

# Automatic generation of paper figures

***Giorgia Guglielmi, Joseph D. Barry, Wolfgang Huber, Stefano De Renzis***

## Contents

1	Introduction	1
2	Load Data	1
3	Analysis	2
3.1	WT	2
3.2	Global	4
3.3	2p titration	8
3.4	Sktl, Zip, Dark	9
3.5	Boxes Far vs Boxes Close	11
3.6	Global 2p	13

## 1 Introduction

---

In this vignette we automatically generate the data analysis figures produced for the paper by Guglielmi *et al.*

## 2 Load Data

---

```
data("opto")
library(dplyr)
library(ggplot2)
```

## 3 Analysis

### 3.1 WT

To identify the furrowing line in the control (WT) samples, we first chose by inspection a time point for each sample that showed a clear band of constriction. An average cell area profile was computed along the dorsal-ventral (d-v) axis, to which a smoothing line was fitted. We defined the position of minimum area as the furrowing position. Here we compute this for all of the control samples but display the results for only one sample.

```
tabWT <- filter(opto, condition == "WT")
myCex <- 1.4
tabWT[tabWT$sample == "Global Activation/1_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5"] >#
tabWT[tabWT$sample == "Global Activation/2_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5"] >#
tabWT[tabWT$sample == "Global Activation/3_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5"] >#
tabWT[tabWT$sample == "Global Activation/4_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5"] >#
tabWT[tabWT$sample == "Global Activation/5_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5"] >#
```

```
plotSamples <- "Global Activation/4_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5.rda"
fnPos <- function(x, nbinsExclude=150, myCex=1.4, plotSamples) {
  doPlot <- x$sample[1] %in% plotSamples
  pos <- identifyFurrowPosition(x, nbinsExclude=nbinsExclude, h=100,
    plot=doPlot, myCex=myCex, px=x$px[1])
  return(data.frame(furrowPos=pos))
}
dvFurrowPos <- tabWT %>%
  group_by(sample) %>%
  filter(t == tstar) %>%
  do(fnPos(., plotSamples=""))

## Warning: `as.tbl()` is deprecated as of dplyr 1.0.0.
## Please use `tibble::as_tibble()` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_warnings()` to see where this warning was generated.

for (i in seq_along(dvFurrowPos$sample))
  tabWT[tabWT$sample %in% dvFurrowPos$sample[i], "furrowPos"] <-
    dvFurrowPos$furrowPos[i]
print(dvFurrowPos)

## # A tibble: 5 x 2
## # Groups:   sample [5]
##   sample          furrowPos
##   <chr>           <int>
## 1 Global Activation/1_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5      301
## 2 Global Activation/2_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5      258
## 3 Global Activation/3_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5      280
## 4 Global Activation/4_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5      207
## 5 Global Activation/5_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5      266
```

## Automatic generation of paper figures

We wished to focus only on cells along or near to the furrowing line. Therefore we subsetted the control data to include only cells located in a box of dimensions  $30 \times 10$  microns that was centered on the furrowing line.

```
L <- function(x, px) round(microns2px(x, px))
fnBox <- function(x, w=512, Lx=30, Ly=10) {
  Lx <- L(Lx, px=x$px[1])
  Ly <- L(Ly, px=x$px[1])
  box <- constructBox(x$furrowPos[1], Lx=Lx, Ly=Ly, w=w)
  return(data.frame(t(box)))
}
boxes <- tabWT %>% group_by(sample) %>% do(fnBox(.))

fnIsolate <- function(x, boxes) {
  box <- unlist(filter(boxes, sample == x$sample[1])[2:5])
  x <- isolateBoxCells(x, box)
  return(x)
}
tabWTs <- tabWT %>% group_by(sample) %>% do(fnIsolate(., boxes=boxes))
```

We then proceeded to also automatically identify the time point of tissue invagination. We noticed that as the tissue furrowed, adjacent cells moved into the observation box from both dorsal and ventral sides, and tended to be larger in area than the cells that initiated the constriction. By fitting a smoothing line to the time profile of median cell area in the observation box it was straightforward to identify the time at which a minimum area was reached, which we chose to define as the time point of tissue invagination.

```
fnMinArea <- function(x, plot=FALSE) {
  px <- x$px[1]
  res <- identifyTimeMinArea(x=x, px=px, plot=plot)
  return(data.frame(t=res["tindex"]))
}
tMinArea <- tabWTs %>% group_by(sample) %>% do(fnMinArea(.))
for (i in seq_along(tMinArea$sample))
  tabWTs[tMinArea$sample == tMinArea$sample[i], "tstar"] <- tMinArea$t[i]
tabWTs <- filter(tabWTs, t <= tstar)
```

