Immunedeconv

Kevin Ryan

2023-05-08

R Markdown

```
library(biomaRt)
library(immunedeconv)
library(corrplot)
library(forcats)
library(stringr)
library(here)
library(dplyr)
library(tidyr)
library(ggplot2)
library(RColorBrewer)
```

```
annotate_genes <- function(df){</pre>
    df$hgnc_symbol <- df$gene</pre>
    mart <- useMart(biomart = "ensembl", dataset = "hsapiens_gene_ensembl") #, host="uswest.ensembl.org</pre>
    info <- getBM(attributes=c("hgnc_symbol",</pre>
                                  "ensembl_gene_id_version",
                                  "chromosome_name",
                                  "start_position",
                                  "end_position",
                                  "strand",
                                  "entrezgene_description"),
                   filters = c("hgnc_symbol"),
                   values = df$gene,
                   mart = mart,
                   useCache=FALSE)
    tmp <- merge(df, info, by="hgnc_symbol")</pre>
    tmp\$strand <- gsub("-1", "-", tmp\$strand)
    tmp$strand <- gsub("1", "+", tmp$strand)</pre>
    tmp <- tmp[!grepl("CHR", tmp$chromosome_name),]</pre>
    return(tmp)
}
convert_ensg_version_to_hgnc_df_rownames <- function(df, ensdb = EnsDb.Hsapiens.v86){</pre>
  library(AnnotationDbi)
```

```
library(ensembldb)
  library (EnsDb. Hsapiens. v86)
  genes_base <- str_split_fixed(string = rownames(df), pattern = "\\.", n = 2)[,1]</pre>
  newnames_original <- suppressWarnings(mapIds(EnsDb.Hsapiens.v86,</pre>
    keys = genes_base,
    column = 'SYMBOL',
    keytype = 'GENEID'))
# keep ensg version of newnames is na or is duplicated
newnames <- ifelse(is.na(newnames_original) | duplicated(newnames_original),</pre>
    rownames(df), newnames_original)
rownames(df) <- newnames
return(df)
}
salmon_tpm_path <- "/home/rstudio/Documents/PhD/CAF_data/nfcore_results/inhouse_data_nfcore_results_ver</pre>
salmon_tpm_in <- read.table(salmon_tpm_path, header = T)</pre>
colnames(salmon_tpm_in) <- gsub("X", "", colnames(salmon_tpm_in))</pre>
tx2gene_path = "~/Documents/PhD/subtypes/caf-subtype-analysis/nf-subpop/outdir/tx2gene/tx2gene.txt"
tx2gene <- read.table(tx2gene_path, header = T)</pre>
salmon tpm in [1:5,1:5]
                                                  3533
                                                             3536
##
                gene_id gene_name
                                        3532
## 1 ENSG0000000003.14 TSPAN6 15.867374 18.118736 17.147614
## 2 ENSG0000000005.6
                             TNMD 0.000000 0.000000 0.041895
## 3 ENSG0000000419.12
                             DPM1 76.108329 73.852506 73.578070
## 4 ENSG0000000457.14
                            SCYL3 4.126492 4.089532 3.484089
## 5 ENSG00000000460.17 Clorf112 16.378414 12.994835 12.456041
salmon_tpm_hgnc <- convert_ensg_version_to_hgnc_df_rownames(salmon_tpm_in)</pre>
## Loading required package: stats4
## Loading required package: BiocGenerics
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:dplyr':
##
##
       combine, intersect, setdiff, union
## The following objects are masked from 'package:stats':
##
##
       IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
       anyDuplicated, aperm, append, as.data.frame, basename, cbind,
##
```

```
colnames, dirname, do.call, duplicated, eval, evalq, Filter, Find,
##
##
       get, grep, grepl, intersect, is.unsorted, lapply, Map, mapply,
##
       match, mget, order, paste, pmax, pmax.int, pmin, pmin.int,
##
       Position, rank, rbind, Reduce, rownames, sapply, setdiff, sort,
       table, tapply, union, unique, unsplit, which.max, which.min
## Loading required package: Biobase
## Welcome to Bioconductor
##
##
       Vignettes contain introductory material; view with
##
       'browseVignettes()'. To cite Bioconductor, see
       'citation("Biobase")', and for packages 'citation("pkgname")'.
##
## Loading required package: IRanges
## Loading required package: S4Vectors
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:tidyr':
##
##
       expand
## The following objects are masked from 'package:dplyr':
##
##
       first, rename
## The following object is masked from 'package:utils':
##
##
       findMatches
## The following objects are masked from 'package:base':
##
##
       expand.grid, I, unname
##
## Attaching package: 'IRanges'
## The following objects are masked from 'package:dplyr':
##
##
       collapse, desc, slice
## Attaching package: 'AnnotationDbi'
## The following object is masked from 'package:dplyr':
##
##
       select
```

```
## Loading required package: GenomicRanges
## Loading required package: GenomeInfoDb
## Loading required package: GenomicFeatures
## Loading required package: AnnotationFilter
## Attaching package: 'ensembldb'
## The following object is masked from 'package:dplyr':
##
##
       filter
## The following object is masked from 'package:stats':
##
##
       filter
salmon_tpm_hgnc_not_duplicated <- salmon_tpm_hgnc[!duplicated(salmon_tpm_hgnc$gene_name),]</pre>
salmon_tpm_hgnc_not_duplicated_hgnc_col <- subset(salmon_tpm_hgnc_not_duplicated, select = -c(gene_id)
colnames(salmon_tpm_hgnc_not_duplicated_hgnc_col) <- gsub("X", "Sample_", colnames(salmon_tpm_hgnc_not_</pre>
\#write.table(salmon\_tpm\_hgnc\_not\_duplicated\_hgnc\_col, file = "/home/rstudio/Downloads/salmon\_tpm\_hgnc\_c
rownames(salmon_tpm_hgnc_not_duplicated) <- salmon_tpm_hgnc_not_duplicated$gene_name</pre>
salmon_tpm_hgnc_not_duplicated <- subset(salmon_tpm_hgnc_not_duplicated, select = -c(gene_name, gene_i
metadata <- read.csv(file = here("~/Documents/PhD/subtypes/caf-subtype-analysis/intermediate_files/meta
colnames(metadata)[1] <- "sample"</pre>
metadata$sample <- as.character(metadata$sample)</pre>
metadata$Condition <- ifelse(metadata$Condition == "Tumour", "CAF", "TAN")
metadata
      sample Patient Condition Age Size
                                           Grade Histology
        4033
                   1
                           CAF 46
                                      45 Grade_2
                                                   Lobular ER_positive PR_positive
```

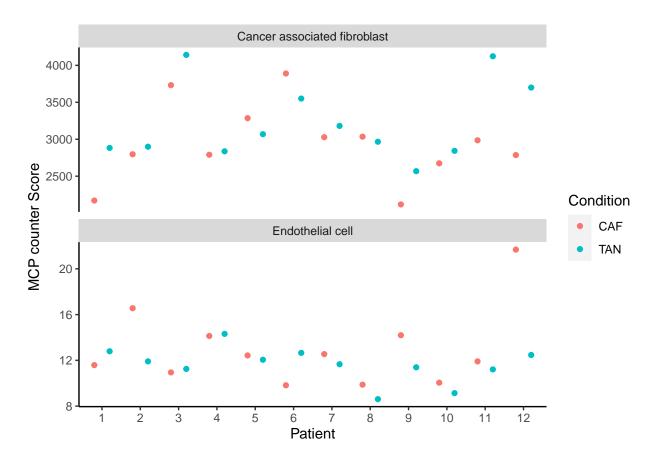
1 ## 2 4034 1 TAN 46 45 Grade_2 Lobular ER_positive PR_positive ## 3 4027 2 CAF 77 40 Grade_3 Ductal ER_negative PR_negative ## 4 4028 2 TAN 77 40 Grade_3 Ductal ER_negative PR_negative 3 ## 5 4112 CAF 62 12 Grade_3 Ductal ER_positive PR_negative ## 6 4113 3 TAN 62 8 Grade_3 Ductal ER_positive PR_negative ## 7 4116 4 CAF 45 35 Grade_2 Lobular ER_positive PR_positive ## 8 4 4117 TAN 45 13 Grade_2 Lobular ER_positive PR_positive ## 9 4214 5 CAF 78 90 Grade_2 Lobular ER_positive PR_negative 5 ## 10 4215 78 Lobular ER_positive PR_negative TAN 90 Grade_2 ## 11 4315 6 CAF 84 30 Grade_2 Ductal ER_positive PR_positive ## 12 4316 6 TAN 84 22 Grade_2 Ductal ER_positive PR_positive ## 13 4340 7 CAF 62 100 Grade_2 Lobular ER_positive PR_positive 7 ## 14 4341 TAN 62 100 Grade_2 Lobular ER_positive PR_positive 8 Ductal ER_positive PR_positive ## 15 4344 CAF 50 28 Grade 2 8 ## 16 4345 TAN 50 28 Grade_2 Ductal ER_positive PR_positive

```
## 17
               3532
                                    9
                                                   CAF
                                                             48
                                                                       16 Grade 2
                                                                                                 Ductal ER_positive PR_positive
## 18
               3533
                                   9
                                                   TAN
                                                             48
                                                                       16 Grade_2
                                                                                                 Ductal ER_positive PR_positive
## 19
               3536
                                  10
                                                   CAF
                                                             50
                                                                      52 Grade 3
                                                                                               Lobular ER_positive PR_positive
## 20
               3537
                                  10
                                                   TAN
                                                           50
                                                                      52 Grade_3
                                                                                               Lobular ER_positive PR_positive
## 21
               4299
                                  11
                                                   CAF
                                                             84
                                                                      40 Grade 3
                                                                                                 Ductal ER_positive PR_positive
## 22
               4300
                                  11
                                                   TAN
                                                             84
                                                                       40 Grade 3
                                                                                                 Ductal ER_positive PR_positive
## 23
                                                                       52 Grade 2
                                                                                               Lobular ER_positive PR_negative
               4722
                                  12
                                                   CAF
                                                             81
                                  12
                                                                                               Lobular ER_positive PR_negative
## 24
               4723
                                                   TAN 81
                                                                       52 Grade 2
##
                            Her2 Subtype
## 1 Her2_negative LuminalA LVI_negative
          Her2_negative LuminalA LVI_negative
          Her2_negative
                                             TNBC LVI_positive
## 3
## 4 Her2_negative
                                             TNBC LVI_positive
## 5 Her2_negative LuminalA LVI_negative
## 6 Her2_negative LuminalA LVI_negative
## 7
          Her2_negative LuminalA LVI_positive
## 8 Her2_negative LuminalA LVI_positive
## 9 Her2_negative LuminalA LVI_positive
## 10 Her2_negative LuminalA LVI_positive
## 11 Her2_negative LuminalA LVI_positive
## 12 Her2_negative LuminalA LVI_positive
## 13 Her2_negative LuminalA LVI_negative
## 14 Her2_negative LuminalA LVI_negative
## 15 Her2_negative LuminalA LVI_negative
## 16 Her2_negative LuminalA LVI_negative
## 17 Her2_negative LuminalA LVI_negative
## 18 Her2_negative LuminalA LVI_negative
## 19 Her2_negative LuminalA LVI_negative
## 20 Her2_negative LuminalA LVI_negative
## 21 Her2_negative LuminalA LVI_positive
## 22 Her2_negative LuminalA LVI_positive
## 23 Her2_negative LuminalA LVI_negative
## 24 Her2_negative LuminalA LVI_negative
# includes cancer-associated fibroblast
deconvolution_mcp <- immunedeconv::deconvolute(salmon_tpm_hgnc_not_duplicated, "mcp_counter")</pre>
## >>> Running mcp_counter
deconvolution_mcp_filtered_long <- deconvolution_mcp %% dplyr::filter(cell_type %in% c("Cancer associa
deconvolution_mcp_filtered_long_metadata <- full_join(deconvolution_mcp_filtered_long, metadata)
## Joining with 'by = join_by(sample)'
ord <- as.character(seq(1,12,1))
# Plotting
deconvolution_mcp_filtered_long_metadata_ordered <- deconvolution_mcp_filtered_long_metadata[order(deconvolution_mcp_filtered_long_metadata]
mcp_plot \leftarrow ggplot(deconvolution_mcp_filtered_long_metadata_ordered, aes(x = as.character(Patient), y = acceptance for the second contracts of the se
   geom_point(position = position_dodge(width = 0.8)) +
       scale x discrete(limits = ord) +
```

