

Unified classification and risk-stratification in Acute Myeloid Leukemia

Tutorial for Bayesian Dirichlet Process for finding molecular clusters :

This notebook is independent of the paper and can be used for clustering any dataset

```
In [8]: library(hdp)
library(ggplot2)
library('ggrepel')
library(dplyr)

library('IRdisplay')
source('src/tools.R')
source('src/hdp_tools.R')    ## containing useful functions for HDP
library(gridExtra)
```

```
In [9]: df_hdp1 <- read.table("data/aml_bdp1.tsv")

df_molecular <- read.table("data/aml_molecular_bdp.tsv")

genes_columns <- colnames(df_molecular)[grepl("^[:upper:]", colnames(df_molecular))]

cytos_columns <- colnames(df_molecular)[!grepl("^[:upper:]", colnames(df_molecular))]
```

BDP : A clustering tool to identify dominant molecular drivers

I) BDP1 : BDP algorithm applied on 2150 Patients and 153 genes and cytos

```
In [10]: # HDP Parameters

data <- df_hdp1
posterior_samples <- 250
initial_clusters <- 17
burn <- 7000
chains <- 3
base_dist <- prepare_distributions(df_hdp1)$gaussian
aa <- 0.5
ab <- -1.5
```

```
In [12]: dd_predicted_10_components <- get_prediction_result_dataframe(multi_output_cluster_10, df_hdp1)
tab <- get_table(dd_predicted_10_components[, 'predicted_component'])
tab

#plot components

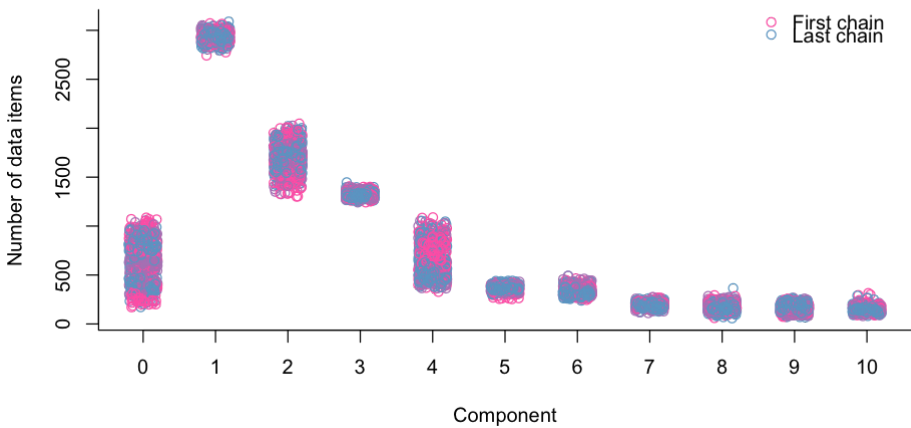
plot_components_size(multi_output_cluster_10, 8, 4.5)
set_notebook_plot_size(22, 2.6)
plot_comp_distn(multi_output_cluster_10,
                cat_names = colnames(df_hdp1),
                col       = "skyblue3",
                col_nonsig = "black",
                plot_title=c("Component 0 (going to HDP 2)", "Component 1 (NPM1)", "Component 2 (going
to HDP 2)", "Component 3 (TP53-Complex)", "Component 4 (going to HDP 2)",
                           "Component 5 (t(8;21))", "Component 6 (inv(16))", "Component 7 (CEBPA_bi)"
                           , "Component 8 (t(15;17) & t(11))", "Component 9 (additions)", "Component 10 (inv(3))")
                )
```

Number of components: 10

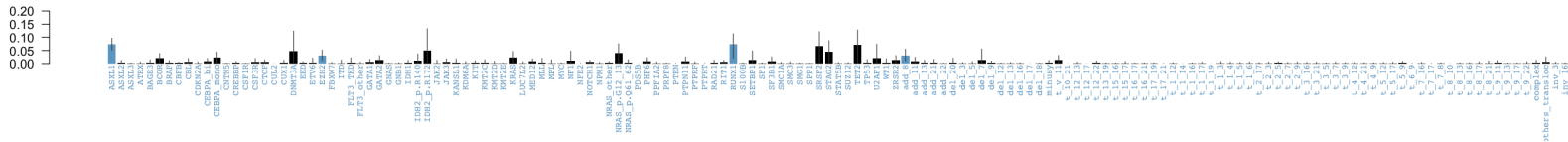
Number of NA rows : 51

A data.frame: 13 × 3

	values	count	freq
	<fct>	<int>	<chr>
2	1	833	38.7441860465116
3	2	418	19.4418604651163
4	3	229	10.6511627906977
5	4	128	5.95348837209302
6	5	114	5.30232558139535
7	6	114	5.30232558139535
1	0	96	4.46511627906977
9	8	61	2.83720930232558
12	NaN	51	2.37209302325581
8	7	50	2.32558139534884
11	10	29	1.34883720930233
10	9	27	1.25581395348837
13	-- total --	2150	100%