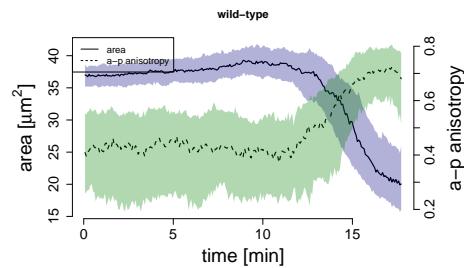
Next we inspected the dynamics of cell area and anisotropy.

```
plotSample <- "Global_Activation/4_CTRL_VFF.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closingSize5.rda"
tabSample <- filter(tabWTs, sample == plotSample)
tabSample$t <- tabSample$t*tabSample$dt
plotFeatureEvolution(tabSample, dt=1/60, tMax=Inf, myTitle="wild-type",
  px=tabSample$px[1], cex=2, cex.axis=1.5, line=3.3,
  mar=c(5.1, 5.6, 4.1, 4.6))
```

A summary of WT cell features at the time point of furrowing was then taken.

```
summaryWT <- tabWTs %>%
  group_by(sample) %>%
  filter(t == tstar) %>%
  summarize(APanisotropy=median(e.x, na.rm=TRUE),
    APanisotropy25=quantile(e.x, probs=0.25, na.rm=TRUE),
```

## Automatic generation of paper figures



**Figure 1: Fig 2J: Cell area and a-p anisotropy are plotted over time for cells within an area of 30 microns x 10 microns along the furrowing line for one control sample**

Solid and dashed lines indicate the median over cells for area and a-p anisotropy, respectively, while shaded regions show the interquartile range.

```
APanisotropy75=quantile(e.x, probs=0.75, na.rm=TRUE),
area=median(x.0.s.area, na.rm=TRUE),
area25=quantile(x.0.s.area, probs=0.25, na.rm=TRUE),
area75=quantile(x.0.s.area, probs=0.75, na.rm=TRUE))

## `summarise()` ungrouping output (override with `.groups` argument)

summaryWTapicalSlice <- tabWTs %>%
  group_by(sample) %>%
  filter(t == tstar, z == 1) %>%
  summarise(APanisotropy=median(e.x, na.rm=TRUE),
            APanisotropy25=quantile(e.x, probs=0.25, na.rm=TRUE),
            APanisotropy75=quantile(e.x, probs=0.75, na.rm=TRUE),
            area=median(x.0.s.area, na.rm=TRUE),
            area25=quantile(x.0.s.area, probs=0.25, na.rm=TRUE),
            area75=quantile(x.0.s.area, probs=0.75, na.rm=TRUE))

## `summarise()` ungrouping output (override with `.groups` argument)
```

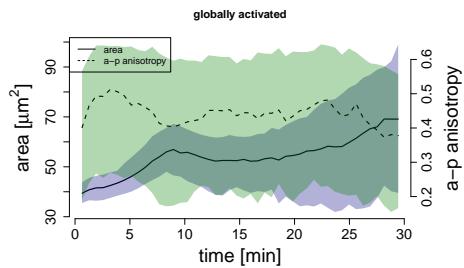
## 3.2 Global

We first inspected the dynamics of samples that were globally activated.

```
tabGlobal <- filter(opto, condition == "Global")
tabGlobal$t <- tabGlobal$t*tabGlobal$dt
plotSample <- "Global Activation/Global Activation/Image31.lsm - C=0/furrowSeg1.0.47_threshOffset5e-04_closing"
tabSample <- filter(tabGlobal, sample == plotSample)
plotFeatureEvolution(tabSample, dt=1/60, tMax=30, myTitle="globally activated",
                     px=tabSample$px[1], cex=2, cex.axis=1.5, line=3.3,
                     mar=c(5.1, 5.6, 4.1, 4.6))
```

We then asked if the cell behaviour of the control and photo-activated samples were quantitatively different. Since there was no clear way to choose equivalent time points in the control and photo-activated samples, we decided to compare control cells at the time of invagination to photo-activated cells at 10, 20 and 30 minutes post-activation.

## Automatic generation of paper figures



**Figure 2: Fig 2K: Cell area and a-p anisotropy are plotted over time for all cells in a globally light-activated sample**

