

Figure 7

This notebook reproduces the main analyses from figure 7 of the manuscript. First we import the necessary packages.

```
library(Seurat)
library(tidyverse)
library(patchwork)
library(voxhunt)
```

Now we set up some colors and load the data. The loaded seurat object contains the bulk RNA-seq data from the patterning screen. Each sample represents one organoid. We also do some feature selection with VoxHunt.

```
data("voxel_meta")
load_aba_data('voxhunt_data/')
patscreen_data <- read_rds('patscreen_srt.rds')

ps_names <- c(
  'Ctrl_no_dose',
  'SHH_high', 'SHH_low',
  'CHIR_high', 'CHIR_low',
  'RSP02_high', 'RSP02_low',
  'RSP03_high', 'RSP03_low',
  'SB_DM_high', 'SB_DM_low'
)
ps_colors <- c(
  'gray',
  '#f57f17', '#ffc107',
  '#186a3b', '#28b463',
  '#c2185b', '#f48fb1',
  '#512da8', '#9575cd',
  '#2874a6', '#5dade2'
)
names(ps_colors) <- ps_names

struct_markers <- structure_markers('E13', 'custom_2')
genes_use <- struct_markers %>%
  group_by(group) %>%
  top_n(10, auc) %>%
  pull(gene) %>% unique()

print(patscreen_data)

## An object of class Seurat
## 98973 features across 61 samples within 3 assays
## Active assay: RNA (32991 features, 0 variable features)
## 2 other assays present: counts, TPM
```

Now we perform PCA to get an impression how the samples are distributed.

```
patscreen_data <- patscreen_data %>% FindVariableFeatures(nfeatures=2000) %>% ScaleData() %>% RunPCA(np
```

```
## Warning in FindVariableFeatures.Assay(object = assay.data, selection.method =  
## selection.method, : selection.method set to 'vst' but count slot is empty; will  
## use data slot instead
```

```
## Centering and scaling data matrix
```

```
## PC_ 1
```

```
## Positive: PTRF, LGALS3, S100A14, GSTA2, KYNU, KDELR3, TMEM37, GSTA1, APOE, SOAT2
```

```
## LCP1, MIR4435-2HG, VIL1, NPNT, ANXA4, GOS2, AKR1D1, HNF4A, IGF2, LGALS2
```

```
## ARSE, F2, AMBP, TMEM92, TF, FABP1, PLA2G12B, AGT, GGT1, SLC02B1
```

```
## Negative: TCEAL7, TMSB15A, H2AFZ, HMG2, SMC4, PLEKH01, FABP7, BTBD17, CENPV, GINS2
```

```
## FAM64A, TCEAL5, PCNA, BEX2, IMMP1L, HMGB2, PIK3R3, CCDC144NL-AS1, HIST1H4C, TUBA1B
```

```
## CENPF, H2AFV, HMGB1, MARCKSL1, C16orf45, RTN1, HN1, ACOT7, TMSB15B, NCALD
```

```
## PC_ 2
```

```
## Positive: FAM212A, FABP5, TNNC2, SIX1, SCRG1, RGS10, NPFFR2, CD163L1, ROPN1, S100A6
```

```
## RP11-45A12.1, ACTC1, S100A5, TFAP2B, MIA, GPHA2, FAM110A, PLAC9, GAS2, SOX10
```

```
## ISL1, PGF, SLC9A9, NRP2, LRRC53, ROPN1B, ERVMER61-1, DLX1, S100B, EGFL8
```

```
## Negative: CXADR, VAT1L, NUA2, POR, SDC2, COL4A6, PROM1, DDR1, ELP4, GDDP2
```

```
## GGA2, KLHDC8B, CLU, C5orf22, GJA1, ODC1, FGFR2, PALLD, ALPL, PAM
```

```
## PLAGL1, UBE3D, EPHA7, TUBB4A, CD82, MID1, TRPM3, TMEM2, TSC22D1, EFHD1
```

```
## PC_ 3
```

```
## Positive: CTNNA1, MTHFD2, PLS3, ACLY, HDAC2, PPP2R1B, VCP, DDX5, EFTUD2, CTPS1
```

```
## CCNC, ZNF385C, EYA1, UNC5C, CCT8, HNRNP, HEYL, XPOT, API5, SRSF1
```

```
## SF3B3, C5orf30, CAPRIN1, XRCC6, TRIM9, SRSF6, DLD, MAPRE1, FAR2, CSE1L
```

```
## Negative: RP4-616B8.4, RP6-65G23.5, RP11-1035H13.3, RP11-461A8.1, BARHL2, RP11-371E8.4, RP11-256P1.1
```

```
## C1orf194, RP11-2B6.3, C2orf80, PRLH, GRIK1, FETUB, RP11-316M1.12, ORM1, HLA-B, CORIN
```

```
## AP001615.9, HELT, MUC12, ENKUR, AC079742.4, CIB1, RP11-849I19.1, RP11-109P14.10, RP11-493L12.4, I
```

```
## PC_ 4
```

```
## Positive: VIT, TUBB2A, ALCAM, RSP02, WNT1, UNC5C, TNFRSF19, CDK6, CHGB, ASCL1
```

```
## PCDH9, C8orf4, PCSK1N, CTHRC1, ZIC4, RP11-77H9.2, ZIC1, GAP43, PAM, NHLH2
```

```
## SLC17A6, STMN2, WNT10B, RP11-363J20.2, SYT4, NDRG2, WFIKN1, WNT4, RSP03, RASD1
```

```
## Negative: RAX, SIX3, SHISA2, SIX6, RP11-89K21.1, LINC01551, FZD5, FAM69C, LHX2, DLK1
```

```
## F2RL2, DLX6, PPP2R2B, TNNT1, RP11-665C14.2, GPR160, ATP6V1B1, CAPN6, ADAM28, FEZF1-AS1
```

```
## RP11-114G22.1, EDN3, FEZF1, FGF8, SCUBE1, DANT1, CDA, BST2, PAMR1, CLDN10
```

```
## PC_ 5
```

```
## Positive: NOS2, NTN5, PAX3, LINC01198, ZNF436-AS1, COL11A2, MUSTN1, PAX6, LMX1A, GREM1
```

```
## ARHGAP15, CRABP2, RP11-96L14.7, NRIP3, TPPP3, RP11-314M24.1, AGR3, NEUROG1, SLC39A8, CST4
```

```
## ZIC2, ID3, LHX5-AS1, FOXD3-AS1, EGFL8, DKK1, WNT1, LGI1, CDH9, TAL2
```

```
## Negative: TECRL, RP11-162P23.2, RP11-568A7.4, PCAT14, GDF3, RP11-267L5.1, CH507-513H4.4, PLVAP, SLP
```