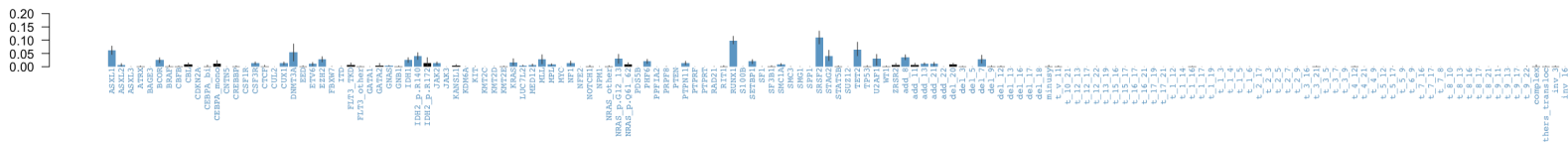
Component 0 (going to HDP 2)



Component 1 (NPM1)

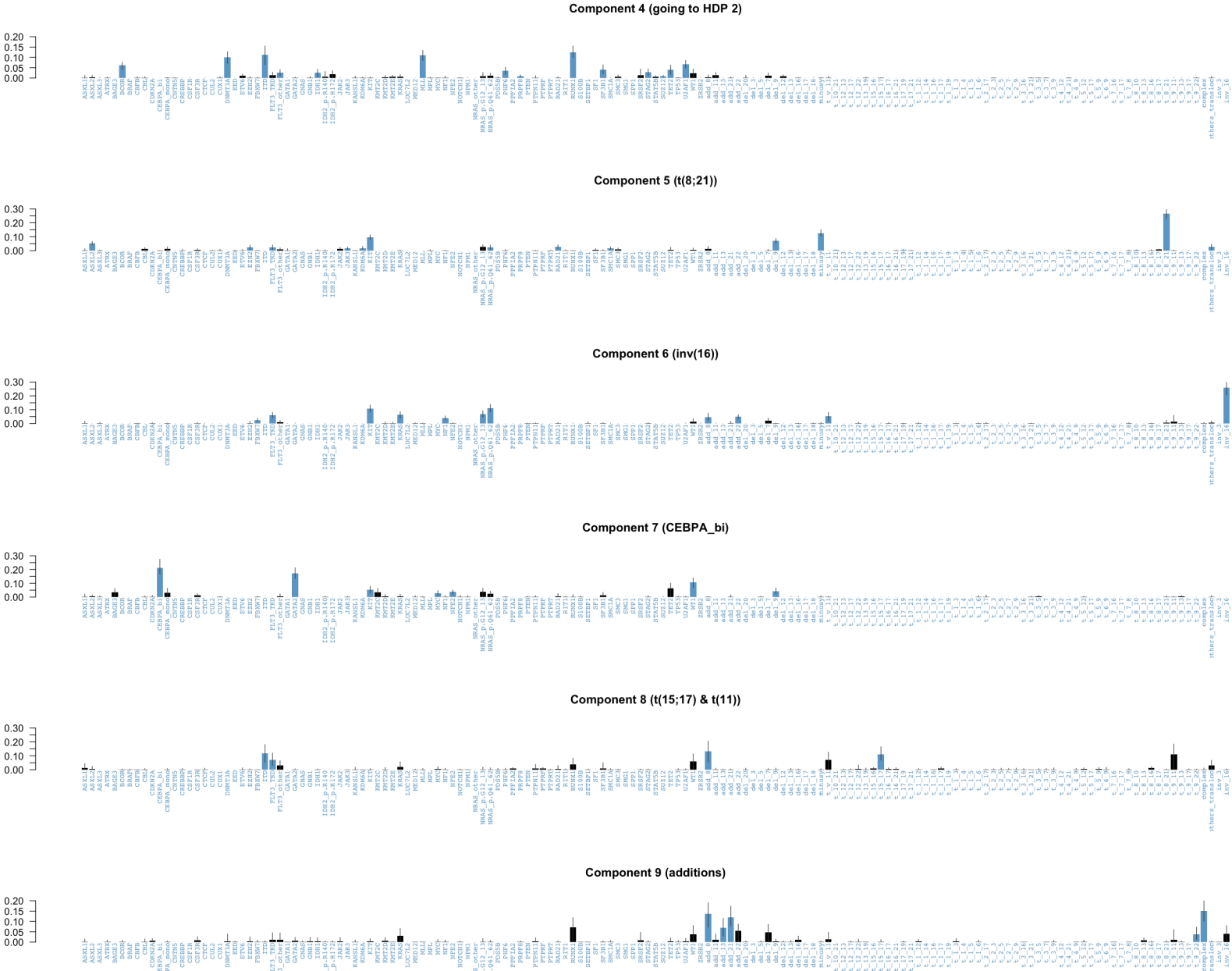


Component 2 (going to HDP 2)



Component 3 (TP53-Complex)







```

In [33]: # png("figures/figures_miscellaneous/bdpl_comp.png",width=8000,height=8000,res=200)
dd_predicted_10_components <- get_prediction_result_dataframe(multi_output_cluster_10, df_hdp1)
tab <- get_table(dd_predicted_10_components[, 'predicted_component'])
tab

# par(mfrow=c(12,1))
# plot_components_size(multi_output_cluster_10,8,4.5)
# plot_comp_distn(multi_output_cluster_10,
#                 cat_names = colnames(df_hdp1),
#                 col        = "skyblue3",
#                 col_nonsig = "black",
#                 plot_title=c("Component 0 (going to HDP 2)", "Component 1 (NPM1)", "Component 2 (going to HDP 2)", "Component 3 (TP53-Complex)", "Component 4 (going to HDP 2)",
#                               "Component 5 (t(8;21))", "Component 6 (inv(16))", "Component 7 (CEBPA_b i)", "Component 8 (t(15;17) & t(11))", "Component 9 (additions)", "Component 10 (inv(3))")
#                 )
# dev.off()
# png("figures/figures_miscellaneous/bdpl_proba.png",width=3000,height=800,res=140)
plot_assignment_probability_by_component(dd_predicted_10_components)#++scale_fill_manual(values="blue")
# ylab("Max Proba of assignment")+xlab("Predicted Component") +
#   theme(plot.title=element_text(size=30,face="bold",hjust=0.5),axis.text=element_text(size=20),axis.title=element_text(size=25,face="bold")) +
#   ggtitle("Assignment probability of Top Component")
# dev.off()

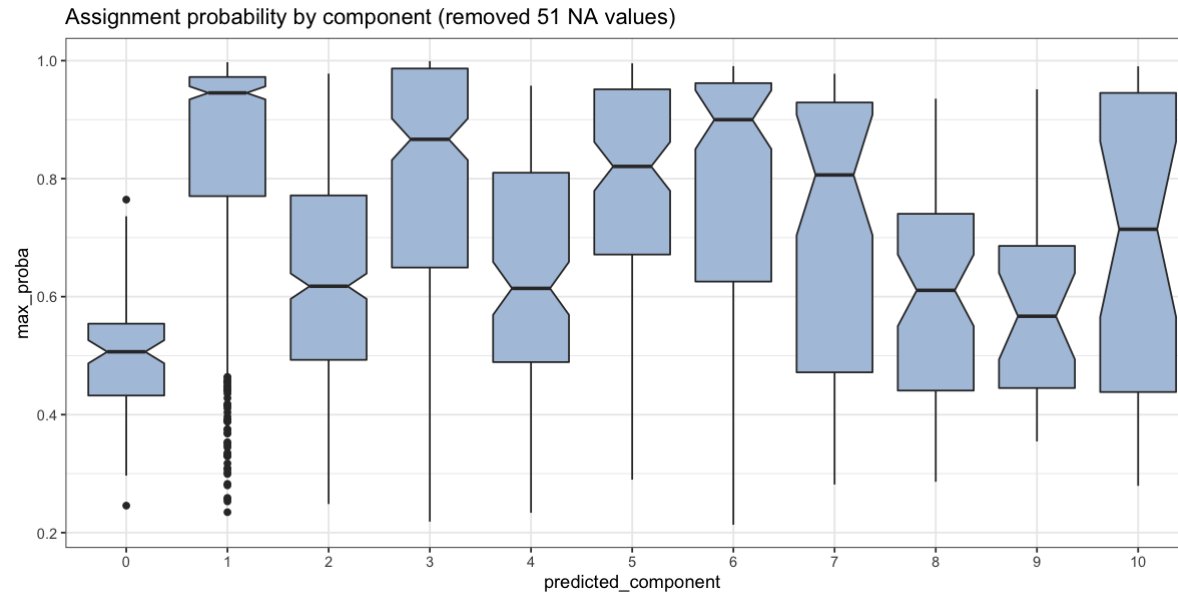
```