Solid and dashed lines indicate the median over cells for area and a-p anisotropy, respectively, while shaded regions show the interquartile range.

```
sampleCenter <- tabGlobal %>% group_by(sample) %>% summarize(pos=mean(x.0.m.cy))

## `summarise()` ungrouping output (override with `groups` argument)

for (i in seq_len(nrow(sampleCenter)))
  tabGlobal[tabGlobal$sample == sampleCenter$sample[i], "furrowPos"] <-
    sampleCenter$pos[i]
subsetTimes <- function(x, tstar) {
  times <- unique(x$t)
  ind <- which.min(abs(tstar-times/60))
  xs <- filter(x, t == times[ind])
  return(xs)
}
compareTimes <- c(10, 20, 30)
tabGlobals <- lapply(compareTimes, function(tstar) {
  y <- tabGlobal %>% group_by(sample) %>% do(subsetTimes(., tstar=tstar))
  y$tstar <- tstar
  return(y)
})
tabGlobals <- bind_rows(tabGlobals)
tabGlobals$condition <- paste0("Global", tabGlobals$tstar)
compareGlobal <- tabGlobals %>% group_by(condition, sample) %>%
  summarise(APanisotropy=median(e.x, na.rm=TRUE),
           APanisotropy25=quantile(e.x, probs=0.25, na.rm=TRUE),
           APanisotropy75=quantile(e.x, probs=0.75, na.rm=TRUE),
           area=median(x.0.s.area, na.rm=TRUE),
           area25=quantile(x.0.s.area, probs=0.25, na.rm=TRUE),
           area75=quantile(x.0.s.area, probs=0.75, na.rm=TRUE))

## `summarise()` regrouping output by `condition` (override with `groups` argument)
summaryWT$condition <- "control"
compareGlobal <- bind_rows(compareGlobal, summaryWT)
```

To assess differences we then performed two-sample t-tests on the median cell area and a-p anisotropy for each sample. Since we compared between multiple groups, multiple testing correction was performed using the method of Bonferroni.

## Automatic generation of paper figures

```
wtArea <- compareGlobal %>% filter(condition == "control") %>%
  summarize(meanArea=mean(area))

## 'summarise()' ungrouping output (override with '.groups' argument)

wtArea <- wtArea$meanArea
compareGlobal$normArea <- log2(compareGlobal$area/wtArea)
compareGlobal$normArea25 <- log2(compareGlobal$area25/wtArea)
compareGlobal$normArea75 <- log2(compareGlobal$area75/wtArea)
ttAPanisotropy <- pairwise.t.test(compareGlobal$APanisotropy,
  compareGlobal$condition, p.adjust.method="bonferroni")
print(ttAPanisotropy)

##
## Pairwise comparisons using t tests with pooled SD
##
## data: compareGlobal$APanisotropy and compareGlobal$condition
##
##          Global10 Global20 Global30
## Global20 1.00000 -      -
## Global30 1.00000 1.00000 -
## control   0.00016  0.00075  0.00166
##
## P value adjustment method: bonferroni

ttArea <- pairwise.t.test(compareGlobal$normArea, compareGlobal$condition,
  p.adjust.method="bonferroni")
print(ttArea)

##
## Pairwise comparisons using t tests with pooled SD
##
## data: compareGlobal$normArea and compareGlobal$condition
##
##          Global10 Global20 Global30
## Global20 1      -      -
## Global30 1      1      -
## control   9.3e-07 8.8e-07 1.2e-07
##
## P value adjustment method: bonferroni
```

For visualization purposes, p-values were converted to stars indicating significance levels. For the comparison of median cell areas, the testing was performed on log2 transformed values to ensure a more constant variance between the conditions. To ensure that we were log transforming a dimensionless quantity, we first divided each measurement by the mean of the areas in the control group. We also decided to show the corresponding absolute area measurements on the right-hand axis.

```
mySignIf <- function(pv) symnum(pv, cutpoints=c(0, 0.001, 0.01, 0.05, 1),
  symbols=c("***", "**", "*", "n.s."))
myCol <- c("red", rep("blue", 3))
compareGlobal$condition <- factor(compareGlobal$condition,
  levels=c("control", "Global10", "Global20", "Global30"))
myNames <- levels(compareGlobal$condition)
```