theme(#panel.grid.major = element_blank(),

```
#panel.grid.minor = element_blank(),
    panel.background = element_blank(),
    axis.line = element_line(colour = "black")) +
facet_wrap(~ cell_type, nrow = 2, scales = "free_y") +
xlab("Patient") +
ylab("MCP counter Score")
```

mcp_plot



```
# no fibroblast cell type
quantiseq <- immunedeconv::deconvolute(salmon_tpm_hgnc_not_duplicated, method= "quantiseq")</pre>
```

```
##
## >>> Running quantiseq

##
## Running quanTIseq deconvolution module

## Gene expression normalization and re-annotation (arrays: FALSE)
```

Removing 17 noisy genes

Removing 15 genes with high expression in tumors

```
## Signature genes found in data set: 135/138 (97.83%)
## Mixture deconvolution (method: lsei)
## Deconvolution successful!
set_cibersort_binary("/home/rstudio/sw/CIBERSORT/CIBERSORT.R")
#set_cibersort_binary("/home/rstudio/sw/CIBERSORT/CIBERSORT_1.4.R")
set_cibersort_mat("/home/rstudio/sw/CIBERSORT/LM22.txt")
#cibersort_out <- immunedeconv::deconvolute(salmon_tpm_hqnc_not_duplicated, "cibersort") # or 'cibers
# includes cancer-associated fibroblast
deconvolution_epic <- immunedeconv::deconvolute(salmon_tpm_hgnc_not_duplicated, "epic")
##
## >>> Running epic
## Warning in (function (bulk, reference = NULL, mRNA_cell = NULL, mRNA_cell_sub = NULL, : The optimiza
## 4723
## - check fit.gof for the convergeCode and convergeMessage
## Warning in (function (bulk, reference = NULL, mRNA_cell = NULL, mRNA_cell_sub =
## NULL, : mRNA_cell value unknown for some cell types: CAFs, Endothelial - using
## the default value of 0.4 for these but this might bias the true cell
## proportions from all cell types.
#cibersort_signature_matrix <- read.table("/home/rstudio/Documents/PhD/")</pre>
deconvolution_timer <- deconvolute(salmon_tpm_hgnc_not_duplicated, "timer",</pre>
            indications=rep("BRCA", 24))
## >>> Running timer
## ## Enter batch mode
## ## Loading immune gene expression
## [1] "Outlier genes: ACTB ACTG1 COL1A1 COL1A2 EEF1A1 FN1 GAPDH MT-ATP8 MT-CO1 MT2A SPARC TGFBI TIMP1"
## ## Removing the batch effect of /tmp/RtmpaoXgxF/fileeba5249320a
## Found 1033 genes with uniform expression within a single batch (all zeros); these will not be adjust
## Found2batches
## Adjusting forOcovariate(s) or covariate level(s)
## Standardizing Data across genes
```

```
## Fitting L/S model and finding priors
## Finding parametric adjustments
## Adjusting the Data
# timer default is for immune cells - no cafs
deconvolution timer
## # A tibble: 6 x 25
                    '3532' '3533' '3536' '3537' '4027' '4028' '4033' '4034' '4112'
##
     cell type
##
    <chr>
                     <dbl> <
## 1 B cell
                     0.151 0.139 0.146 0.141 0.151 0.141 0.144 0.132 0.144
## 2 T cell CD4+
                     0.167  0.162  0.161  0.163  0.163  0.164  0.164  0.165  0.159
                     0.162  0.149  0.146  0.147  0.150  0.157  0.146  0.159  0.148
## 3 T cell CD8+
## 4 Neutrophil
                     0.115  0.127  0.134  0.124  0.120  0.120  0.126  0.119  0.116
## 5 Macrophage
                     0.113 0.115 0.112 0.124 0.108 0.115 0.129 0.131 0.133
## 6 Myeloid dendri~ 0.356 0.331 0.329 0.328 0.365 0.333 0.325 0.318 0.323
## # i 15 more variables: '4113' <dbl>, '4116' <dbl>, '4117' <dbl>, '4214' <dbl>,
      '4215' <dbl>, '4299' <dbl>, '4300' <dbl>, '4315' <dbl>, '4316' <dbl>,
      '4340' <dbl>, '4341' <dbl>, '4344' <dbl>, '4345' <dbl>, '4722' <dbl>,
## #
       '4723' <db1>
deconvolution_xcell <- immunedeconv::deconvolute(salmon_tpm_hgnc_not_duplicated, "xcell")</pre>
## >>> Running xcell
colnames(deconvolution_xcell) <- gsub("X", "", colnames(deconvolution_xcell))</pre>
# cancer-associated fibroblast, hematopoeitic stem cell, microenvironment score, stroma score
deconvolution_xcell_long <- deconvolution_xcell %% pivot_longer(!cell_type, names_to = "sample", value
# ggplot(data = deconvolution_xcell, aes(x = cell_type, y = xcell_score)) +
   geom_bar(stat = "identity", alpha = 0.7) +
#
   facet_grid(. ~sample) +
   ylim(0,800) +
   geom_text(aes(label = Freq), fontface = "bold", vjust = 1.5, colour = "white", size = 4) +
  labs(x = "\n Coin Flip Outcome", y = "Frequency\n", title = "\n Coin Flip Results \n") +
#
   theme(plot.title = element_text(hjust = 0.5),
#
          axis.title.x = element_text(face="bold", colour="darkgreen", size = 12),
#
          axis.title.y = element_text(face="bold", colour="darkgreen", size = 12),
          legend.title = element_text(face="bold", size = 10),
          strip.background = element_rect(fill="lightblue", colour="black", size=1),
          strip.text = element_text(face="bold", size=rel(1.2)))
# Create the applot object
ggplot(deconvolution_xcell_long, aes(x = sample, y = xcell_score)) +
  # Add the facet wrap
 facet_wrap(~ cell_type, ncol = 5) +
  # Add the points
 geom_point() +
```

```
# Add the x-axis label
xlab("Sample") +
# Add the y-axis label
ylab("Y Axis Value")
```

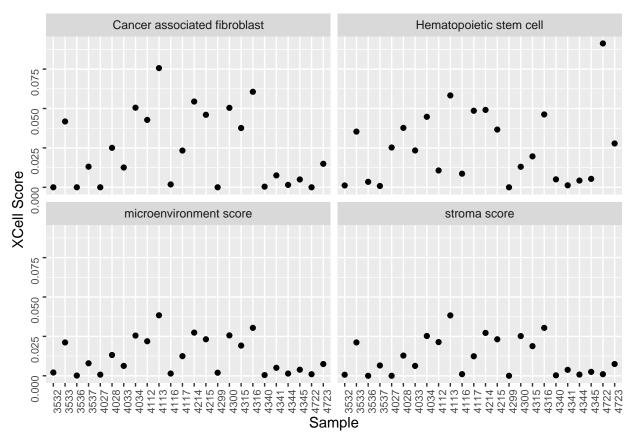
		B cell	B cell memory	B cell naive	B cell plasma	cer associated fibrol
Y Axis Value						
	0.075	-switched memory	mon lymphoid proge	nmon myeloid proge	Endothelial cell	Eosinophil
	0.075	ocyte-monocyte pro	ematopoietic stem co	immune score	Macrophage	Macrophage M1
	۵ ۵۷5 -	Macrophage M2	Mast cell	icroenvironment sco	Monocyte	Myeloid dendritic cel
		oid dendritic cell acti -	Neutrophil	NK cell	smacytoid dendritic	stroma score
		************				Andrew Strategic
	Q.QZ5:	CD4+ (non-regula -	Il CD4+ central men	Il CD4+ effector mer	T cell CD4+ memory	T cell CD4+ naive
		T cell CD4+ Th1	T cell CD4+ Th2	T cell CD8+	III CD8+ central men	II CD8+ effector mer
	8:898		T call gamma dalta	T and NIZ	all requileters / Tra	\$\$\$\$\tag{\text{\tiny{\text{\tiny{\tiny{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\texitinx}\\ \text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tinit}\\ \text{\tin\tint{\text{\text{\tinit}\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\texi}\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\ti}\tint{\text{\ti}\titit{\text{\texi}\tint{\text{\ti}\titt{\text{\titil\titt{\text{\ti}\titit
	9:935	T cell CD8+ naive	T cell gamma delta	T cell NK	cell regulatory (Treg	
			```````			23
				Sample		

deconvolution_xcell_filtered_long <- deconvolution_xcell %>% dplyr::filter(cell_type %in% c("Cancer ass deconvolution_xcell_filtered_long_metadata <- full_join(deconvolution_xcell_filtered_long, metadata)