Number of components: 10

Number of NA rows : 51

A data.frame: 13 × 3

	values	count	freq
	<fct>	<int>	<chr>
2	1	833	38.7441860465116
3	2	418	19.4418604651163
4	3	229	10.6511627906977
5	4	128	5.95348837209302
6	5	114	5.30232558139535
7	6	114	5.30232558139535
1	0	96	4.46511627906977
9	8	61	2.83720930232558
12	NaN	51	2.37209302325581
8	7	50	2.32558139534884
11	10	29	1.34883720930233
10	9	27	1.25581395348837
13	-- total --	2150	100%



```
In [14]: # NA Rows correspond to no defining events i.e all columns are 0 for those 51 patients
df_molecular <- read.table("data/aml_molecular_bdp.tsv")
initial_components <- cbind(df_molecular, add_first_second_predicted_component(multi_output_cluster_10
))
```

Number of components: 10
Number of NA rows : 51

```
In [15]: write.table(initial_components, file="data/initial_components_with_proba.tsv", sep='\t')
```

Rules : Patients that do not have molecular drivers defined by HDP will be going to HDP 2

Step 1: Defining feature in each component

```
In [16]: tmp <- read.table("data/initial_components_with_proba.tsv")

tmp$initial_refined_component <- "not_assigned"

tmp[tmp$NPM1==1,]$initial_refined_component <- "NPM1"

tmp[tmp$initial_refined_component=="not_assigned" & (tmp$TP53==1 & tmp$complex==1),]$initial_refined_
component <- "TP53_complex"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==3 & (tmp$TP53==1
| tmp$complex==1),]$initial_refined_component <- "TP53_complex"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==5 & tmp$t_8_21==
1,]$initial_refined_component <- "t_8_21"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==6 & tmp$inv_16==
1,]$initial_refined_component <- "inv_16"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==7 & tmp$CEBPA_bi
==1,]$initial_refined_component <- "CEBPA_bi"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==8 & (tmp$t_v_11=
=1 | tmp$t_9_11==1),]$initial_refined_component <- "t_11"
tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==8 & tmp$t_15_17=
=1,]$initial_refined_component <- "t_15_17"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==9 & (tmp$add_8==
1 | tmp$add_11==1 | tmp$add_13==1 | tmp$add_21==1 | tmp$add_22==1),]$initial_refined_component <- "ad
ditions"

tmp[tmp$initial_refined_component=="not_assigned" & tmp$initial_predicted_component==10 & tmp$inv_3==
1,]$initial_refined_component <- "inv_3"

tmp[is.na(tmp$max_proba),]$initial_refined_component <- "no_events"
```

```
In [17]: t(data.frame(table(tmp$initial_refined_component)))
```

A matrix: 2 × 11 of type chr

Var1	additions	CEBPA_bi	inv_16	inv_3	no_events	not_assigned	NPM1	t_11	t_15_17	t_8_21	TP53_complex
Freq	21	35	85	13	51	915	682	31	19	99	199

Patients that did not pass the first test (not defining feature of their component)

```
In [18]: table_tmp <- t(data.frame(table(tmp[tmp$initial_refined_component=="not_assigned" & !is.element(tmp$initial_predicted_component,c(0,2,4)),]$initial_predicted_component)))
rownames(table_tmp) <- c("Components", "Number of Patients")
cat("Number of patients to test for second max proba in HDP 1: ")
cbind(table_tmp, c("Total", sum(table(tmp[tmp$initial_refined_component=="not_assigned" & !is.element(tmp$initial_predicted_component,c(0,2,4)),]$initial_predicted_component))))
```

Number of patients to test for second max proba in HDP 1:

