# ACMG EH

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## Load required packages

```
library(tidyverse)
```

## Import data

```
data anno <- read delim("data/toy example/annovar annotations.txt",
                         col names = TRUE,
                         na = ".",
                         delim="\t")
data_info <- read_delim("data/toy_example/sample_info.txt",</pre>
                         delim="\t",
                         col_select = c(CHROM:NCALLED)) %>%
  dplyr::rename(case_ac=AC, case_af=AF, case_an=AN)
data_vep_csq <- read_delim("data/toy_example/vep_annotations.txt",</pre>
                            col_names = c("CHROM", "POS", "REF", "ALT", "Allele", "Consequence",
                                           "IMPACT", "SYMBOL", "Gene", "Feature_type", "Feature", "BIOTYPE", "EXON", "INTRON", "HGVSc", "HGVSp", "cDNA_position"
                                           "CDS_position", "Protein_position", "Amino_acids", "Codons",
                                           "Existing_variation", "DISTANCE", "STRAND", "FLAGS", "SYMBOL_S
                                           "HGNC_ID", "CANONICAL", "MANE_SELECT", "MANE_PLUS_CLINICAL", "
                                           "GENE_PHENO", "DOMAINS", "HGVS_OFFSET", "LoF", "LoF_filter", "
                                           "LoF_info", "SpliceAI_pred_DP_AG", "SpliceAI_pred_DP_AL", "Spl
                                           "SpliceAI_pred_DP_DL", "SpliceAI_pred_DS_AG", "SpliceAI_pred_D
                                           "SpliceAI_pred_DS_DL", "SpliceAI_pred_SYMBOL", "existing_InFra
                                           "existing_OutOfFrame_oORFs", "existing_uORFs",
                                           "five_prime_UTR_variant_annotation", "five_prime_UTR_variant_c
data_GT = read_delim("data/toy_example/sample_GT.txt")
clinvar <- read_delim("data/processed/clinvar_PLP_BLB_CLNSIG_CLNSTAT.tsv.gz")</pre>
uniprot_variations <- read_delim("data/processed/humsvar_corrected_withID.txt.gz") %>%
 filter(Variant category uniprot=="LP/P") %>%
  mutate(POS_uniprot=str_extract(AAchange_uniprot, "[0-9]+"),
         AA1_uniprot=str_sub(AAchange_uniprot, 3, 5),
         AA2_uniprot=str_sub(AAchange_uniprot, -3, -1))
```

```
pm1_functional_domains_withP <- read_delim("data/processed/uniprot_functional_regions_no_benign_1_patho
pm1_functional_domains <- read_delim("data/processed/uniprot_functional_regions_no_benign.tsv.gz")
PER <- readxl::read_xlsx("data/processed/PER_coordinates.xlsx", sheet = 4)
repeat_masker <- read_delim("data/processed/hg38_rmsk_cleaned.txt.gz", col_names = c("Chr", "Start", "E
  mutate(Start=Start+1) #data from UCSC is O-based. Add 1 to Start to become 1-based.
bp1_genes <- read_delim("data/processed/BP1_genes.txt.gz", delim="\n") %%
  mutate(bp1_gene=1)
pp2_genes <- read_delim("data/processed/PP2_genes.txt.gz", delim="\n") %>%
  mutate(pp2_gene=1)
data <- data_anno %>%
  bind_cols(data_vep_csq) %>%
  bind_cols(data_info %>% dplyr::select(-CHROM, -POS, -REF, -ALT)) %>%
  bind_cols(data_GT %>% dplyr::select(-CHROM, -POS, -REF, -ALT))
rm(data_vep_csq, data_info, data_GT, data_anno)
data <- data %>%
  filter(IMPACT %in% c("HIGH", "MODERATE")) %>%
  filter(Chr != "chrM") %>%
  filter(VAR >= 1)
data <- data %>%
  replace_na(list(gnomad41_genome_AF=0,
                  gnomad41_exome_AF=0,
                  gnomad41_genome_AF_nfe=0,
                  gnomad41_genome_AF_afr=0,
                  gnomad41_genome_AF_amr=0,
                  gnomad41_genome_AF_sas=0,
                  gnomad41_genome_AF_eas=0))
data$popmax <- pmax(data$gnomad41_genome_AF,</pre>
                    data$gnomad41_exome_AF)
data$spliceai_max <- pmax(as.numeric(data$SpliceAI_pred_DS_AG),</pre>
                          as.numeric(data$SpliceAI_pred_DS_AL),
                          as.numeric(data$SpliceAI_pred_DS_DG),
                          as.numeric(data$SpliceAI_pred_DS_DL))
```