## Automatic generation of paper figures

```
myStarsAPanisotropy <- apply(ttAPanisotropy$p.value, c(1, 2), mySignIf)
info <- sessionInfo()
par(mar=c(7.1, 6.1, 3.1, 5.6))
stripchart(APanisotropy ~ condition, data=compareGlobal, vertical=TRUE, pch=20,
           xlab="", ylab="a-p anisotropy", cex.lab=2, ylim=c(0, 1.1), col=myCol,
           axes=FALSE, method="jitter", jitter=0.05, cex=2)
axis(side=1, at=1:4, labels=FALSE)
text(x=1, par("usr")[3]-0.17, labels="control", srt=45, xpd=TRUE, cex=2)
text(x=2:4, par("usr")[3]-0.12, labels=rep("light", 3), srt=45, xpd=TRUE, cex=2)
text(x=2:4+0.05, par("usr")[3]-0.2, labels=c("+10min", "+20min", "+30min"),
      srt=45, pos=NULL, xpd=TRUE, cex=2)
axis(side=2, cex.axis=1.5, lwd=1.5)
drawBar <- function(x0, x1, y, star, offset=0.1, cex=2) {
  segments(x0=x0, y0=y, x1=x1, y1=y)
  text(x=mean(c(x0, x1)), y=y+offset, labels=star, cex=cex)
}
drawBar(1, 2, y=0.85, star=myStarsAPanisotropy["control", "Global10"],
        offset=0.05)
drawBar(1, 3, y=0.95, star=myStarsAPanisotropy["control", "Global20"],
        offset=0.05)
drawBar(1, 4, y=1.05, star=myStarsAPanisotropy["control", "Global30"],
        offset=0.05)
drawBar(2, 3, y=0.23, star=myStarsAPanisotropy["Global20", "Global10"],
        offset=-0.05)
drawBar(2, 4, y=0.65, star=myStarsAPanisotropy["Global30", "Global10"],
        offset=0.05)
drawBar(3, 4, y=0.13, star=myStarsAPanisotropy["Global30", "Global20"],
        offset=-0.05)
conditionSummary <- compareGlobal %>% group_by(condition) %>%
  summarize(median=median(APanisotropy),
            q25=quantile(APanisotropy, probs=0.25),
            q75=quantile(APanisotropy, probs=0.75))

## 'summarise()' ungrouping output (override with '.groups' argument)

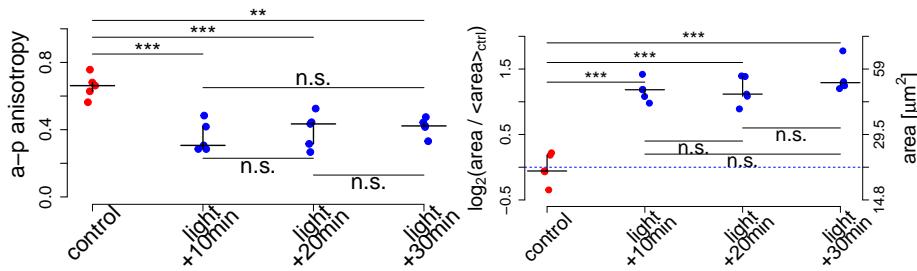
conditionSummary <- as.data.frame(conditionSummary)
conditionSummary <- conditionSummary[order(conditionSummary$condition), ]
for (i in seq_len(nrow(conditionSummary))) {
  segments(x0=i-0.2, x1=i+0.2, y0=conditionSummary[i, "median"], lwd=2, lty=1)
  segments(x0=i, y0=conditionSummary[i, "q25"], y1=conditionSummary[i, "q75"],
            lwd=2, lty=1)
}
myStarsArea <- apply(ttArea$p.value, c(1, 2), mySignIf)
par(mar=c(7.1, 6.1, 3.1, 5.6))
stripchart(normArea ~ condition, data=compareGlobal, vertical=TRUE, pch=20,
           xlab="", ylab=expression(paste("log"[2], "(area / <area>"[ctrl], ")")),
           cex.lab=2, group.names=myNames, ylim=c(-0.5, 2.0), col=myCol,
           axes=FALSE, method="jitter", jitter=0.05, cex=2)
axis(side=1, at=1:4, labels=FALSE)
text(x=1, par("usr")[3]-0.47, labels="control", srt=45, xpd=TRUE, cex=2)
text(x=2:4, par("usr")[3]-0.32, labels=rep("light", 3), srt=45, xpd=TRUE, cex=2)
text(x=2:4+0.05, par("usr")[3]-0.53, labels=c("+10min", "+20min", "+30min"),
```

## Automatic generation of paper figures

```

srt=45, pos=NULL, xpd=TRUE, cex=2)
axis(side=2, cex.axis=1.5, lwd=1.5)
drawBar(1, 2, y=1.3, star=myStarsArea["control", "Global10"], offset=0.12)
drawBar(1, 3, y=1.6, star=myStarsArea["control", "Global20"], offset=0.12)
drawBar(1, 4, y=1.9, star=myStarsArea["control", "Global30"], offset=0.12)
drawBar(2, 3, y=0.40, star=myStarsArea["Global20", "Global10"], offset=-0.12)
drawBar(2, 4, y=0.20, star=myStarsArea["Global30", "Global10"], offset=-0.12)
drawBar(3, 4, y=0.60, star=myStarsArea["Global30", "Global20"], offset=-0.12)
mapMicronsSq <- function(x) px2area(wtArea*2**x, px=0.293)
rangeNormAreaSeq <- seq(-0.5, 2.0, by=0.5)
newAxisLabels <- round(mapMicronsSq(rangeNormAreaSeq), digits=1)
axis(4, at=rangeNormAreaSeq, labels=newAxisLabels, cex.axis=1.5, lwd=1.5)
abline(h=0, lty=2, col="blue")
mtext(expression(paste("area [", mu, m^2, "]")), side=4, cex=2, line=4)
conditionSummary <- compareGlobal %>% group_by(condition) %>%
  summarize(median=median(normArea), q25=quantile(normArea, probs=0.25),
            q75=quantile(normArea, probs=0.75))
## 'summarise()' ungrouping output (override with '.groups' argument)
conditionSummary <- as.data.frame(conditionSummary)
conditionSummary <- conditionSummary[order(conditionSummary$condition), ]
for (i in seq_len(nrow(conditionSummary))) {
  segments(x0=i-0.2, x1=i+0.2, y0=conditionSummary[i, "median"], lwd=2, lty=1)
  segments(x0=i, y0=conditionSummary[i, "q25"], y1=conditionSummary[i, "q75"],
            lwd=2, lty=1)
}