```
## Joining with 'by = join_by(sample)'
```

#deconvolution_xcell %>% pivot_longer(!cell_type, names_to = "sample", values_to = "xcell_score")

```
ggplot(deconvolution_xcell_filtered_long, aes(x = sample, y = xcell_score)) +
# Add the facet wrap
facet_wrap(~ cell_type, ncol = 2) +
# Add the points
geom_point() +
# Add the x-axis label
xlab("Sample") +
# Add the y-axis label
ylab("XCell Score") +
theme(axis.text = element_text(size = 8, angle = 90))
```

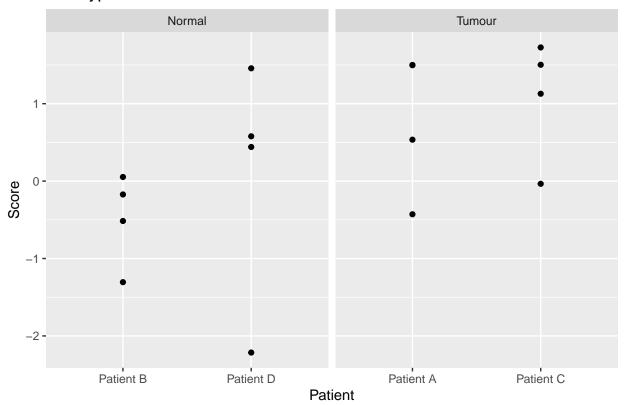


```
xcell_plot <- deconvolution_xcell_filtered_long_metadata %>%
  mutate(Condition = fct_relevel(Condition, "TAN", "CAF")) %>%
  ggplot(aes(x = Patient, y = xcell_score)) +
  #geom_col() +
  facet_wrap(~Patient, nrow=1, scales = "free_x", strip.position = "bottom") +
    facet_wrap(~cell_type, ncol = 2) +
    #theme(plot.title = element_text(hjust = 0.5, size = 14, face = "bold"),
          #axis.text.y = element_text(size = 12, face = "bold"),
          #axis.text.x = element_text(size = 8),
          #strip.text.y = element_text(size = rel(100))
          #) +
  #ggtitle("CAF subpopulation proportions\n determined by CIBERSORTx") +
    scale_fill_manual(name=NULL,
                    values = c(brewer.pal(3, "Dark2"), "gray")
  xlab(label = "Patient") +
  ylab("XCell score") +
  theme(#panel.grid.major = element_blank(),
        #panel.grid.minor = element_blank(),
       panel.background = element_blank(),
       axis.line = element_line(colour = "black")) +
     \# scale_y_continuous(expand = c(0,0)) +
  #geom_hline(yintercept = c(25, 50, 75), color = "gray", linetype = "dashed") +
  theme(legend.text=element_text(size=rel(1.2)),
       axis.text.x = element_text(size = 8, angle = 90))
```

```
# Example data frame
df <- data.frame(
    Patient = rep(c("Patient A", "Patient B", "Patient C", "Patient D"), 4),
    CellType = rep(c("Cell Type 1", "Cell Type 2", "Cell Type 3", "Cell Type 4"), each = 4),
    Status = rep(c("Tumour", "Normal"), times = 8),
    Score = rnorm(16)
)

# Plotting
ggplot(df, aes(x = Patient, y = Score)) +
    geom_point() +
    facet_wrap(~ CellType, scales = "free_y") +
    facet_grid(~ Status, scales = "free_x") +
    xlab("Patient") +
    ylab("Score") +
    ggtitle("Cell Type Scores")</pre>
```

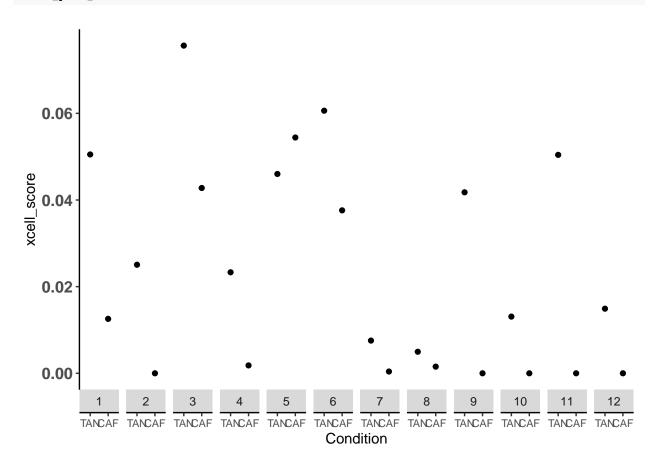
Cell Type Scores



```
xcell_plot_test <- deconvolution_xcell_filtered_long_metadata %>%
    dplyr::filter(cell_type == "Cancer associated fibroblast") %>%
    mutate(Condition = fct_relevel(Condition, "TAN", "CAF")) %>%
    ggplot(aes(x = Condition, y = xcell_score)) +
    geom_point() +
    facet_wrap(~Patient, nrow=1, scales = "free_x", strip.position = "bottom") +
        theme(plot.title = element_text(hjust = 0.5, size = 14, face = "bold"),
        axis.text.y = element_text(size = 12, face = "bold"),
```

```
axis.text.x = element_text(size = 8),
        strip.text.y = element_text(size = rel(100))
#ggtitle("CAF subpopulation proportions\n determined by CIBERSORTx") +
    scale_fill_manual(name=NULL,
                    values = c(brewer.pal(3, "Dark2"), "gray")
#
                    ) +
#xlab(label = "Patient") +
theme(#panel.grid.major = element_blank(),
       #panel.grid.minor = element_blank(),
       panel.background = element_blank(),
      axis.line = element_line(colour = "black")) +
    # scale y continuous(expand = c(0,0.08)) +
#geom_hline(yintercept = c(25, 50, 75), color = "gray", linetype = "dashed") +
theme(legend.text=element_text(size=rel(1.2)),
     axis.text.x = element_text(size = 8))
```

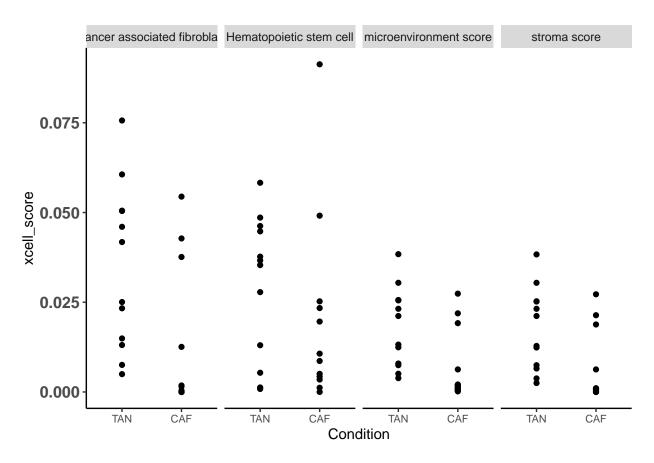
xcell_plot_test



```
xcell_plot_test2 <- deconvolution_xcell_filtered_long_metadata %>%
#dplyr::filter(cell_type == "Cancer associated fibroblast") %>%
mutate(Condition = fct_relevel(Condition, "TAN", "CAF")) %>%
ggplot(aes(x = Condition, y = xcell_score)) +
geom_point() +
facet_wrap(~Patient, nrow=1, scales = "free_x", strip.position = "bottom") +
```

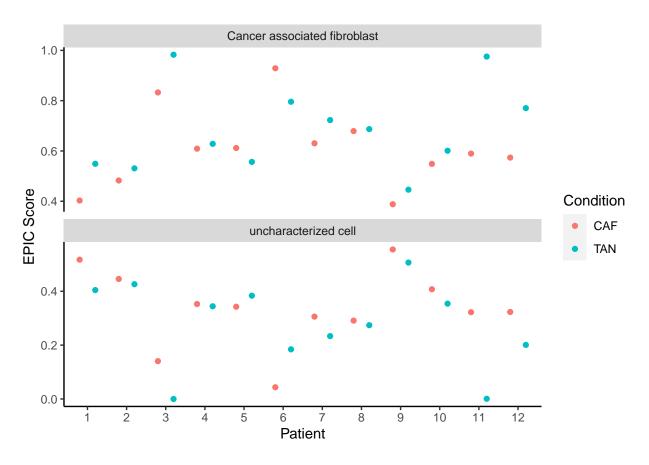
```
theme(plot.title = element_text(hjust = 0.5, size = 14, face = "bold"),
        axis.text.y = element_text(size = 12, face = "bold"),
        axis.text.x = element_text(size = 8),
        strip.text.y = element_text(size = rel(100))
        ) +
 facet_grid(~ cell_type, scales = "free_y") +
#ggtitle("CAF subpopulation proportions\n determined by CIBERSORTx") +
    scale_fill_manual(name=NULL,
                    values = c(brewer.pal(3, "Dark2"), "gray")
#xlab(label = "Patient") +
theme(#panel.grid.major = element_blank(),
       #panel.grid.minor = element_blank(),
      panel.background = element_blank(),
      axis.line = element_line(colour = "black")) +
    \# scale_y_continuous(expand = c(0,0.08)) +
#geom_hline(yintercept = c(25, 50, 75), color = "gray", linetype = "dashed") +
theme(legend.text=element_text(size=rel(1.2)),
      axis.text.x = element_text(size = 8))
```

xcell_plot_test2

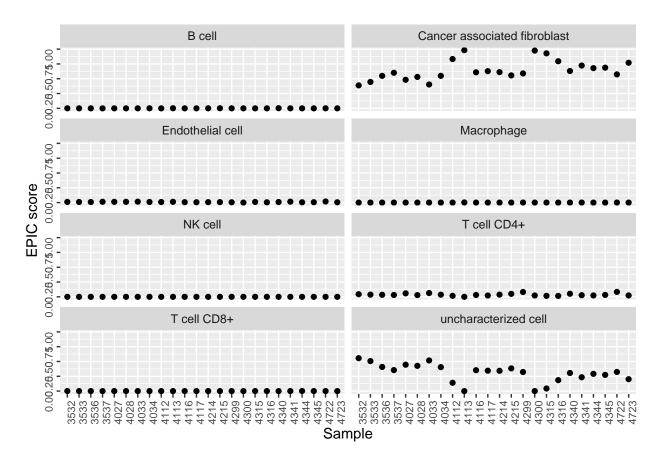