## ACMG/AMP rules

### PVS1 and PVS1\_strong

High confidence loss-of-function variants get PVS1 = 1 and low confidence loss-of-function variants get  $PVS1\_strong = 1$ . For both of them, the transcript should be CANONICAL or MANE:

```
data$pvs1 <- 0
data$pvs1_strong <- 0

data[((!is.na(data$MANE_SELECT)) | (!is.na(data$CANONICAL))) &</pre>
```

#### • PS3

If the variant is reported in ClinVar (with status of practice\_guideline/reviewed\_by\_expert\_pane) or Uniprot, is pathogenic/likely pathogenic:

```
#clinvar
data <- data %>%
  left join(clinvar %>% dplyr::select(CHROM:ALT, CLINSIG, CLINSTAT))
data$ps3 <- 0
data <- data %>%
  mutate(ps3 = case_when(str_detect(CLINSIG, "athogenic") &
                        (CLINSTAT %in% c("practice_guideline", "reviewed_by_expert_panel")) ~ 1,
#uniprot
data <- data %>%
  separate(col=HGVSp, into=c("protein_id", "AAchange_data"), sep=":", remove = F) %>%
  mutate(POS_data=str_extract(AAchange_data, "[0-9]+"),
         AA1_data=str_sub(AAchange_data, 3, 5),
         AA2_data=str_sub(AAchange_data, -3, -1)) %>%
  left_join(uniprot_variations %>% dplyr::select(Gene, POS_uniprot, AA1_uniprot, AA2_uniprot, Variant_c
            by=c("Gene"="Gene", "POS_data"="POS_uniprot", "AA1_data"="AA1_uniprot", "AA2_data"="AA2_uni
data[!is.na(data$Variant category uniprot) & data$Variant category uniprot=="LP/P" &
      ((!str detect(data$CLINSIG, "enign")) | (is.na(data$CLINSIG))), "ps3"] <- 1
```

## • PS1 and PM5

If variant is missense, and reported as pathogenic in Clinvar/Uniprot, different nucleotide which results in the same amino acid should get PS1. Different nucleotide with different amino acide gets PM5:

```
#clinvar
clinvar_missense <- clinvar %>%
  filter(str_detect(CLINSIG, "athogenic")) %>%
  filter((str_detect(Consequence, "missense") |
          str_detect(Consequence, "start_lost") |
          str_detect(Consequence, "stop_lost") &
          (str_length(REF)==1 & str_length(ALT)==1))) %>%
  dplyr::rename(ALT_clinvar=ALT, Amino_acids_clinvar=Amino_acids) %>%
  dplyr::select(CHROM:ALT_clinvar, Amino_acids_clinvar)
data <- data %>%
  left_join(clinvar_missense)
#uniprot
data <- data %>%
  left_join(uniprot_variations %>%
              dplyr::select(Gene, POS_uniprot, AA1_uniprot, AA2_uniprot, Variant_category_uniprot) %>%
              dplyr::rename(Variant_category_uniprot_pm5=Variant_category_uniprot),
```

```
by=c("Gene"="Gene", "POS_data"="POS_uniprot", "AA1_data"="AA1_uniprot"))
#ps1
data$ps1 <- 0
data[(!is.na(data$ALT_clinvar)) &
      (data$ALT != data$ALT_clinvar) &
      (data$Amino_acids == data$Amino_acids_clinvar) &
      (data\$ps3==0) \&
      (data$spliceai_max<0.1 | is.na(data$spliceai_max)) &</pre>
      (str_detect(data$Consequence, "missense") | str_detect(data$Consequence, "start_lost") | str_dete
      (str_length(data$REF)==1 & str_length(data$ALT)==1), "ps1"] <- 1
#pm5
data$pm5 <- 0
data[(!is.na(data$ALT_clinvar)) &
      (data$ALT != data$ALT_clinvar) &
      (data$Amino_acids != data$Amino_acids_clinvar) &
      (data$ps3==0) &
      (data$spliceai_max<0.1 | is.na(data$spliceai_max)) &</pre>
      (data$ps1==0) &
      (str_detect(data$Consequence, "missense") | str_detect(data$Consequence, "start_lost") | str_dete
      (str_length(data$REF)==1 & str_length(data$ALT)==1), "pm5"] <- 1
data[(!is.na(data$AA2_uniprot)) &
      (data$AA2 data!=data$AA2 uniprot) &
      (data$ps3==0) &
       (data$spliceai_max<0.1 | is.na(data$spliceai_max)) &</pre>
      (data$ps1==0) &
      (str_detect(data$Consequence, "missense") | str_detect(data$Consequence, "start_lost") | str_dete
      (str_length(data$REF)==1 & str_length(data$ALT)==1), "pm5"] <- 1
```