```



**Figure 3: Fig 2L-M:** The median cell area and a-p anisotropy features at the time point of tissue invagination in the control samples were compared to three different time points in the photo-activated samples

The statistical testing was performed using pairwise two-sample t-tests. As can be seen from the figure, standard deviations were comparable between groups. To attain a higher precision we therefore used a pooled standard deviation for all of the tests. p-values were corrected for multiple testing using the method of Bonferroni. We performed the area testing on the log-transformed values but included the absolute areas on the right-hand y-axis as a reference. For both cell area and a-p anisotropy, the control samples were significantly different from the photo-activated samples, while the latter showed no significant changes over time. The crosses show group median (horizontal line) and interquartile range (vertical line).

### 3.3 2p titration

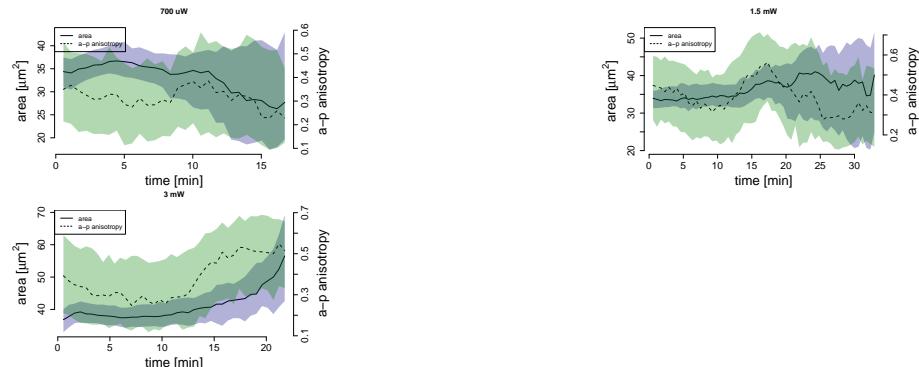
To assess the effect of increasing laser power on cell behaviour we extracted cell features from the following samples.

## Automatic generation of paper figures

```

titrationCond <- paste0("2pTitration", c("2.5", "5", "8"), "percent")
tab2pTitration <- filter(opto, condition %in% titrationCond)
tab2pTitration <- filter(tab2pTitration, !(sample == "New data/2p titration/2.5%/1_CIBN0CRL_R001_GR1_B1_SUM"))
tab2pTitration <- filter(tab2pTitration, !(sample == "New data/2p titration/2.5%/2_CIBN0CRL_R001_GR1_B1_SUM"))
tab2pTitration <- filter(tab2pTitration, !(sample == "New data/2p titration/2.5%/3_CIBN0CRL_R001_GR1_B1_SUM"))
tab2pTitration$t <- tab2pTitration$t*tab2pTitration$dt
plotSamples <- c(
  "700 uW"="New data/2p titration/2.5%/2_CIBN0CRL_R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closing",
  "1.5 mW"="New data/2p titration/5%/1_CIBN0CRL_R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closing",
  "3 mW"="New data/2p titration/8%/4_CIBN0CRL_R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closing"
)
for (i in seq_along(plotSamples)) {
  s <- plotSamples[i]
  tab2pTitrations <- filter(tab2pTitration, sample == s)
  px <- tab2pTitrations$px[1]
  plotFeatureEvolution(tab2pTitrations, dt=1/60, tMax=Inf, myTitle=names(s),
    px=px, cex=2, cex.axis=1.5, line=3.3, mar=c(5.1, 5.6, 4.1, 4.6))
}

```



**Figure 4: Fig 5M-O: Cell area and a-p anisotropy are plotted over time for locally light-activated cells in individual samples**

Different two-photon laser powers were used to monitor changes in cell behaviour resulting from varying levels of activation. Solid and dashed lines indicate the median over cells for area and a-p anisotropy, respectively, while shaded regions show the interquartile range.