```
# Example data frame
ord <- as.character(seq(1,12,1))</pre>
```

```
# Plotting
deconvolution_xcell_filtered_long_metadata_ordered <- deconvolution_xcell_filtered_long_metadata[order(
xcell plot \leftarrow ggplot(deconvolution xcell filtered long metadata ordered, aes(x = as.character(Patient),
  geom_point(position = position_dodge(width = 0.8)) +
    scale_x_discrete(limits = ord) +
    theme(#panel.grid.major = element_blank(),
         #panel.grid.minor = element_blank(),
         panel.background = element_blank(),
         axis.line = element_line(colour = "black")) +
  facet_wrap(~ cell_type, nrow = 2, scales = "free_y") +
  xlab("Patient") +
  ylab("xCell Score")
colnames(deconvolution_epic) <- gsub("X", "", colnames(deconvolution_epic))</pre>
deconvolution_epic_long <- deconvolution_epic %% pivot_longer(!cell_type, names_to = "sample", values_</pre>
deconvolution_epic_filtered_long <- deconvolution_epic %% dplyr::filter(cell_type %in% c("Cancer assoc
deconvolution_epic_filtered_long_metadata <- full_join(deconvolution_epic_filtered_long, metadata)
## Joining with 'by = join_by(sample)'
# Plotting
deconvolution_epic_filtered_long_metadata_ordered <- deconvolution_epic_filtered_long_metadata[order(de
epic_plot <- ggplot(deconvolution_epic_filtered_long_metadata_ordered, aes(x = as.character(Patient), y
  geom_point(position = position_dodge(width = 0.8)) +
    scale_x_discrete(limits = ord) +
   theme(#panel.grid.major = element_blank(),
         #panel.grid.minor = element blank(),
         panel.background = element_blank(),
         axis.line = element_line(colour = "black")) +
  facet_wrap(~ cell_type, nrow = 2, scales = "free_y") +
  xlab("Patient") +
  ylab("EPIC Score")
epic_plot
```



```
ggplot(deconvolution_epic_long, aes(x = sample, y = epic_score)) +
# Add the facet wrap
facet_wrap(~ cell_type, ncol = 2) +
# Add the points
geom_point() +
# Add the x-axis label
xlab("Sample") +
# Add the y-axis label
ylab("EPIC score") +
theme(axis.text = element_text(size = 8, angle = 90))
```



```
deconvolution_epic_long_metadata <- full_join(deconvolution_epic_long, metadata)</pre>
```

```
## Joining with 'by = join_by(sample)'
```

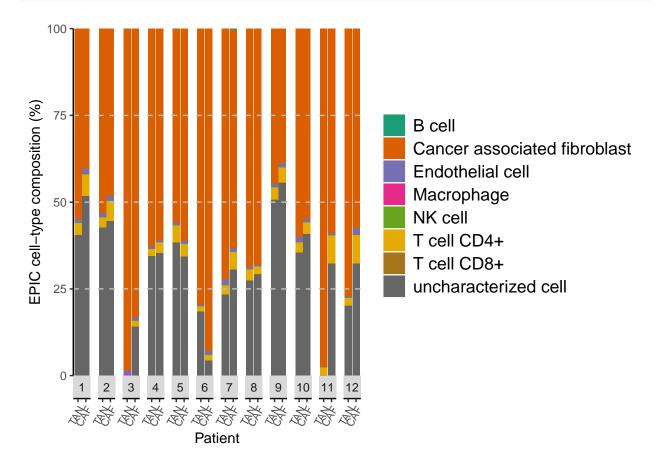
```
\#deconvolution\_epic\_filtered\_long\_metadata\$cell\_type <- relevel(deconvolution\_epic\_filtered\_long\_metadata\$cell\_type <- relevel(deconvolution\_epic\_filtered\_long\_metadata*cell\_type <- relevel(deconvolution\_epic\_filtered\_long\_filtered\_long\_filtered\_long\_filtered\_long\_filtered\_long\_f
epic plot stack <- deconvolution epic long metadata %>%
      mutate(Condition = fct_relevel(Condition, "TAN", "CAF")) %>%
      mutate(epic_score = 100*epic_score) %>%
      ggplot(aes(x = Condition, y = epic_score, fill = cell_type)) +
      geom_col() +
      facet_wrap(~Patient, nrow=1, scales = "free_x", strip.position = "bottom") +
              #theme(plot.title = element_text(hjust = 0.5, size = 14, face = "bold"),
                                  #axis.text.y = element_text(size = 12, face = "bold"),
                                  \#axis.text.x = element\_text(size = 8),
                                  #strip.text.y = element_text(size = rel(100))
                                  #) +
      \#ggtitle("CAF \ subpopulation \ proportions \ n \ determined \ by \ CIBERSORTx") \ +
             scale_fill_manual(name=NULL,
                                                                   values = c(brewer.pal(8, "Dark2"))
      xlab(label = "Patient") +
      ylab("EPIC cell-type composition (%)") +
      theme(#panel.grid.major = element blank(),
                           #panel.grid.minor = element_blank(),
```

```
panel.background = element_blank(),
    axis.line = element_line(colour = "black")) +
    scale_y_continuous(expand = c(0,0)) +
    geom_hline(yintercept = c(25, 50, 75), color = "gray", linetype = "dashed") +
    theme(legend.text=element_text(size=rel(1.2)),
        axis.text.x = element_text(size = 8))

# labs(tag = "TAN = Tumour-associated normal\nCAF=Cancer-associated fibroblast") +
    # theme(plot.tag.position = c(0.9, 0.3),)

# scale_y_continuous(expand = c(0, 0), limits = c(0, 1.0000001))
```

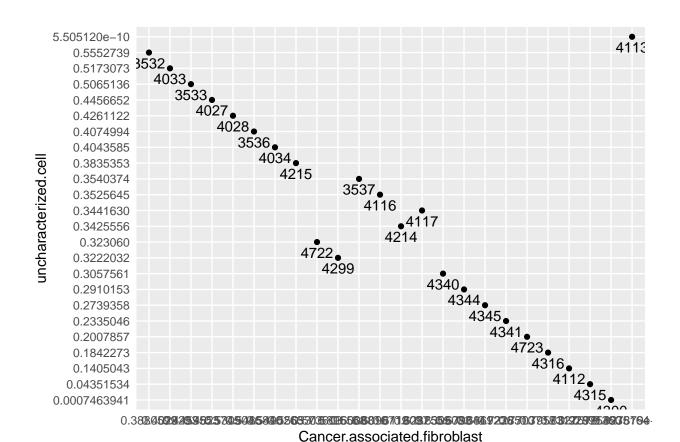
```
epic_plot_stack <- epic_plot_stack + theme(axis.text.x=element_text(angle=60, hjust=1))
epic_plot_stack</pre>
```



cor.test(as.numeric(deconvolution_epic[which(deconvolution_epic\$cell_type == "Cancer associated fibrobl

```
##
## Pearson's product-moment correlation
##
## data: as.numeric(deconvolution_epic[which(deconvolution_epic$cell_type == "Cancer associated fibrob
## t = -42.528, df = 22, p-value < 2.2e-16</pre>
```

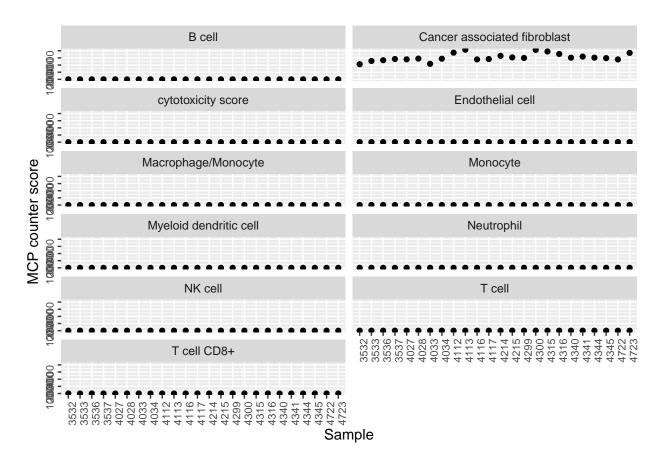
```
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.9974334 -0.9858799
## sample estimates:
          cor
## -0.9939729
as.numeric(deconvolution_epic[which(deconvolution_epic$cell_type == "Cancer associated fibroblast"),c(2
## [1] 0.3880599 0.4459553 0.5485840 0.6008816 0.4825745 0.5309046 0.4028293
## [8] 0.5490263 0.8327579 0.9827879 0.6090719 0.6282559 0.6118097 0.5565316
## [15] 0.5895568 0.9754935 0.9289136 0.7957312 0.6305083 0.7228510 0.6790441
## [22] 0.6869236 0.5736030 0.7703718
\#cor.test(cibersort\_caf\_subpopulation\_results\$S1, as.numeric(deconvolution\_epic[which(deconvolution\_epic])
#deconvolution_epic_filter_long <- deconvolution_epic %>% dplyr::filter(cell_type %in% c("Cancer associ
deconvolution_epic_filter <- deconvolution_epic %% dplyr::filter(cell_type %in% c("Cancer associated f
deconvolution_epic_filter <- t(deconvolution_epic_filter)</pre>
colnames(deconvolution_epic_filter) <- deconvolution_epic_filter[1,]</pre>
deconvolution_epic_filter <- data.frame(deconvolution_epic_filter[-c(1),])</pre>
deconvolution_epic_filter <- deconvolution_epic_filter %>% mutate(Sample = rownames(deconvolution_epic_
ggplot(data = deconvolution_epic_filter) +
  geom_point(mapping = aes(x = Cancer.associated.fibroblast, y = uncharacterized.cell)) +
   \verb|theme| (\#axis.text.x=element\_blank(), \#remove x axis labels| \\
        axis.ticks.x=element_blank(), #remove x axis ticks
        #axis.text.y=element_blank(), #remove y axis labels
        axis.ticks.y=element_blank() #remove y axis ticks
  \#geom\_text(data=subset(deconvolution\_epic\_filter, Sample == "4113"), aes(x = Cancer.associated.fibrob)
  geom_text(data=deconvolution_epic_filter, aes(x = Cancer.associated.fibroblast, y = uncharacterized.c
```



cibersort_caf_subpopulation_results <- read.csv("/home/rstudio/Documents/PhD/subtypes/caf-subtype-analy</pre>

