Figures for Paper

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```
library(readr)
library(VennDiagram)
## Loading required package: grid
## Loading required package: futile.logger
library(magrittr)
library(rtracklayer)
## Loading required package: GenomicRanges
## Loading required package: stats4
## Loading required package: BiocGenerics
## Loading required package: parallel
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:parallel':
##
       clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
##
       clusterExport, clusterMap, parApply, parCapply, parLapply,
##
       parLapplyLB, parRapply, parSapply, parSapplyLB
## The following objects are masked from 'package:stats':
##
##
       IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
##
       anyDuplicated, append, as.data.frame, cbind, colMeans,
##
       colnames, colSums, do.call, duplicated, eval, evalq, Filter,
##
       Find, get, grep, grepl, intersect, is.unsorted, lapply,
##
       lengths, Map, mapply, match, mget, order, paste, pmax,
##
       pmax.int, pmin, pmin.int, Position, rank, rbind, Reduce,
##
       rowMeans, rownames, rowSums, sapply, setdiff, sort, table,
       tapply, union, unique, unsplit, which, which.max, which.min
## Loading required package: S4Vectors
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:base':
##
##
       expand.grid
## Loading required package: IRanges
## Loading required package: GenomeInfoDb
```

Converting the Dataframes to Genomic Ranges

```
ChIPSeqEnrichedRegionsGRanges <- makeGRangesFromDataFrame(ChIPSeqEnrichedRegions,
                         keep.extra.columns=TRUE,
                         ignore.strand=TRUE,
                         seqinfo=NULL,
                         seqnames.field="X1",
                         start.field="X2",
                         end.field="X3",
                         strand.field="X8",
                         starts.in.df.are.Obased=FALSE)
# Expaning the ChIP-enriched region by 200bp up and down stream to identify motifs in surrounding areas
ChIPSeqEnrichedRegionsGRanges <- ChIPSeqEnrichedRegionsGRanges+200
MotifOverlapRPredictedSitesGRanges <- makeGRangesFromDataFrame(MotifOverlapRPredictedSites,
                         keep.extra.columns=TRUE,
                         ignore.strand=TRUE,
                         seqinfo=NULL,
                         seqnames.field="X1",
                         start.field="X2",
                         end.field="X3",
                         strand.field="X5",
                         starts.in.df.are.Obased=FALSE)%>%reduce()
```

```
AllMotifInstancesGRanges <- makeGRangesFromDataFrame(AllMotifInstances,
                                              keep.extra.columns=TRUE,
                                              ignore.strand=TRUE,
                                              seginfo=NULL,
                                              seqnames.field="X1",
                                              start.field="X2",
                                              end.field="X3",
                                              strand.field="X5",
                                              starts.in.df.are.Obased=FALSE)
CRMMotifInstancesGRanges <- makeGRangesFromDataFrame(CRMMotifInstances,</pre>
                                              keep.extra.columns=TRUE,
                                              ignore.strand=TRUE,
                                              seqinfo=NULL,
                                              seqnames.field="X1",
                                              start.field="X2",
                                              end.field="X3",
                                              strand.field="X5",
                                              starts.in.df.are.Obased=FALSE)
```

Lets get the overlaps between Predicted sites and ChIP-enriched regions within Regulatory modules

We select for ChIP-enriched regions within Regulatory modules as these regions are usually the ones that affect gene regulation and therefore the ones we're interested/

```
## Identifing ChIP-seq regions in regulatory modules

# SitesInRegulatoryModules <- c(subsetByOverlaps(ChIPSeqEnrichedRegionsGRanges, enhancers),

# subsetByOverlaps(ChIPSeqEnrichedRegionsGRanges, promoters))%>% unlist(

SitesInRegulatoryModules <- ChIPSeqEnrichedRegionsGRanges

OverlapPredictedChIPseq <- (findOverlaps(MotifOverlapRPredictedSitesGRanges, SitesInRegulatoryModules)%

OverlapSequneceChIPseq <- (findOverlaps(AllMotifInstancesGRanges, SitesInRegulatoryModules)%>%countLnod

OverlapPredictedSequence <- (findOverlaps(MotifOverlapRPredictedSitesGRanges, AllMotifInstancesGRanges)

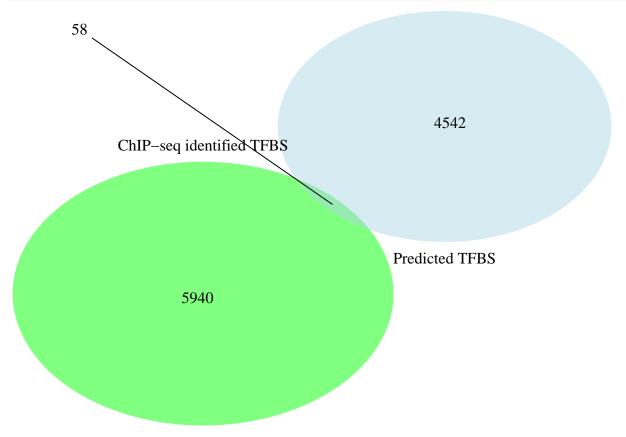
OverlapCRMEnriched <- (findOverlaps(CRMMotifInstancesGRanges, SitesInRegulatoryModules)%>%countLnodeHit