## 3.4 Sktl, Zip, Dark

Next we tested if a-p anisotropy from various control samples was different from wild-type samples at the time point of furrowing. The most apical z-stack that gave an accurate cell segmentation was kept for the following analysis. A box isolating a group of constricting cells was chosen manually by the biologist.

```

tab <- filter(opto, condition %in% c("Dark", "Sktl", "Zip"))
summaryCnds <- tab %>% group_by(condition, sample) %>%
  summarize(APanisotropy=median(e.x, na.rm=TRUE),
            APanisotropy25=quantile(e.x, probs=0.25, na.rm=TRUE),
            APanisotropy75=quantile(e.x, probs=0.75, na.rm=TRUE))

```

## Automatic generation of paper figures

```
## 'summarise()' regrouping output by 'condition' (override with '.groups' argument)

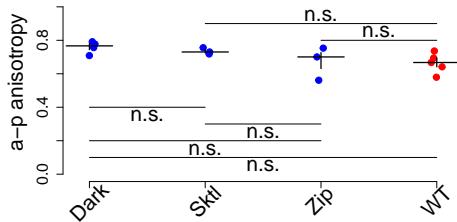
summaryWTapicalSlice$condition <- "WT"
summaryConds <- bind_rows(summaryConds, summaryWTapicalSlice)
myCol <- c(rep("blue", 3), "red")
summaryConds$condition <- factor(summaryConds$condition,
    levels=c("Dark", "Sktl", "Zip", "WT"))
summaryConds$sample <- factor(summaryConds$sample)
sampleOrdering <- levels(summaryConds$sample)[c(6:15, 1:5)]
summaryConds$sample <- factor(summaryConds$sample, levels=sampleOrdering)
anisotropyLab <- "a-p anisotropy"
par(mar=c(6.1, 6.1, 3.1, 2.1))
stripchart(APanisotropy ~ condition, data=summaryConds, vertical=TRUE, pch=20,
    xlab="", ylab=anisotropyLab, cex.lab=2, axes=FALSE, col=myCol,
    ylim=c(0, 1.1), method="jitter", jitter=0.05, cex=2)
axis(side=1, at=1:4, labels=FALSE, cex.axis=1.5, lwd=1.5)
axis(side=2, cex.axis=1.5, lwd=1.5)
text(x=1:4, par("usr")[3]-0.12, labels=levels(summaryConds$condition), srt=45,
    xpd=TRUE, cex=2)
conditionSummary <- summaryConds %>% group_by(condition) %>%
    summarize(median=median(APanisotropy),
        q25=quantile(APanisotropy, probs=0.25),
        q75=quantile(APanisotropy, probs=0.75))

## 'summarise()' ungrouping output (override with '.groups' argument)

conditionSummary <- as.data.frame(conditionSummary)
for (i in seq_len(nrow(conditionSummary))) {
    segments(x0=i-0.2, x1=i+0.2, y0=conditionSummary[i, "median"], lwd=2, lty=1)
    segments(x0=i, y0=conditionSummary[i, "q25"], y1=conditionSummary[i, "q75"],
        lwd=2, lty=1)
}
ttConds <- pairwise.t.test(summaryConds$APanisotropy, summaryConds$condition,
    p.adjust.method="bonferroni")
myStarsConds <- apply(ttConds$p.value, c(1, 2), mySignIf)
print(myStarsConds)

##      Dark   Sktl   Zip
## Sktl "n.s." "?" "?"
## Zip  "n.s." "n.s." "?"
## WT   "n.s." "n.s." "n.s."

drawBar(1, 2, y=0.40, star=myStarsConds["Sktl", "Dark"], offset=-0.05)
drawBar(2, 3, y=0.30, star=myStarsConds["Zip", "Sktl"], offset=-0.05)
drawBar(1, 3, y=0.20, star=myStarsConds["Zip", "Dark"], offset=-0.05)
drawBar(1, 4, y=0.10, star=myStarsConds["WT", "Dark"], offset=-0.05)
drawBar(2, 4, y=0.90, star=myStarsConds["WT", "Sktl"], offset=0.05)
drawBar(3, 4, y=0.80, star=myStarsConds["WT", "Zip"], offset=0.05)
```



**Figure 5: Fig 3A: The median a-p anisotropy of constricting cells in different condition are compared to the a-p anisotropy of control cells at the time point of furrowing**

The statistical testing was performed using pairwise two-sample t-tests. As can be seen from the figure, standard deviations were comparable between groups. To attain a higher precision we therefore used a pooled standard deviation for all of the tests. p-values were corrected for multiple testing using the method of Bonferroni. No significant differences were detected between the groups. The crosses show group median (horizontal line) and interquartile range (vertical line).