```
cibersort_caf_subpopulation_results$Mixture <- gsub("X", "",cibersort_caf_subpopulation_results$Mixture
colnames(cibersort_caf_subpopulation_results)[1] <- "Sample"
cibersort_caf_subpopulation_results <- cibersort_caf_subpopulation_results %>% dplyr::select(-c("P.valu
```

```
deconvolution_mcp_long <- deconvolution_mcp %% pivot_longer(!cell_type, names_to = "sample", values_to
deconvolution_mcp_long$sample <- gsub("X", replacement = "", deconvolution_mcp_long$sample)
ggplot(deconvolution_mcp_long, aes(x = sample, y = mcp_counter_score)) +
    # Add the facet wrap
facet_wrap(~ cell_type, ncol = 2) +
    # Add the points
geom_point() +
    # Add the x-axis label
xlab("Sample") +
    # Add the y-axis label
ylab("MCP counter score") +
theme(axis.text = element_text(size = 8, angle = 90))</pre>
```



```
deconvolution_mcp_filter <- data.frame(t(deconvolution_mcp[which(deconvolution_mcp$cell_type =="Cancer colnames(deconvolution_mcp_filter) <- deconvolution_mcp_filter[1,]
deconvolution_mcp_filter <- data.frame(deconvolution_mcp_filter[-1,])
deconvolution_mcp_filter$Sample <- colnames(deconvolution_mcp)[2:ncol(deconvolution_mcp)]
colnames(deconvolution_mcp_filter)[1] <- "mcp_counter_score_CAF"</pre>
```

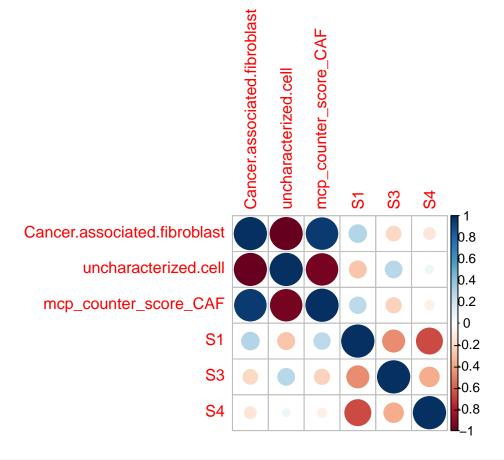
deconvolution_mcp_filter

```
##
      mcp_counter_score_CAF Sample
## 1
                    2118.187
                                3532
                    2568.089
## 2
                                3533
## 3
                    2674.509
                                3536
## 4
                    2843.439
                                3537
## 5
                    2797.128
                                4027
## 6
                    2899.125
                                4028
## 7
                    2170.477
                                4033
## 8
                    2883.104
                                4034
## 9
                    3732.086
                                4112
                    4141.333
## 10
                                4113
## 11
                     2790.22
                                4116
## 12
                     2836.76
                                4117
## 13
                    3285.481
                                4214
## 14
                    3068.483
                                4215
## 15
                    2985.898
                                4299
                    4122.946
                                4300
## 16
```

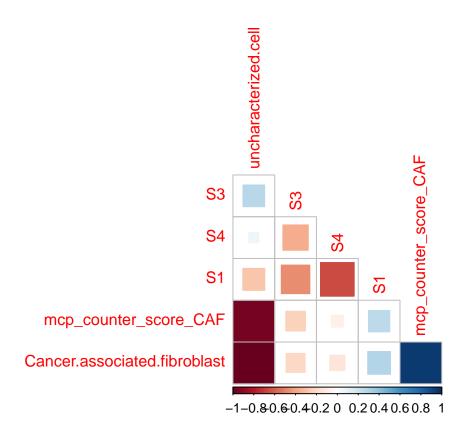
```
3889.782
## 17
                                4315
## 18
                    3550.484
                                4316
## 19
                                4340
                    3027.929
                    3180.459
                                4341
## 20
## 21
                    3035.335
                                4344
## 22
                    2966.457
                                4345
## 23
                    2786.021
                                4722
## 24
                    3699.696
                                4723
```

deconvolution_epic_cibersort_mcp_combined <- full_join(deconvolution_epic_filter, deconvolution_mcp_fil deconvolution_epic_cibersort_mcp_combined <- full_join(deconvolution_epic_cibersort_mcp_combined, ciber rownames(deconvolution_epic_cibersort_mcp_combined) <- deconvolution_epic_cibersort_mcp_combined \$Sample deconvolution_epic_cibersort_mcp_combined <- deconvolution_epic_cibersort_mcp_combined %>% dplyr::selections

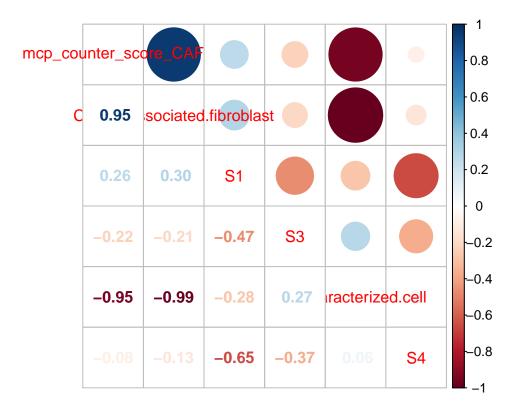
corrplot(cor(deconvolution_epic_cibersort_mcp_combined))



corrplot(cor(deconvolution_epic_cibersort_mcp_combined), method = 'square', order = 'FPC', type = 'lower'



corrplot.mixed(cor(deconvolution_epic_cibersort_mcp_combined), order = 'AOE')



deconvolution_estimate <- immunedeconv::deconvolute_estimate(salmon_tpm_hgnc_not_duplicated)

```
## [1] "Merged dataset includes 9963 genes (449 mismatched)."
```

deconvolution_estimate

```
##
                          3532
                                         3533
                                                     3536
                                                                 3537
                                                                                4027
## StromalScore
                  -398.9779264
                                 -77.0901415 -367.80715 -234.41848
                                                                      -421.6918121
                 -1357.1717989 -1241.8458068 -1266.27719 -1320.35708 -1187.1899556
## ImmuneScore
## ESTIMATEScore -1756.1497253 -1318.9359483 -1634.08434 -1554.77556 -1608.8817677
## TumorPurity
                     0.9403275
                                   0.9165688
                                                  0.93408
                                                              0.92986
                                                                          0.9327527
                                         4033
##
                          4028
                                                       4034
                                                                     4112
## StromalScore
                  -310.5104970 -145.7250766
                                               -10.3216652
                                                              -20.8256033
                 -1313.4536825 -1220.6344024 -1155.0647891 -1235.5174420
## ImmuneScore
## ESTIMATEScore -1623.9641796 -1366.3594790 -1165.3864543 -1256.3430453
##
  TumorPurity
                     0.9335485
                                   0.9193302
                                                  0.9073238
                                                                0.9128561
##
                          4113
                                         4116
                                                       4117
                                                                     4214
## StromalScore
                   152.0644980
                               -300.7698916
                                              -249.6659772
                                                              -96.2201331
## ImmuneScore
                 -1139.3841557 -1288.4770855 -1219.8616524 -1155.4397974
## ESTIMATEScore
                  -987.3196577 -1589.2469771 -1469.5276296 -1251.6599305
                                   0.9317097
## TumorPurity
                     0.8960258
                                                  0.9251835
                                                                0.9125752
##
                          4215
                                         4299
                                                       4300
                                                                     4315
## StromalScore
                  -172.1903869 -236.2956163
                                              -71.9668837
                                                               21.6951624
```

^{## [1] &}quot;1 gene set: StromalSignature overlap= 136"

^{## [1] &}quot;2 gene set: ImmuneSignature overlap= 140"

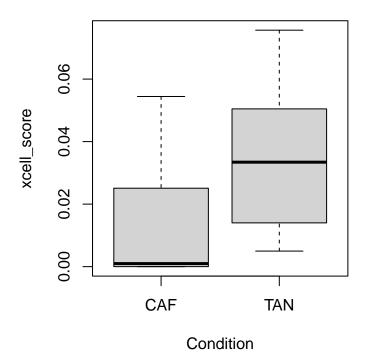
```
-1212.7012973 -1171.9712346 -1201.7329345 -1142.4733879
## ESTIMATEScore -1384.8916842 -1408.2668509 -1273.6998182 -1120.7782255
## TumorPurity
                     0.9203972
                                   0.9217333
                                                 0.9138934
                                                               0.9045514
##
                          4316
                                        4340
                                                      4341
                                                                    4344
## StromalScore
                    30.4199104 -274.1544943
                                             -173.9367732
                                                            -255.1282555
## ImmuneScore
                -1138.2189328 -1236.3081962 -1193.6425982 -1300.0417197
## ESTIMATEScore -1107.7990224 -1510.4626906 -1367.5793713 -1555.1699753
                                   0.9274472
                                                 0.9194006
## TumorPurity
                     0.9037374
                                                               0.9298813
##
                          4345
                                        4722
                                                     4723
## StromalScore
                  -367.5866878 -240.7848074 -165.711409
## ImmuneScore
                -1326.4001046 -981.5805628 -1159.640857
## ESTIMATEScore -1693.9867924 -1222.3653702 -1325.352266
## TumorPurity
                     0.9371835
                                   0.9108085
                                                 0.916945
```

deconvolution_xcell_filtered_long_metadata_caf <- deconvolution_xcell_filtered_long_metadata %>% dplyr: summary(deconvolution_xcell_filtered_long_metadata_caf)

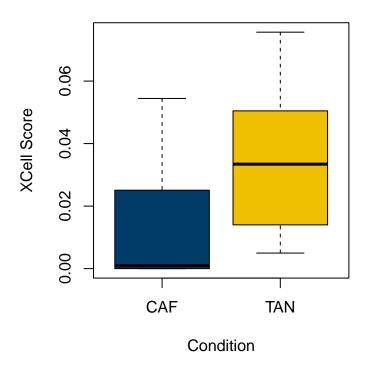
```
##
                          sample
                                                                 Patient
     cell_type
                                           xcell_score
                                                 :0.000000
##
  Length:24
                       Length:24
                                          Min.
                                                             Min.
                                                                   : 1.00
## Class :character
                       Class : character
                                          1st Qu.:0.001253
                                                              1st Qu.: 3.75
                                                             Median: 6.50
                                          Median :0.014000
## Mode :character
                       Mode :character
##
                                                 :0.023543
                                                             Mean : 6.50
                                          Mean
##
                                          3rd Qu.:0.043588
                                                              3rd Qu.: 9.25
##
                                                  :0.075644
                                                             Max.
                                                                    :12.00
                                          Max.
##
    Condition
                                            Size
                                                            Grade
                            Age
                              :45.00
##
  Length:24
                                       Min. : 8.00
                                                        Length:24
                       Min.
   Class : character
                       1st Qu.:49.50
                                       1st Qu.: 26.50
                                                         Class : character
   Mode :character
                                       Median : 40.00
##
                       Median :62.00
                                                        Mode :character
##
                       Mean
                              :63.92
                                             : 43.58
                                       Mean
##
                       3rd Qu.:78.75
                                       3rd Qu.: 52.00
                              :84.00
                                              :100.00
##
                       Max.
                                       Max.
                            ER
                                               PR
##
    Histology
                                                                  Her2
   Length:24
##
                       Length:24
                                          Length:24
                                                             Length:24
                                                              Class : character
##
   Class :character
                       Class :character
                                          Class : character
   Mode :character
                       Mode :character
                                          Mode :character
                                                             Mode : character
##
##
##
##
      Subtype
                           LVI
##
   Length:24
                       Length:24
##
   Class :character
                       Class : character
##
   Mode :character
                       Mode :character
##
##
##
```

