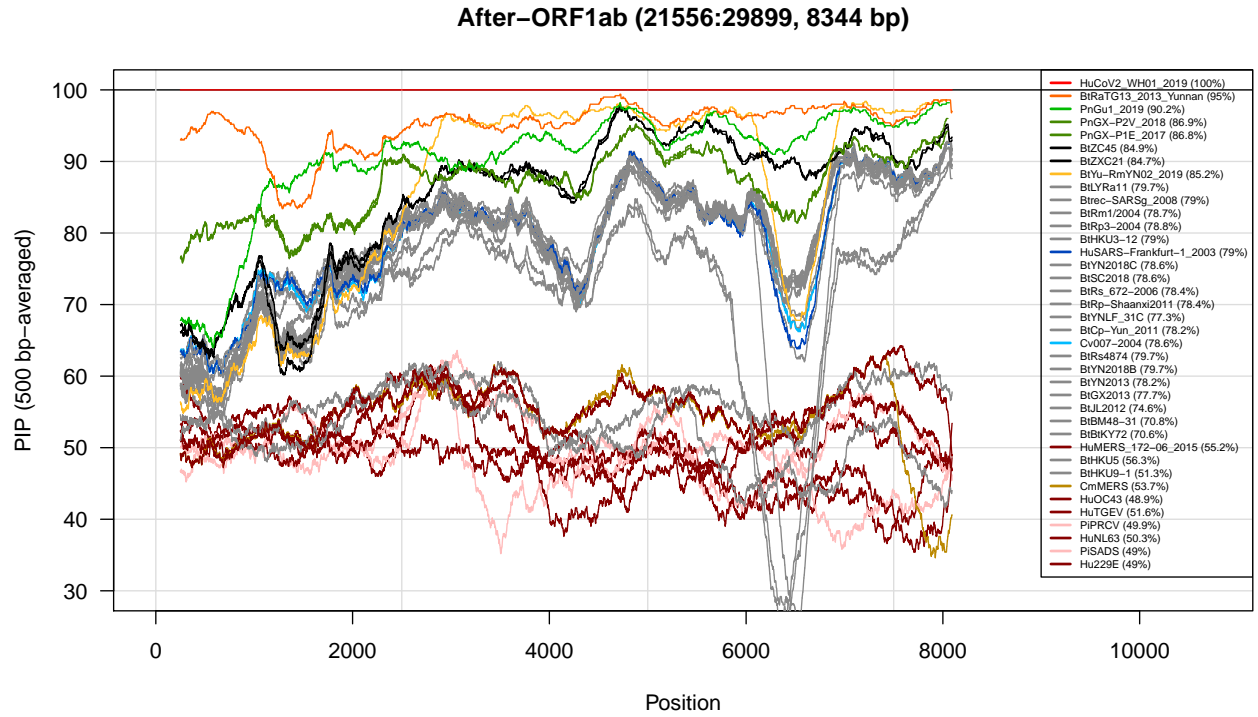


Figure 18: Feature-specific Percent Identical Positions (PIP) profiles.

Memory image

We store the result in a memory image, in order to be able to reload it to plot PIP profiles with different parameters.

```
save.image(file = outfiles["Memory image"])
```

Session info

```
sessionInfo()
```

R version 3.6.1 (2019-07-05)

Platform: x86_64-apple-darwin15.6.0 (64-bit)

Running under: macOS Mojave 10.14.6

Matrix products: default

BLAS: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib

LAPACK: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib

locale:

[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8

attached base packages:

[1] stats4 parallel stats graphics grDevices utils datasets methods base

other attached packages:

[1] Biostrings_2.52.0 XVector_0.24.0 IRanges_2.18.3 S4Vectors_0.22.1 BiocGenerics_0.30.0

loaded via a namespace (and not attached):

[1] Rcpp_1.0.4 digest_0.6.25 magrittr_1.5 evaluate_0.14 highr_0.8

[15] BiocManager_1.30.10 htmltools_0.4.0