### 3.5 Boxes Far vs Boxes Close

We wished to compare samples from the double activation experiments where two boxes of activation were placed at different distances from one another. Specifically, we wished to quantify the a-p anisotropy of cells in the central area that was not activated. As the mCherry signal in this region was predominantly cytoplasmic, it was not possible to obtain an accurate segmentation of cell shape. We therefore instead imaged a single time point at the end of each experiment in the GFP channel, with which cell shapes could be resolved. For this section the most apical z-stack that gave an accurate cell segmentation was kept for the analysis. A box isolating a group of constricting cells was chosen manually by the biologist.

```
tabBoxes <- filter(opto, condition %in% c("BoxesFar", "NewBoxesFar",
                                         "BoxesClose"))
summaryBoxes <- tabBoxes %>% group_by(condition, sample) %>%
  summarize(APanisotropy=median(e.x, na.rm=TRUE),
            APanisotropy25=quantile(e.x, probs=0.25, na.rm=TRUE),
            APanisotropy75=quantile(e.x, probs=0.75, na.rm=TRUE))

## `summarise()` regrouping output by 'condition' (override with '.groups' argument)
summaryBoxes$condition[summaryBoxes$condition == "NewBoxesFar"] <- "BoxesFar"
summaryBoxes$condition <- droplevels(summaryBoxes$condition)

ttDouble <- t.test(APanisotropy ~ condition, summaryBoxes)
print(ttDouble)

##
## Welch Two Sample t-test
##
## data: APanisotropy by condition
## t = -3.7062, df = 8.3088, p-value = 0.005603
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.35250238 -0.08317076
## sample estimates:
## mean in group BoxesClose   mean in group BoxesFar
##           0.4211512              0.6389877
```

## Automatic generation of paper figures

```

diff(ttDouble$estimate)

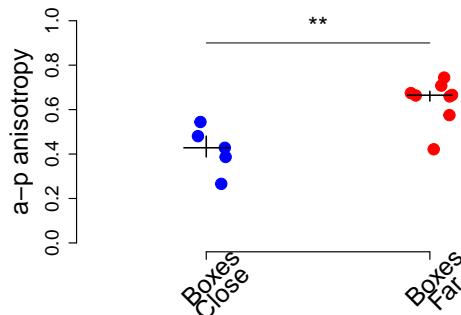
## mean in group BoxesFar
##          0.2178366

par(mar=c(6.1, 6.1, 3.1, 2.1))
stripchart(APanisotropy ~ condition, summaryBoxes, vertical=TRUE, cex.lab=2,
           xlab="", ylab="a-p anisotropy", xlim=c(0.5, 2.5), ylim=c(0, 1), pch=19,
           col=c("blue", "red"), method="jitter", jitter=0.1, axes=FALSE, cex=2)
axis(side=1, at=1:2, labels=FALSE, lwd=1.5)
text(x=1:2, par("usr")[3]-0.06, labels=rep("Boxes", 2), srt=45, pos=1, xpd=TRUE,
      cex=2)
text(x=1:2+0.05, par("usr")[3]-0.13, labels=c("Close", "Far"), srt=45, pos=1,
      xpd=TRUE, cex=2)
axis(2, cex.axis=1.5, lwd=1.5)
drawBar(1, 2, y=0.9, star=mySignIf(ttDouble$p.value), offset=0.1, cex=2)
conditionSummary <- summaryBoxes %>% group_by(condition) %>%
  summarize(median=median(APanisotropy),
            q25=quantile(APanisotropy, probs=0.25),
            q75=quantile(APanisotropy, probs=0.75))

## 'summarise()' ungrouping output (override with '.groups' argument)

conditionSummary <- as.data.frame(conditionSummary)
for (i in seq_len(nrow(conditionSummary))) {
  segments(x0=i-0.1, x1=i+0.1, y0=conditionSummary[i, "median"], lwd=2, lty=1)
  segments(x0=i, y0=conditionSummary[i, "q25"], y1=conditionSummary[i, "q75"],
            lwd=2, lty=1)
}

```



**Figure 6: Fig 7G: The median a-p anisotropy of groups of constricting cells were compared between samples where the distance between two boxes of light activation was varied**

A two-sample t-test revealed that cells were more elongated along the embryo a-p axis in the experiment where the boxes were placed farther apart. As can be seen from the plot, the two groups showed unequal variances. We therefore used Welch's t-test. The crosses show group median (horizontal line) and interquartile range (vertical line). The magnitude of difference in a-p anisotropy between the two groups was found to be 0.23.