A matrix: 2 × 9 of type chr

Components	1	3	5	6	7	8	9	10	Total
Number of Patients	160	32	14	29	14	11	5	16	281

Step 2: Second Max Proba with defining features

In [19]: *### Get Min Proba for each component*

```
min_proba_NPM1 = min(tmp[tmp$initial_predicted_component==1,]$max_proba, na.rm=T)
min_proba_TP53_complex = min(tmp[tmp$initial_predicted_component==3,]$max_proba, na.rm=T)
min_proba_t_8_21 = min(tmp[tmp$initial_predicted_component==5,]$max_proba, na.rm=T)
min_proba_inv_16 = min(tmp[tmp$initial_predicted_component==6,]$max_proba, na.rm=T)
min_proba_CEBPA_bi = min(tmp[tmp$initial_predicted_component==7,]$max_proba, na.rm=T)
min_proba_transloc = min(tmp[tmp$initial_predicted_component==8,]$max_proba, na.rm=T)
min_proba_additions = min(tmp[tmp$initial_predicted_component==9,]$max_proba, na.rm=T)
min_proba_inv_3 = min(tmp[tmp$initial_predicted_component==10,]$max_proba, na.rm=T)
```

In [20]: *## Refine patients with second max proba*

```
threshold <- 0.75

tmp[tmp$initial_refined_component=="not_assigned" ,]$initial_refined_component <- apply(tmp[tmp$initial_refined_component=="not_assigned" ,], 1, function(x) {
  if (all(is.na(x['second_predicted_component'])))
    return(NaN)
  else
    return(ifelse(x['second_predicted_component']==3 & x['second_max_proba']>=threshold*min_p
roba_TP53_complex & (x['complex']==1 | x["TP53"]==1), "TP53_complex",

    ifelse(x['second_predicted_component']==5 & x['second_max_proba']>=threshold*min_p
roba_t_8_21 & x['t_8_21']==1 , "t_8_21",

    ifelse(x['second_predicted_component']==6 & x['second_max_proba']>=threshold*min_p
roba_inv_16 & x['inv_16']==1, "inv_16",

    ifelse(x['second_predicted_component']==7 & x['second_max_proba']>=threshold*min_
proba_CEBPA_bi & x['CEBPA_bi']==1, "CEBPA_bi",

    ifelse(x['second_predicted_component']==8 & x['second_max_proba']>=threshold*min_p
roba_transloc & (x['t_v_11']==1 | x['t_9_11']==1), "t_11",
    ifelse(x['second_predicted_component']==8 & x['second_max_proba']>=threshold*min_p
roba_transloc & x['t_15_17']==1 , "t_15_17",

    ifelse(x['second_predicted_component']==9 & x['second_max_proba']>=threshold*min_p
roba_additions & (x['add_8']==1 | x['add_11']==1 | x['add_13']==1 | x['add_21']==1 | x['add_22']==1
), "additions",

    ifelse(x['second_predicted_component']==10 & x['second_max_proba']>=threshold*min_
proba_inv_3 & x['inv_3']==1, "inv_3",
    "not_assigned"))))))))

}))
t(data.frame(table(tmp$initial_refined_component)))
```

A matrix: 2 × 11 of type chr

Var1	additions	CEBPA_bi	inv_16	inv_3	no_events	not_assigned	NPM1	t_11	t_15_17	t_8_21	TP53_complex
Freq	24	39	88	13	51	888	682	41	20	99	205

```
In [21]: write.table(tmp, "data/bdp1_assignment.tsv", sep='\t')
```

Patients in category not assigned do not have defining features of their component nor of their second predicted component. They will also go in BDP 2.