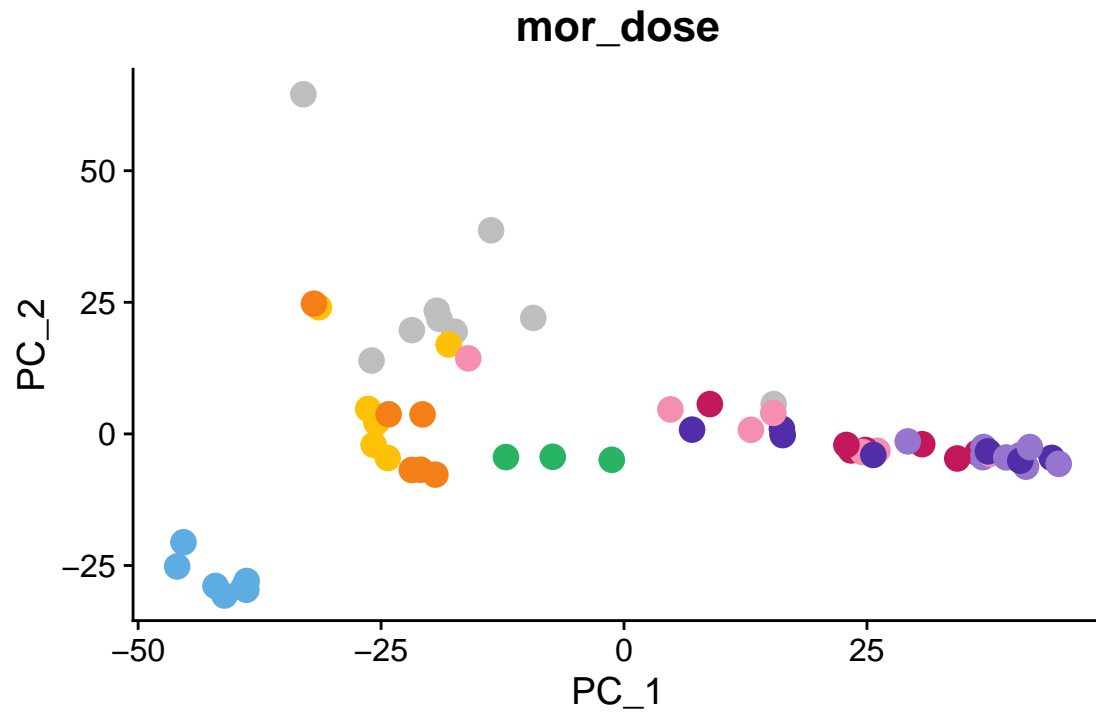
```
## PCOLCE, RGS1, CH507-513H4.6, TDGF1, PRRX1, RP11-85G21.3, SPINK4, HOXB6, LBH, CD300E
```

```
## HOXB-AS3, NANOG, CH507-513H4.3, ECSCR, SLC7A3, CH507-513H4.5, ICAM2, MYH10, POU5F1, ALDH2
```

```
patscreen_data@reductions$pca@stdev
```