## • PM2 (suporting)

Variant is absent from the controls and is rare in the population:

```
data$pm2_supporting <- 0
data[data$popmax<=0.01, "pm2_supporting"] <- 1</pre>
```

#### • PM1

Missense variant located in a functional domain without benign variant, or in a hotspot.

- 1. Functional domains (active site, binding site, calcium-binding region, DNA-binding region, metal ion-binding site, nucleotide phosphate-binding region, site, and zinc finger region) are retrieved from Uniprot and intersected with ClinVar. For a given domain, if there is no missense benign variant inside the domain, it is considered as functional domain.
- 2. Hotspots were retreive from "Identification of pathogenic variant enriched regions across genes and gene families" paper

```
data$pm1 <- 0

#domains
domains_input <- rbind(pm1_functional_domains, pm1_functional_domains_withP) %>%
    distinct()

data_ranges <- GenomicRanges::GRanges(seqnames = data$Chr,</pre>
```

```
ranges = IRanges::IRanges(start = data$Start, end = data$End))
domains_ranges <- IRanges::reduce(GenomicRanges::GRanges(seqnames = domains_input$Chr,</pre>
                                                           ranges = IRanges::IRanges(start = domains_inpu
                                                                                      end = domains_input$;
data_domain_overlap <- IRanges::subsetByOverlaps(data_ranges, domains_ranges)</pre>
if(length(data_domain_overlap)>0){
  data_domain_df <- data.frame(Chr=GenomicRanges::seqnames(data_domain_overlap),</pre>
                               Start=GenomicRanges::start(data_domain_overlap),
                               End=GenomicRanges::end(data_domain_overlap),
                               pm1_domain=1)
  data <- data %>%
    left_join(data_domain_df) %>%
    replace_na(list(pm1_domain=0))
  data[(data$pm1_domain==1) &
       (str_detect(data$Consequence, "missense") |
          str_detect(data$Consequence, "start_lost") |
          str_detect(data$Consequence, "stop_lost") &
       (str_length(data$REF)==1 & str_length(data$ALT)==1)), "pm1"] <- 1
}
#hotspots
PER$Codon <- as.character(PER$Codon)</pre>
data <- data %>%
  left_join(PER %>% dplyr::select(Transcript, Codon, Stats), by=c("Feature"="Transcript", "POS_data"="C
data[!is.na(data$Stats) &
     (str_detect(data$Consequence, "missense") |
     str_detect(data$Consequence, "start_lost") |
     str_detect(data$Consequence, "stop_lost")) &
     (str_length(data$REF)==1 & str_length(data$ALT)==1), "pm1"] <- 1</pre>
```

### • PM4

If the variant is inframe indel and falls inside a non-repeat region or if the variant is start\_lost/stop\_lost. Also, variant should not be a loss-of-function. To detect repeat regions, we use repeat masker from UCSC:

```
rmsk=1)
  data <- data %>%
   left_join(data_rmsk_df) %>%
   replace_na(list(rmsk=0))
  for(i in 1:nrow(data)){
    if(!is.na(str_locate(data$Consequence[i], "inframe"))[1] &
       (data$rmsk[i]==0 & (is.na(data$DOMAINS[i]) | !str_detect(data$DOMAINS[i], "ow_complexity"))) &
       is.na(data$LoF[i])){
      data$pm4[i] <- 1
   }
   if(!is.na(str_locate(data$Consequence[i], "stop_lost")[1]) & is.na(data$LoF[i])){
      data$pm4[i] <- 1
   }
    if(!is.na(str_locate(data$Consequence[i], "start_lost")[1]) & is.na(data$LoF[i])){
      data$pm4[i] <- 1
   }
  }
}
```