II- BDP2 : BDP reapplied on 888 Patients

888 Patients from Component 0,2 and 4 + Patients from other components that do not have defining features from their component.

```
In [22]: hdp1_assignment <- read.table('data/bdp1_assignment.tsv')

# We reapply HDP to those patients :

df_to_recluster <- hdp1_assignment[hdp1_assignment$initial_refined_component=="not_assigned",c(genes_
columns,cytos_columns)] ## Reapply HDP only on not assigned patients and keep only gene and cyto co
lums
```

```

In [23]: num_cols = ncol(df_to_recluster[colSums(df_to_recluster) > 0]) ## keep only column that have at least a mutation (otherwise HDP is not applicable)
num_cols
bin <- function(x){
  set.seed(123)
  (rbinom(1, num_cols, mean(x))+1)/num_cols
}

###Normal

normal <- function(x){
  set.seed(123)
  abs(rnorm(1,mean(x),sd(x)))
}

###Poisson

poisson <- function(x){
  set.seed(123)
  (rpois(num_cols,1)+1)/num_cols
}

###Uniform equally over all columns

equally <- function(x){
  set.seed(123)
  1/num_cols
}

###Repet 1

repet <- function(x){
  set.seed(123)
  1
}

binomial <- unlist(sapply(df_to_recluster[colSums(df_to_recluster) > 0],bin))
gaussian <- unlist(sapply(df_to_recluster[colSums(df_to_recluster) > 0],normal))
pois <- as.numeric(unlist(sapply(df_to_recluster[colSums(df_to_recluster) > 0],poisson)))
unif <- unlist(sapply(df_to_recluster[colSums(df_to_recluster) > 0],equally))

```


129

```
In [24]: # HDP2 Parameters

data <- df_to_recluster[colSums(df_to_recluster) > 0]
posterior_samples <- 150
initial_clusters <- 5
burn <- 5000
chains <- 3
base_dist <- prepare_distributions(df_to_recluster[colSums(df_to_recluster) > 0])$gaussian
aa <- 2
ab <- -6
```

```
In [25]: multi_output <- launch(df_to_recluster[colSums(df_to_recluster) > 0],base_dist,initial_clusters,burn,  
posterior_samples,chains,aa,ab)  
  
multi_output_recluster_5 <- extract_components(multi_output)
```

```

Initialise HDP on a 888 x 129 dataframe
  → create HDP structure... done!
  → add DP node for each patient... done!
  → assign the data to the nodes... done!
### Experiment 1 (seed = 100) ###
Activate HDP nodes and run posterior sampling
  → activate HDP nodes... done!
  → run posterior sampling...
[1] "5000 burn-in iterations in 0.2 mins"

### Experiment 2 (seed = 200) ###
Activate HDP nodes and run posterior sampling
  → activate HDP nodes... done!
  → run posterior sampling...
[1] "5000 burn-in iterations in 0.2 mins"

### Experiment 3 (seed = 300) ###
Activate HDP nodes and run posterior sampling
  → activate HDP nodes... done!
  → run posterior sampling...
[1] "5000 burn-in iterations in 0.2 mins"

Object of class hdpSampleMulti
Number of chains: 3
Total posterior samples: 450
Components: NO. Run hdp_extract_components
-----
Final hdpState from first chain:
Object of class hdpState
Number of DP nodes: 889
Index of parent DP: 0 1 1 1 1 1 1 1 1 1 ...
Number of data items per DP: 0 8 6 5 4 3 4 4 7 4 ...
Index of conparam per DP: 1 1 1 1 1 1 1 1 1 1 ...
Conparam hyperparameters and current value:
      Shape Rate      Value
Conparam 1      2      6 2.65891
Number of data categories: 129
Number of clusters: 34
Initialised with 5 clusters, using random seed 100
Extract HDP components from posterior sampling
  → extract components... done!
* 5 components found

```



```
In [26]: dd_predicted_5_components <- get_prediction_result_dataframe(multi_output_recluster_5, df_final)

tab <- get_table(dd_predicted_5_components[, 'predicted_component'])

tab

plot_components_size(multi_output_recluster_5, 8, 4.5)

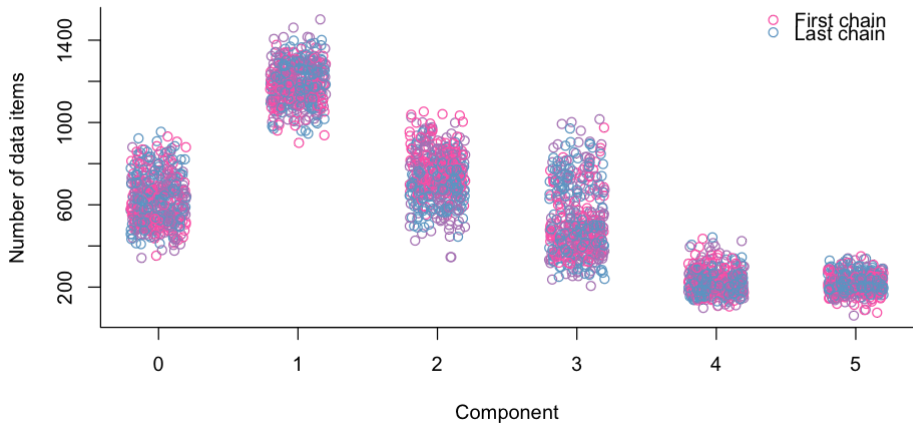
set_notebook_plot_size(16, 2.6)

plot_comp_distn(multi_output_recluster_5,
                cat_names = colnames(df_to_recluster[colSums(df_to_recluster) > 0]),
                col       = "skyblue3",
                col_nonsig = "black",
                plot_title=c("Component 0 (mix)", "Component 1 (chr splicing)", "Component 2 (chr s
plicing)", "Component 3 (DNMT3A IDH1-2)", "Component 4 (chr splicing)", "Component 5 (WT1 and t(6;9))"))
```