```
## Pearson's product-moment correlation
##
## data: caf_xcell and tan_xcell
## t = 3.0238, df = 10, p-value = 0.01281
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1942461 0.9057587
## sample estimates:
## cor
## 0.6911047

par(pty = "s")
boxplot(xcell_score ~ Condition, data = deconvolution_xcell_filtered_long_metadata_caf)
```



XCell CAF score InHouse CAF vs TAN



```
diff = caf_xcell - tan_xcell
summary(diff)
               1st Qu.
                          Median
                                      Mean
                                             3rd Qu.
## -0.050423 -0.034136 -0.022248 -0.021894 -0.011600 0.008418
bartlett.test(xcell_score ~ Condition, data = deconvolution_xcell_filtered_long_metadata_caf)
##
   Bartlett test of homogeneity of variances
##
##
## data: xcell_score by Condition
## Bartlett's K-squared = 0.1614, df = 1, p-value = 0.6879
t.test(formula = xcell_score ~ Condition,
       alternative = "two.sided",
       mu = 0,
       paired = TRUE,
       var.equal = TRUE,
       conf.level = 0.95, data = deconvolution_xcell_filtered_long_metadata_caf)
##
##
   Paired t-test
```

##

```
## data: xcell_score by Condition
## t = -4.4452, df = 11, p-value = 0.0009865
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.03273521 -0.01105366
## sample estimates:
## mean difference
## -0.02189443
```

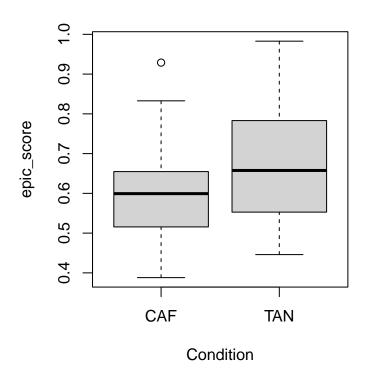
deconvolution_epic_filtered_long_metadata

```
## # A tibble: 48 x 14
      cell_type
                    sample epic_score Patient Condition
                                                          Age Size Grade Histology
                                                        <int> <int> <chr> <chr>
##
      <chr>
                    <chr>
                                <dbl>
                                        <int> <chr>
   1 Cancer assoc~ 3532
                                0.388
                                            9 CAF
                                                           48
                                                                 16 Grad~ Ductal
## 2 Cancer assoc~ 3533
                                0.446
                                            9 TAN
                                                           48
                                                                 16 Grad~ Ductal
## 3 Cancer assoc~ 3536
                                0.549
                                           10 CAF
                                                           50
                                                                 52 Grad~ Lobular
## 4 Cancer assoc~ 3537
                                0.601
                                           10 TAN
                                                           50
                                                                 52 Grad~ Lobular
## 5 Cancer assoc~ 4027
                                0.483
                                           2 CAF
                                                           77
                                                                 40 Grad~ Ductal
                                                           77
## 6 Cancer assoc~ 4028
                                0.531
                                            2 TAN
                                                                 40 Grad~ Ductal
## 7 Cancer assoc~ 4033
                                0.403
                                            1 CAF
                                                           46
                                                                 45 Grad~ Lobular
## 8 Cancer assoc~ 4034
                                0.549
                                            1 TAN
                                                           46
                                                                 45 Grad~ Lobular
## 9 Cancer assoc~ 4112
                                0.833
                                            3 CAF
                                                           62
                                                                 12 Grad~ Ductal
## 10 Cancer assoc~ 4113
                                0.983
                                           3 TAN
                                                           62
                                                                  8 Grad~ Ductal
## # i 38 more rows
## # i 5 more variables: ER <chr>, PR <chr>, Her2 <chr>, Subtype <chr>, LVI <chr>
```

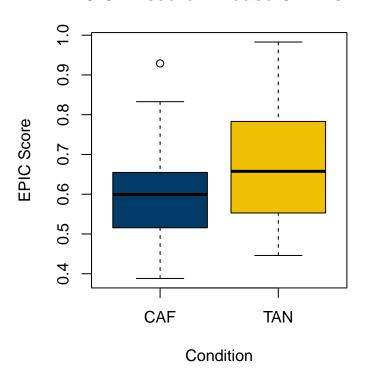
deconvolution_epic_filtered_long_metadata_caf <- deconvolution_epic_filtered_long_metadata %>% dplyr::f
summary(deconvolution_epic_filtered_long_metadata_caf)

```
##
                         sample
                                          epic_score
                                                            Patient
    cell_type
## Length:24
                                                         Min. : 1.00
                      Length:24
                                               :0.3881
                                         Min.
## Class :character
                      Class : character
                                         1st Qu.:0.5489
                                                         1st Qu.: 3.75
## Mode :character Mode :character
                                                         Median : 6.50
                                        Median :0.6104
##
                                         Mean
                                               :0.6468
                                                         Mean : 6.50
##
                                         3rd Qu.:0.7347
                                                         3rd Qu.: 9.25
##
                                         Max.
                                               :0.9828
                                                         Max. :12.00
##
    Condition
                                                         Grade
                           Age
                                          Size
## Length:24
                            :45.00
                                           : 8.00
                                                     Length:24
                      Min.
                                     Min.
## Class :character
                      1st Qu.:49.50
                                      1st Qu.: 26.50
                                                      Class : character
## Mode :character
                      Median :62.00
                                      Median : 40.00
                                                      Mode :character
##
                             :63.92
                                           : 43.58
                      Mean
                                      Mean
##
                      3rd Qu.:78.75
                                      3rd Qu.: 52.00
##
                      Max.
                             :84.00
                                      Max.
                                            :100.00
##
                           ER
                                             PR
    Histology
                                                               Her2
## Length:24
                      Length:24
                                         Length:24
                                                           Length:24
   Class :character
                                         Class :character
##
                      Class :character
                                                           Class : character
  Mode :character
                      Mode :character
                                        Mode :character
                                                           Mode :character
##
##
##
##
                          LVI
     Subtype
##
  Length:24
                      Length:24
```

```
Class :character
                       Class : character
    Mode :character
##
                       Mode :character
##
##
##
caf_epic <- deconvolution_epic_filtered_long_metadata_caf$epic_score[which(deconvolution_epic_filtered_</pre>
tan_epic <- deconvolution_epic_filtered_long_metadata_caf$epic_score[which(deconvolution_epic_filtered_
cor.test(x = caf_epic, y = tan_epic,
         method = c("pearson"),
         conf.level = 0.95)
##
##
   Pearson's product-moment correlation
##
## data: caf_epic and tan_epic
## t = 2.9167, df = 10, p-value = 0.01539
\mbox{\tt \#\#} alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1703742 0.9012231
## sample estimates:
##
         cor
## 0.6779839
par(pty = "s")
boxplot(epic_score ~ Condition, data = deconvolution_epic_filtered_long_metadata_caf)
```



EPIC CAF score InHouse CAF vs TAN



```
##
## Bartlett test of homogeneity of variances
##
## data: epic_score by Condition
## Bartlett's K-squared = 0.075341, df = 1, p-value = 0.7837
```

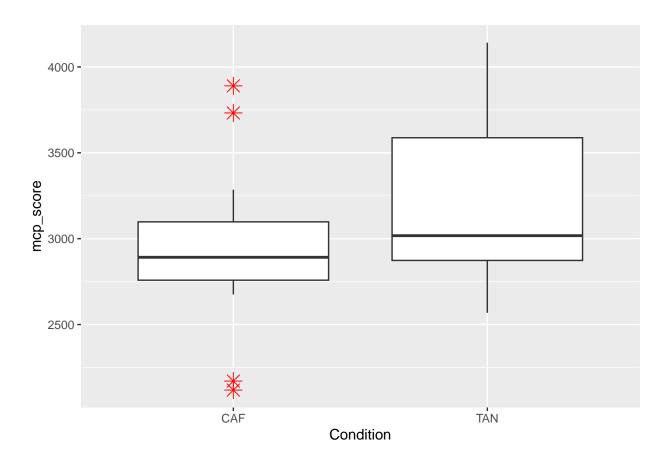
bartlett.test(epic_score ~ Condition, data = deconvolution_epic_filtered_long_metadata_caf)