#### • PP2

If missense variant is inside a gene depleted from pathogenic missense variants. To consider a gene as a gene depleted from pathogenic missense, two criteria must be satisfied for it:

```
1. \frac{pathogenic\_missense}{pathogenic\_total} >= 0.7
2. \frac{benign\_missense}{total\_missense} <= 0.3
```

3. at least 5 pathogenic missense are reported

### • PP3 and BP4

Computational prediction. PP3 is applied when REVEL >= 0.5 and BayesDel\_noAF\_pred predicts damaging

BP4 is applied if REVEL < 0.4 or BayesDel noAF pred predicts tolerated

## • BA1

If a variant is very common (>0.05) this rule is applied, and variant is considered as benign.

```
data$ba1 <- 0
data[data$gnomad41_genome_AF_nfe>0.05 |
    data$gnomad41_genome_AF_afr>0.05 |
    data$gnomad41_genome_AF_amr>0.05 |
    data$gnomad41_genome_AF_sas>0.05 |
    data$gnomad41_genome_AF_sas>0.05 |
    data$gnomad41_genome_AF_eas>0.05, "ba1"] <- 1</pre>
```

### • BS1

If a variant is common (>0.02):

```
data$bs1 <- 0
data[data$gnomad41_genome_AF_nfe>0.02 |
    data$gnomad41_genome_AF_afr>0.02 |
    data$gnomad41_genome_AF_amr>0.02 |
    data$gnomad41_genome_AF_sas>0.02 |
    data$gnomad41_genome_AF_sas>0.02 |
    data$gnomad41_genome_AF_sas>0.02 |
```

### • BS3

Reported as benign in ClinVar and its status is practice\_guideline/reviewed\_by\_expert\_panel:

#### • BP1

Missense variant falls inside a gene which damaging mutations are mostly truncating. To detect such genes, we do the following steps:

- 1. Count pathogenic truncating ('frameshift\_variant', 'splice\_acceptor\_variant', 'splice\_donor\_variant', 'stop\_gained') variants for each gene
- 2. Count total pathogenic variants for each gene (at least 5 pathogenic must be found in a gene)
- 3. Keep genes that  $\frac{truncating\_pathogenic}{total\_pathogenic} >= 0.7$

### • BP3

Variant is an inframe indel and falls inside a repeat region:

```
data$bp3 <- 0
if(length(data_rmsk_overlap)>0){
  for(i in 1:nrow(data)){
    if(!is.na(str_locate(data$Consequence[i], "inframe"))[1] &
        (data$rmsk[i]==1 | (str_detect(data$DOMAINS[i], "ow_complexity") & !is.na(data$DOMAINS[i])) )){
        data$bp3[i] <- 1
    }
}</pre>
```

# Pathogenicity probability

We use the recommended model in "modeling the ACMG/AMP variant classification guidelines as a Bayesian classification framework" to assign a pathogenicity probability to each variant:

```
data$probability <- 0
prior <- 0.1
for (i in 1:nrow(data)){
 PVS <- data$pvs1[i]
  PS <- data$ps1[i]+data$ps3[i]+data$pvs1_strong[i]
  PM <- data$pm1[i]+data$pm4[i]+data$pm5[i]</pre>
  PP <- data$pp2[i]+data$pp3[i]+data$pm2_supporting[i]</pre>
  BS <- data$bs1[i]+data$bs3[i]
  BP <- data$bp1[i]+data$bp3[i]+data$bp4[i]</pre>
  if (data$ba1[i]==1){
    data$probability[i] <- 0</pre>
  }
  else{
    odds_path \leftarrow 350^((PP/8)+(PM/4)+(PS/2)+(PVS/1)-(BP/8)-(BS/2))
    data$probability[i] <- (odds_path*prior)/(((odds_path-1)*prior) + 1)</pre>
  }
}
### to remove duplicated rows (for a given variant, choose the maximum probability)
temp <- data %>%
  group_by(CHROM, POS, REF, ALT) %>%
  summarize(probability=max(probability))
data <- data %>%
  dplyr::select(-probability) %>%
  left_join(temp) %>%
  distinct(CHROM, POS, REF, ALT, probability, .keep_all = T)
```

## **Extract Pathogenic Variants**

The recommended threshold is 0.9, but there are some variants whose probability of pathogenicity is 0.899. We consider them as pathogenic as well.

```
### extract pathogenic
data_patho <- data %>%
  filter(probability>=0.89)
```