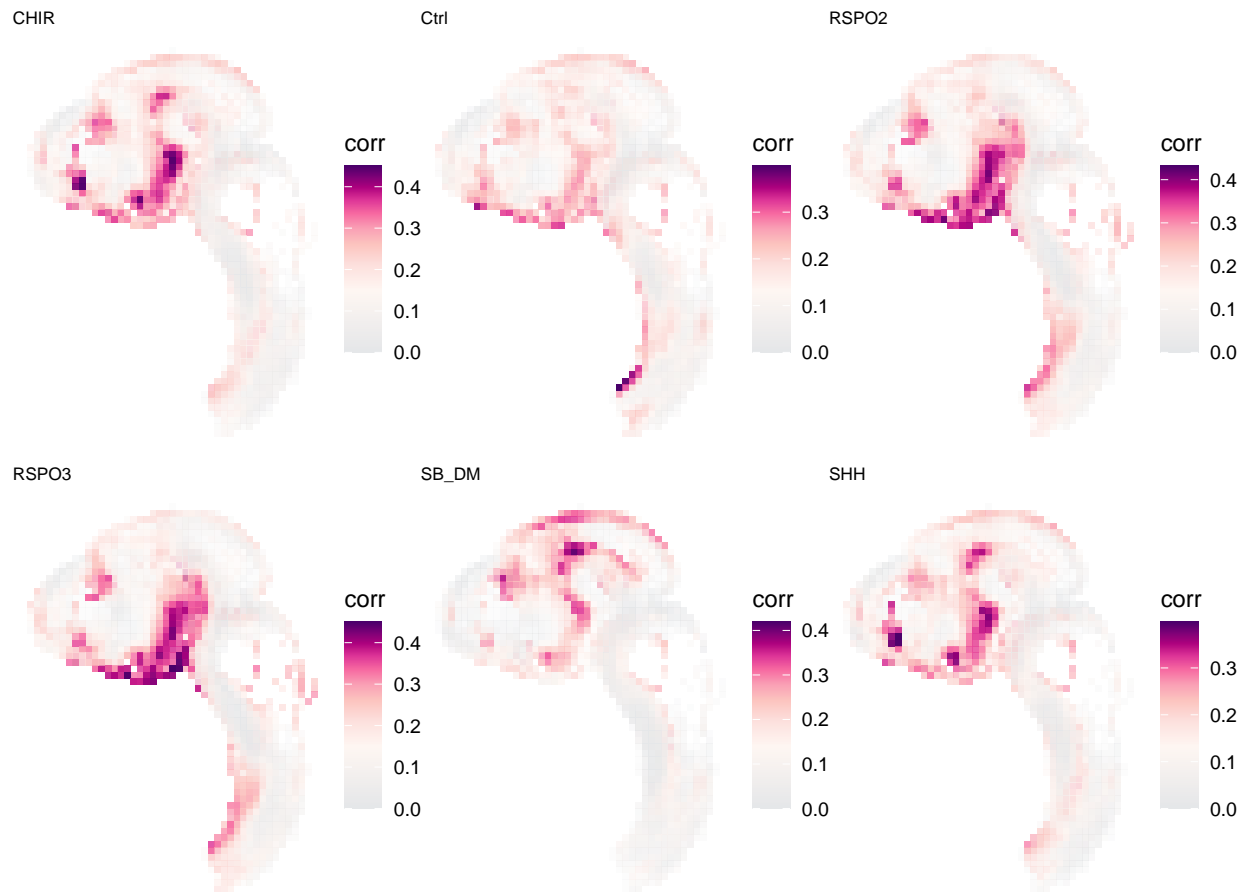
```
## [1] 29.44264 16.70063 12.20770 8.94130 7.89276
```

```
DimPlot(patscreen_data, group.by='mor_dose', pt.size=4) +
  scale_color_manual(values=ps_colors) +
  no_legend()
```