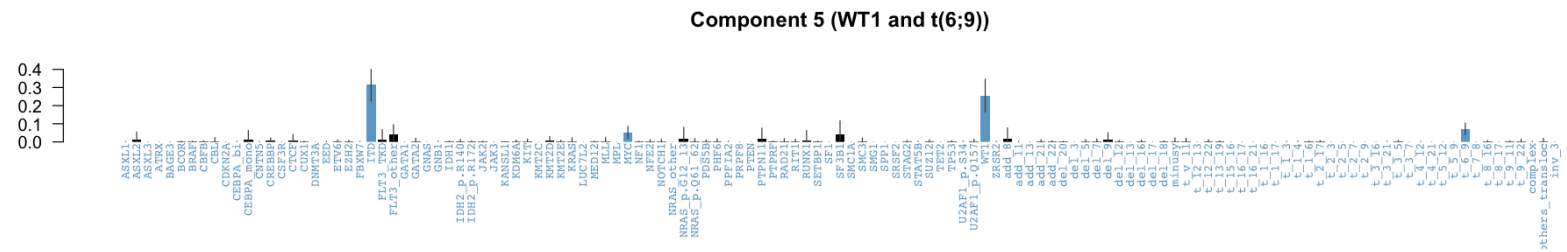
Number of components: 5
Number of NA rows : 0

A data.frame: 7 × 3

	values	count	freq
	<fct>	<int>	<chr>
2	1	319	35.9234234234234
3	2	228	25.6756756756757
1	0	144	16.2162162162162
4	3	101	11.3738738738739
6	5	62	6.98198198198198
5	4	34	3.82882882882883
11	-- total --	888	100%








```

In [27]: # png("figures/figures_miscellaneous/bdp2_comp.png",width=8000,height=8000,res=200)
dd_predicted_5_components <- get_prediction_result_dataframe(multi_output_recluster_5, df_final)

tab <- get_table(dd_predicted_5_components[, 'predicted_component' ])

tab

# par(mfrow=c(12,1))
# plot_components_size(multi_output_recluster_5,8,4.5)
# plot_comp_distn(multi_output_recluster_5,
#                 cat_names = colnames(df_to_recluster[colSums(df_to_recluster) > 0]),
#                 col       = "skyblue3",
#                 col_nonsig = "black",
#                 plot_title=c("Component 0 (mix)", "Component 1 (chr splicing)", "Component 2 (chr
splicing)", "Component 3 (DNMT3A IDH1-2)", "Component 4 (chr splicing)", "Component 5 (WT1 and t(6;
9))"))

# dev.off()
# png("figures/figures_miscellaneous/bdp2_proba.png",width=3000,height=800,res=140)
# plot_assignement_probability_by_component(dd_predicted_5_components)+scale_fill_manual(values="blue")
# ylab("Max Proba of assignment")+xlab("Predicted Component") +
#   theme(plot.title=element_text(size=30,face="bold",hjust=0.5),axis.text=element_text(size=20),axis.title=element_text(size=25,face="bold")) +
#   ggtitle("Assignment probability of Top Component")
# dev.off()

```

Number of components: 5
Number of NA rows : 0

A data.frame: 7 × 3

	values	count	freq
	<fct>	<int>	<chr>
2	1	319	35.9234234234234
3	2	228	25.6756756756757
1	0	144	16.2162162162162
4	3	101	11.3738738738739
6	5	62	6.98198198198198
5	4	34	3.82882882882883
11	-- total --	888	100%

```
In [28]: plot_assignment_probability_by_component(dd_predicted_5_components)
```

notch went outside hinges. Try setting notch=FALSE.