```
##
## Paired t-test
##
## data: epic_score by Condition
```

```
## t = -2.1183, df = 11, p-value = 0.05774
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.164548496 0.003148298
## sample estimates:
## mean difference
        -0.0807001
##
deconvolution_mcp_filtered_long_metadata
## # A tibble: 48 x 14
##
      cell_type
                     sample mcp_score Patient Condition
                                                           Age Size Grade Histology
##
      <chr>
                     <chr>
                                <dbl>
                                        <int> <chr>
                                                         <int> <int> <chr> <chr>
## 1 Endothelial c~ 3532
                                14.2
                                            9 CAF
                                                            48
                                                                  16 Grad~ Ductal
## 2 Endothelial c~ 3533
                                11.4
                                            9 TAN
                                                            48
                                                                  16 Grad~ Ductal
## 3 Endothelial c~ 3536
                                10.0
                                                                  52 Grad~ Lobular
                                           10 CAF
                                                            50
## 4 Endothelial c~ 3537
                                 9.13
                                           10 TAN
                                                            50
                                                                  52 Grad~ Lobular
## 5 Endothelial c~ 4027
                                16.6
                                            2 CAF
                                                            77
                                                                  40 Grad~ Ductal
## 6 Endothelial c~ 4028
                                11.9
                                                            77
                                                                  40 Grad~ Ductal
                                            2 TAN
## 7 Endothelial c~ 4033
                                            1 CAF
                                                                  45 Grad~ Lobular
                                11.6
                                                            46
## 8 Endothelial c~ 4034
                                12.8
                                            1 TAN
                                                            46
                                                                  45 Grad~ Lobular
                                                                 12 Grad~ Ductal
## 9 Endothelial c~ 4112
                                10.9
                                            3 CAF
                                                            62
## 10 Endothelial c~ 4113
                                                                  8 Grad~ Ductal
                                11.2
                                            3 TAN
                                                            62
## # i 38 more rows
## # i 5 more variables: ER <chr>, PR <chr>, Her2 <chr>, Subtype <chr>, LVI <chr>
deconvolution_compare_fibroblast_score_caf_tan <- function(df, score_column) {</pre>
  df_caf <- df %>% dplyr::filter(cell_type == "Cancer associated fibroblast")
  caf_score <- df_caf[[score_column]][which(df_caf$Condition == "CAF")]</pre>
  tan_score <- df_caf[[score_column]][which(df_caf$Condition == "TAN")]</pre>
  print("carrying out correlation test...")
  print(cor.test(x = caf_score, y = tan_score,
         method = c("pearson"),
         conf.level = 0.95))
  plt_title <- paste(score_column, " cancer-associated fibroblast for inhouse CAF vs TAN")
  par(pty = "s")
  #boxplot(score_column ~ Condition, data = df_caf)
  #plot_out <- boxplot(df_caf[,score_column] ~ df_caf[,Condition],</pre>
        col = c("#003C67FF", "#EFC000FF"),
        main = plt_title,
       xlab = "Condition", ylab = score_column)
  print(df_caf)
  plot_out <- ggplot(data = df_caf, aes(Condition, .data[[score_column]])) +</pre>
              geom_boxplot(outlier.colour="red", outlier.shape=8, outlier.size=4)
  bartlet_out <- bartlett.test(df_caf[[score_column]] ~ df_caf$Condition, data = df_caf)</pre>
  if (bartlet_out$p.value > 0.05){
   print("variances equal, carry out t-test")
   t.test.out <- t.test(formula = df_caf[[score_column]] ~ df_caf$Condition,</pre>
       alternative = "two.sided",
       mu = 0,
       paired = TRUE,
       var.equal = TRUE,
```

```
conf.level = 0.95, data = df_caf)
 }
 outputs <- list(plot_out, bartlet_out, t.test.out)</pre>
 return(outputs)
out_mcp <- deconvolution_compare_fibroblast_score_caf_tan(deconvolution_mcp_filtered_long_metadata, "mc
## [1] "carrying out correlation test..."
## Pearson's product-moment correlation
##
## data: caf_score and tan_score
## t = 2.5875, df = 10, p-value = 0.02707
\#\# alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.09324848 0.88538684
## sample estimates:
        cor
## 0.6332609
##
## # A tibble: 24 x 14
                    sample mcp_score Patient Condition Age Size Grade Histology
     cell_type
##
     <chr>>
                    <chr>
                               <dbl> <int> <chr>
                                                      <int> <int> <chr> <chr>
                                           9 CAF
                                                                16 Grad~ Ductal
## 1 Cancer associ~ 3532
                               2118.
                                                          48
## 2 Cancer associ~ 3533
                               2568.
                                          9 TAN
                                                          48
                                                               16 Grad~ Ductal
## 3 Cancer associ~ 3536
                                         10 CAF
                                                          50 52 Grad~ Lobular
                               2675.
## 4 Cancer associ~ 3537
                               2843.
                                         10 TAN
                                                         50
                                                               52 Grad~ Lobular
## 5 Cancer associ~ 4027
                               2797.
                                          2 CAF
                                                          77
                                                               40 Grad~ Ductal
## 6 Cancer associ~ 4028
                                          2 TAN
                                                         77
                                                               40 Grad~ Ductal
                              2899.
## 7 Cancer associ~ 4033
                               2170.
                                          1 CAF
                                                          46
                                                               45 Grad~ Lobular
## 8 Cancer associ~ 4034
                               2883.
                                           1 TAN
                                                          46
                                                                45 Grad~ Lobular
## 9 Cancer associ~ 4112
                               3732.
                                           3 CAF
                                                          62
                                                                12 Grad~ Ductal
## 10 Cancer associ~ 4113
                                                                8 Grad~ Ductal
                               4141.
                                           3 TAN
                                                          62
## # i 14 more rows
## # i 5 more variables: ER <chr>, PR <chr>, Her2 <chr>, Subtype <chr>, LVI <chr>
## [1] "variances equal, carry out t-test"
```

out_mcp[[1]]



```
out_mcp[[2]]
```

```
##
##
   Bartlett test of homogeneity of variances
## data: df_caf[[score_column]] by df_caf$Condition
## Bartlett's K-squared = 0.00074082, df = 1, p-value = 0.9783
out_mcp[[3]]
##
##
   Paired t-test
## data: df_caf[[score_column]] by df_caf$Condition
## t = -2.2206, df = 11, p-value = 0.04831
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -575.333038
                  -2.553832
## sample estimates:
## mean difference
         -288.9434
##
```

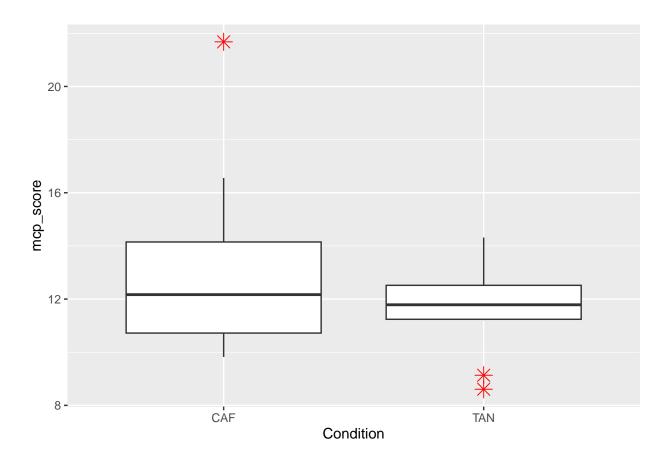
deconvolution_compare_celltype_score_caf_tan <- function(df, score_column, cell_type_in) {</pre>

df_cell_type <- df %>% dplyr::filter(cell_type == cell_type_in)

```
caf_score <- df_cell_type[[score_column]][which(df_cell_type$Condition == "CAF")]</pre>
  tan_score <- df_cell_type[[score_column]][which(df_cell_type$Condition == "TAN")]</pre>
  print("carrying out correlation test...")
  print(cor.test(x = caf_score, y = tan_score,
         method = c("pearson"),
         conf.level = 0.95))
  plot_out <- ggplot(data = df_cell_type, aes(Condition, .data[[score_column]])) +</pre>
              geom boxplot(outlier.colour="red", outlier.shape=8, outlier.size=4)
  bartlet_out <- bartlett.test(df_cell_type[[score_column]] ~ df_cell_type$Condition, data = df_cell_ty
  print(bartlet_out)
  if (bartlet_out$p.value > 0.05) {
    print("variances equal, carry out t-test equal variance")
    t.test.out <- t.test(formula = df_cell_type[[score_column]] ~ df_cell_type$Condition,</pre>
       alternative = "two.sided",
       mu = 0,
       paired = TRUE,
       var.equal = TRUE,
       conf.level = 0.95, data = df_cell_type)
      print("variances not equal, carry out t-test unequal variance")
      t.test.out <- t.test(formula = df_cell_type[[score_column]] ~ df_cell_type$Condition,</pre>
      alternative = "two.sided",
      mu = 0,
      paired = TRUE,
       var.equal = FALSE,
       conf.level = 0.95, data = df_cell_type)
    }
  outputs <- list(plot_out, bartlet_out, t.test.out)</pre>
  return(outputs)
deconvolution_compare_celltype_score_caf_tan(deconvolution_mcp_filtered_long_metadata, "mcp_score", "En
## [1] "carrying out correlation test..."
##
## Pearson's product-moment correlation
##
## data: caf_score and tan_score
## t = 1.437, df = 10, p-value = 0.1813
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.2100771 0.7981123
## sample estimates:
##
         cor
## 0.4137019
##
##
## Bartlett test of homogeneity of variances
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 5.9468, df = 1, p-value = 0.01474
##
```

[1] "variances not equal, carry out t-test unequal variance"

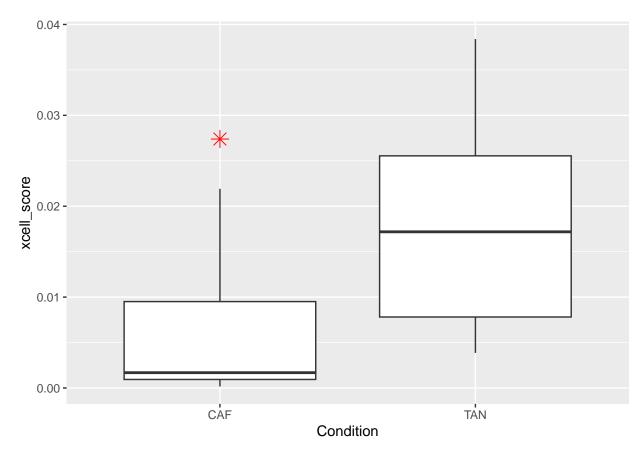
[[1]]



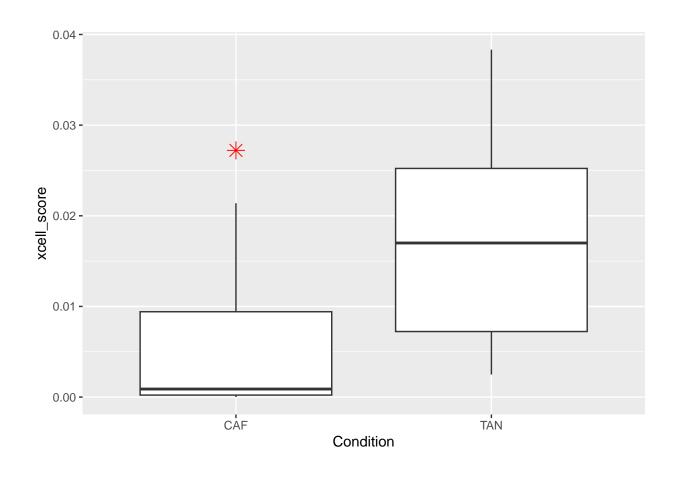
```
##
## [[2]]
##
   Bartlett test of homogeneity of variances
##
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 5.9468, df = 1, p-value = 0.01474
##
##
## [[3]]
##
##
   Paired t-test
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## t = 1.5153, df = 11, p-value = 0.1579
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.6135207 3.3251092
## sample estimates:
## mean difference
##
         1.355794
```