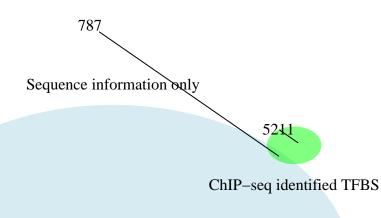
OverlapALL <- (subsetByOverlaps(subsetByOverlaps(MotifOverlapRPredictedSitesGRanges, AllMotifInstancesGRanges)
```

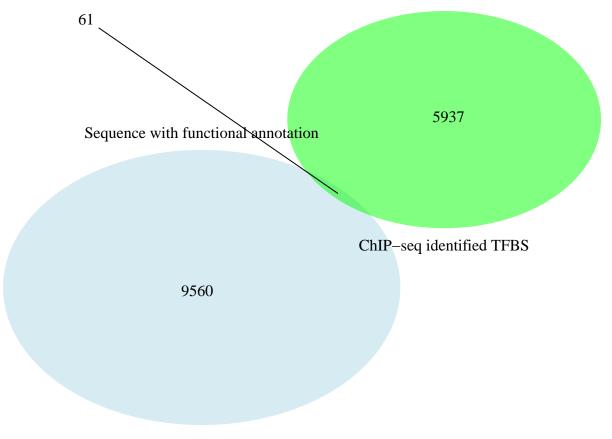
Lets make some pretty diagrams

```
grid.newpage()
draw.pairwise.venn(length(MotifOverlapRPredictedSitesGRanges),
```

```
length(SitesInRegulatoryModules),
OverlapPredictedChIPseq,
category = c("Predicted TFBS", "ChIP-seq identified TFBS"),
lty = rep("blank", 2),
fill = c("light blue", "green"),
alpha = rep(0.5, 2),
cat.pos = c(0, 180),
euler.d = TRUE,
sep.dist = 0.03,
rotation.degree = 45)
```







```
## (polygon[GRID.polygon.32], polygon[GRID.polygon.33], polygon[GRID.polygon.34], polygon[GRID.polygon.
ImprovedAccuracyRelativeToSequence <-</pre>
      function(x){
           foldImprovement <- (subsetByOverlaps(x, SitesInRegulatoryModules)%>%length() / x%>%length()) /
            (subsetByOverlaps(AllMotifInstancesGRanges, SitesInRegulatoryModules)%>%length () / AllMotifInstancesGRanges, SitesInRegulatoryModules)%>%length () / AllMotifInstancesGranges() / AllMotifInstancesGr
     }
PercentageOfSitesThatAreTruePositives<- function(x) {</pre>
         (subsetByOverlaps(x, SitesInRegulatoryModules)%>%length() / x%>%length())*100
AccuracyRelativeToSequence <- lapply(list(AllMotifInstancesGRanges,</pre>
                           CRMMotifInstancesGRanges,
                           MotifOverlapRPredictedSitesGRanges),
                     ImprovedAccuracyRelativeToSequence)%>%rbind.data.frame()%>%set_colnames(c("Sequence", "Functiona")
cbind.data.frame("Sequence" = c(
                                                length(AllMotifInstancesGRanges),
                                                length(ChIPSeqEnrichedRegionsGRanges),
                                                OverlapSequneceChIPseq,
                                                PercentageOfSitesThatAreTruePositives(AllMotifInstancesGRanges),
                                                {\tt ImprovedAccuracyRelativeToSequence(AllMotifInstancesGRanges)),}
                                                "CRM Sites" =
```

```
length(CRMMotifInstancesGRanges),
length(ChIPSeqEnrichedRegionsGRanges),
OverlapCRMEnriched,
PercentageOfSitesThatAreTruePositives(CRMMotifInstancesGRanges),
ImprovedAccuracyRelativeToSequence(CRMMotifInstancesGRanges)),
"motifOverlapR Sites" = c(
length(MotifOverlapRPredictedSitesGRanges),
length(ChIPSeqEnrichedRegionsGRanges),
OverlapPredictedChIPseq,
PercentageOfSitesThatAreTruePositives(MotifOverlapRPredictedSitesGRanges),
ImprovedAccuracyRelativeToSequence(MotifOverlapRPredictedSitesGRanges))
) %>%set_rownames(c ("Number of motifs",
                     "Number of ChIP-seq sites",
                     "Number of sites Predicted",
                     "% of predicted sites that are true positives",
                     "Fold enrichement over sequence alone")
  )%>%pander()
```

	Sequence	CRM Sites	motifOverlapR Sites
Number of motifs	463408	9621	4600
Number of ChIP-seq sites	5998	5998	5998
Number of sites Predicted	787	61	58
% of predicted sites that are true	0.1698	0.634	1.261
$\mathbf{positives}$			
Fold enrichement over sequence	1	3.733	7.424
alone			