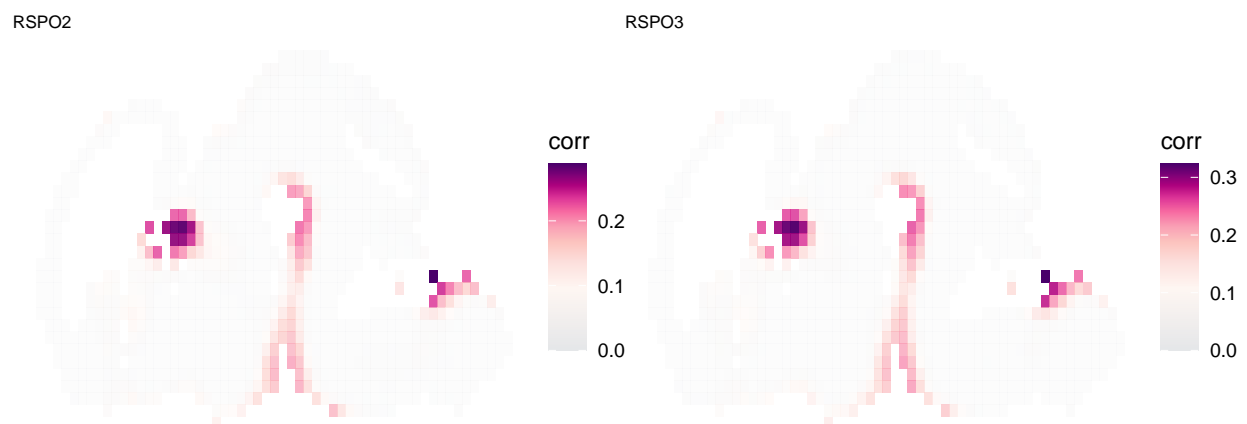
Now we map the data with VoxHunt.

```
ps_map <- voxel_map(patscreen_data, stage='E11', genes_use = genes_use, group_name = 'morphogen')
plot_map(ps_map)
```



Further, we'll have a closer look at the correlation patterns of RSPO2/3.

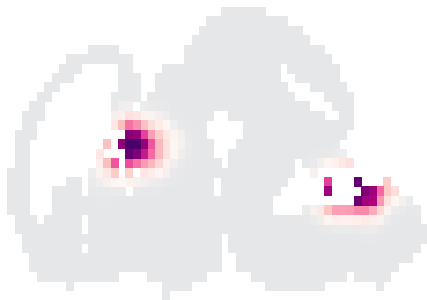
```
rspo_data <- subset(patscreen_data, morphogen%in%c('RSPO2', 'RSPO3'))
ps_map <- voxel_map(rspo_data, stage='E13', genes_use = genes_use, group_name = 'morphogen')
plot_map(ps_map, slices = 6)
```



Lastly, we can look at some markers on the same slice to see that RSPO2 & 3 are expressed adjacent to the highlighted position.

```
plot_expression('E13', slices = 6, genes = c('RSP02', 'RSP03', 'TTR', 'SPINT2'))
```

TTR



RSP02



SPINT2



RSP03