```
## [1] "carrying out correlation test..."
##
##
   Pearson's product-moment correlation
##
## data: caf_score and tan_score
## t = 3.2315, df = 10, p-value = 0.008998
\#\# alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.2387499 0.9138004
## sample estimates:
##
         cor
## 0.7147212
##
##
##
   Bartlett test of homogeneity of variances
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 0.15337, df = 1, p-value = 0.6953
## [1] "variances equal, carry out t-test equal variance"
```

[[1]]



```
##
## [[2]]
##
  Bartlett test of homogeneity of variances
##
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 0.15337, df = 1, p-value = 0.6953
##
##
## [[3]]
##
  Paired t-test
##
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## t = -4.7058, df = 11, p-value = 0.0006441
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.015933207 -0.005778311
## sample estimates:
## mean difference
##
      -0.01085576
deconvolution_compare_celltype_score_caf_tan(deconvolution_xcell_filtered_long_metadata, "xcell_score",
## [1] "carrying out correlation test..."
##
##
  Pearson's product-moment correlation
##
## data: caf_score and tan_score
## t = 3.018, df = 10, p-value = 0.01294
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1929609 0.9055186
## sample estimates:
##
         cor
## 0.6904065
##
##
## Bartlett test of homogeneity of variances
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 0.20576, df = 1, p-value = 0.6501
## [1] "variances equal, carry out t-test equal variance"
## [[1]]
```

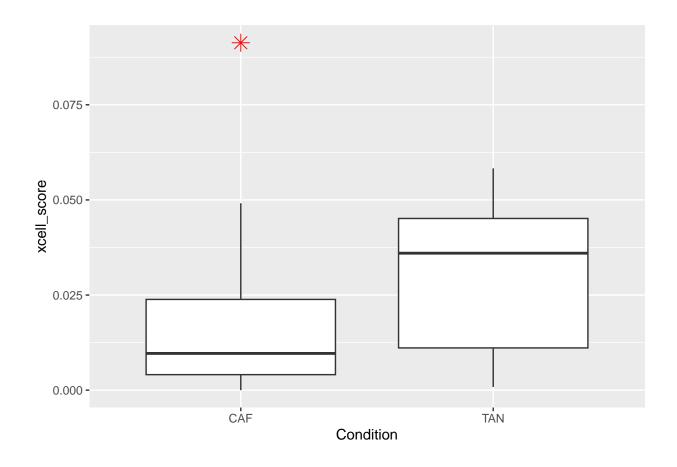


```
##
## [[2]]
##
##
   Bartlett test of homogeneity of variances
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 0.20576, df = 1, p-value = 0.6501
##
##
## [[3]]
##
##
   Paired t-test
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## t = -4.4415, df = 11, p-value = 0.0009926
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.016396379 -0.005530487
## sample estimates:
## mean difference
##
       -0.01096343
```

deconvolution_compare_celltype_score_caf_tan(deconvolution_xcell_filtered_long_metadata, "xcell_score",

[1] "carrying out correlation test..."

```
##
## Pearson's product-moment correlation
##
## data: caf_score and tan_score
## t = 0.66391, df = 10, p-value = 0.5218
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.4176870 0.6971612
## sample estimates:
##
         cor
## 0.2054675
##
##
## Bartlett test of homogeneity of variances
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## Bartlett's K-squared = 0.85096, df = 1, p-value = 0.3563
##
## [1] "variances equal, carry out t-test equal variance"
## [[1]]
```



[[2]]

```
Bartlett test of homogeneity of variances
##
## data: df cell type[[score column]] by df cell type$Condition
  Bartlett's K-squared = 0.85096, df = 1, p-value = 0.3563
##
##
## [[3]]
##
##
  Paired t-test
##
## data: df_cell_type[[score_column]] by df_cell_type$Condition
## t = -1.1102, df = 11, p-value = 0.2906
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
   -0.028307954 0.009325737
## sample estimates:
  mean difference
##
      -0.009491109
```

metadata

```
##
      sample Patient Condition Age Size
                                            Grade Histology
                                                                       ER
                                                                                   PR.
## 1
        4033
                            CAF
                                 46
                    1
                                       45 Grade_2
                                                    Lobular ER_positive PR_positive
## 2
        4034
                    1
                            TAN
                                 46
                                       45 Grade 2
                                                    Lobular ER positive PR positive
## 3
        4027
                    2
                            CAF
                                 77
                                       40 Grade_3
                                                     Ductal ER_negative PR_negative
                    2
## 4
        4028
                            TAN
                                 77
                                       40 Grade_3
                                                      Ductal ER_negative PR_negative
## 5
        4112
                    3
                            CAF
                                 62
                                       12 Grade_3
                                                     Ductal ER_positive PR_negative
## 6
        4113
                    3
                            TAN
                                 62
                                        8 Grade_3
                                                     Ductal ER_positive PR_negative
## 7
                    4
                            CAF
                                 45
                                       35 Grade_2
                                                    Lobular ER_positive PR_positive
        4116
## 8
        4117
                    4
                            TAN
                                 45
                                       13 Grade_2
                                                    Lobular ER_positive PR_positive
## 9
                    5
                            CAF
                                 78
        4214
                                       90 Grade_2
                                                    Lobular ER_positive PR_negative
## 10
                                       90 Grade_2
                    5
        4215
                            TAN
                                 78
                                                    Lobular ER_positive PR_negative
## 11
                                                     Ductal ER_positive PR_positive
                    6
                            CAF
                                 84
                                       30 Grade_2
        4315
## 12
                    6
                                       22 Grade 2
        4316
                            TAN
                                 84
                                                     Ductal ER_positive PR_positive
                    7
## 13
        4340
                            CAF
                                 62
                                      100 Grade_2
                                                    Lobular ER_positive PR_positive
## 14
        4341
                    7
                            TAN
                                 62
                                      100 Grade 2
                                                    Lobular ER_positive PR_positive
                    8
## 15
                            CAF
                                 50
        4344
                                       28 Grade 2
                                                     Ductal ER_positive PR_positive
## 16
        4345
                    8
                            TAN
                                 50
                                       28 Grade 2
                                                     Ductal ER_positive PR_positive
                    9
## 17
        3532
                            CAF
                                 48
                                       16 Grade 2
                                                      Ductal ER_positive PR_positive
## 18
        3533
                   9
                            TAN
                                 48
                                       16 Grade_2
                                                     Ductal ER_positive PR_positive
## 19
        3536
                   10
                            CAF
                                 50
                                       52 Grade_3
                                                    Lobular ER_positive PR_positive
## 20
        3537
                   10
                            TAN
                                 50
                                       52 Grade_3
                                                    Lobular ER_positive PR_positive
## 21
        4299
                   11
                            CAF
                                 84
                                       40 Grade_3
                                                     Ductal ER_positive PR_positive
## 22
        4300
                                 84
                   11
                            TAN
                                       40 Grade_3
                                                     Ductal ER_positive PR_positive
## 23
        4722
                   12
                            CAF
                                 81
                                       52 Grade_2
                                                     Lobular ER_positive PR_negative
## 24
        4723
                                       52 Grade_2
                   12
                            TAN
                                 81
                                                    Lobular ER_positive PR_negative
##
               Her2
                     Subtype
## 1
      Her2_negative LuminalA LVI_negative
      Her2_negative LuminalA LVI_negative
## 3
      Her2_negative
                         TNBC LVI_positive
## 4
      Her2_negative
                         TNBC LVI_positive
## 5
      Her2_negative LuminalA LVI_negative
      Her2_negative LuminalA LVI_negative
      Her2_negative LuminalA LVI_positive
```

```
## 8 Her2_negative LuminalA LVI_positive
## 9 Her2_negative LuminalA LVI_positive
## 10 Her2_negative LuminalA LVI_positive
## 11 Her2_negative LuminalA LVI_positive
## 12 Her2_negative LuminalA LVI_positive
## 13 Her2_negative LuminalA LVI_negative
## 14 Her2_negative LuminalA LVI_negative
## 15 Her2_negative LuminalA LVI_negative
## 16 Her2_negative LuminalA LVI_negative
## 17 Her2_negative LuminalA LVI_negative
## 18 Her2_negative LuminalA LVI_negative
## 19 Her2_negative LuminalA LVI_negative
## 20 Her2_negative LuminalA LVI_negative
## 21 Her2_negative LuminalA LVI_positive
## 22 Her2_negative LuminalA LVI_positive
## 23 Her2_negative LuminalA LVI_negative
## 24 Her2_negative LuminalA LVI_negative
salmon_markers <- data.frame(t(salmon_tpm_hgnc_not_duplicated[c("ACTA2","FAP"),]))</pre>
salmon_markers$sample <- rownames(salmon_markers)</pre>
metadata_markers <- full_join(metadata, salmon_markers)</pre>
## Joining with 'by = join_by(sample)'
# svg("/home/rstudio/Documents/PhD/notes/mini_viva_report/images/current_research/figure3/xcell_deconv_
# xcell_plot
# dev.off()
```

```
 \#svg("/home/rstudio/Documents/PhD/notes/mini\_viva\_report/images/current\_research/figure3/epic\_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca
```

```
 \# \ svg("/home/rstudio/Documents/PhD/notes/mini\_viva\_report/images/current\_research/figure3/mcp\_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_ca_deconv\_
```