## Automatic generation of paper figures

### 3.6 Global 2p

We assessed the a-p anisotropy of samples that were globally activated with the 2-photon microscope at different powers.

```
tabPower700uW <- filter(opto, condition == "700uW")
tabPower700uW[tabPower700uW$sample == "New data/700uW/1_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower700uW[tabPower700uW$sample == "New data/700uW/2_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower700uW[tabPower700uW$sample == "New data/700uW/3_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
plotSamples700uW <- "New data/700uW/3_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"
dvFurrowPos <- tabPower700uW %>% group_by(sample) %>% filter(t == tstar) %>%
  do(fnPos(., plotSamples="", nbinsExclude=0))
for (i in seq_along(dvFurrowPos$sample)) {
  tabPower700uW[tabPower700uW$sample %in% dvFurrowPos$sample[i], "furrowPos"] <- dvFurrowPos$furrowPos[i]
}
boxes <- tabPower700uW %>% group_by(sample) %>% do(fnBox(., w=724))
tabPower700uWs <- tabPower700uW %>% group_by(sample) %>%
  do(fnIsolate(., boxes=boxes))
tMinArea <- tabPower700uWs %>% group_by(sample) %>% do(fnMinArea(., plot=FALSE))
for (i in seq_along(tMinArea$sample)) {
  tabPower700uWs[tabPower700uWs$sample == tMinArea$sample[i], "tstar"] <- tMinArea$t[i]
}
tabPower700uWs <- filter(tabPower700uWs, t <= tstar)

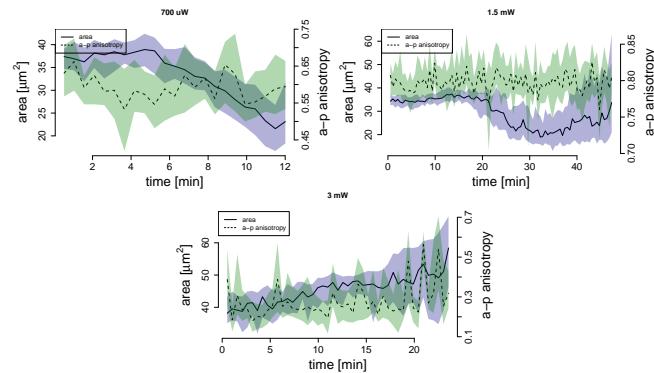
plotSamples1.5mW <- "New data/Global_2p/2_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"
tabPower1.5mW <- filter(opto, condition == "Global2p")
tabPower1.5mW$condition <- "1.5mW"
tabPower1.5mW[tabPower1.5mW$sample == "New data/Global_2p/2_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower1.5mW[tabPower1.5mW$sample == "New data/Global_2p/4_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower1.5mW[tabPower1.5mW$sample == "New data/Global_2p/8_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
dvFurrowPos <- tabPower1.5mW %>% group_by(sample) %>% filter(t == tstar) %>%
  do(fnPos(., plotSamples="", nbinsExclude=0))
for (i in seq_along(dvFurrowPos$sample)) {
  tabPower1.5mW[tabPower1.5mW$sample %in% dvFurrowPos$sample[i], "furrowPos"] <- dvFurrowPos$furrowPos[i]
}
boxes <- tabPower1.5mW %>% group_by(sample) %>% do(fnBox(., w=724))
tabPower1.5mWs <- tabPower1.5mW %>% group_by(sample) %>%
  do(fnIsolate(., boxes=boxes))

plotSamples3mW <- "New data/3mW/1_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"
tabPower3mW <- filter(opto, condition == "3mW")
tabPower3mW[tabPower3mW$sample == "New data/3mW/1_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower3mW[tabPower3mW$sample == "New data/3mW/2_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
tabPower3mW[tabPower3mW$sample == "New data/3mW/3_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda"] <- NULL
boxes <- tabPower3mW %>% group_by(sample) %>% do(fnBox(., w=724))
tabPower3mWs <- tabPower3mW %>% group_by(sample) %>% fnIsolate(., boxes=boxes)

tabPower <- bind_rows(tabPower700uWs, tabPower1.5mWs, tabPower3mWs)
tabPower$t <- tabPower$t*tabPower$dt
plotSamples <- c(
  "700 uW"="New data/700uW/3_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda",
  "1.5 mW"="New data/Global_2p/2_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda",
  "3 mW"="New data/3mW/1_CIBNOCRL__R001_GR1_B1_SUM/furrowSeg1.0.47_threshOffset5e-04_closingSize3.rda")
for (i in seq_along(plotSamples)) {
```

## Automatic generation of paper figures

```
s <- plotSamples[i]
tabPowers <- filter(tabPower, sample == s)
px <- tabPowers$px[1]
plotFeatureEvolution(tabPowers, dt=1/60, tMax=Inf, myTitle=names(s),
px=px, cex=2, cex.axis=1.5, line=3.3, mar=c(5.1, 5.6, 4.1, 4.6))
}
```



**Figure 7: Fig S4N-P: Cell area and a-p anisotropy are plotted over time for all cells in a globally light-activated sample that was activated with different two-photon laser powers**

Solid and dashed lines indicate the median over cells for area and a-p anisotropy, respectively, while shaded regions show the interquartile range.