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Age-related remodelling of oesophageal epithelia by mutated cancer drivers

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Supplementary Method 1: Flow chart of mutation calling

Sequencing of samples with matched germline controls



Sequencing data alignment to hg19 (bwa v.0.7.10)



SNV and insertion/deletion calling by Genomon 2 pipeline



Mapping quality ≥ 30 , Base quality ≥ 15

Reads having < 5 SNVs or indels

Filtering somatic mutation with high fidelity (n=8,056)



Total reads in sample ≥ 10 , Variant reads in sample ≥ 3

Total reads in control ≥ 10 , Variant reads in control ≥ 1

VOF in sample ≥ 0.05 , VOF in control ≤ 0.02

EBCall *p-values* $\leq 10^{-5}$, Fisher *p-value* $\leq 10^{-2}$

Strand ratio $\neq 0$ or 1

Mutations outside of repetitive region



Visual inspection on IGV (n=126)

Somatic mutations (n=7,930)



Validation by PCR-based amplicon sequencing



595/622 validated (True positive rate: 95.7%)

Rescue for Driver genes

Total reads in sample ≥ 10 , Variant reads in sample ≥ 3

Total reads in control ≥ 10 , Variant reads in control ≤ 1

VOF in sample ≥ 0.05 , VOF in control ≤ 0.02

EBCall *p-values* $\leq 10^{-5}$, Fisher *p-value* $\leq 10^{-2}$

hg19; Homo sapiens (human) genome assembly GRCh37

bwa; Burrows-Wheeler Aligner

SNV; single nucleotide variant

VOF; variant allele fraction

EBCall; Empirical Bayesian mutation Calling

IGV; Integrative Genomics Viewer

Supplementary Method 2; a sample script for a random effects model meta-analysis

```
#####  
# load library  
library(metafor)  
#####  
# function for calculating the coordinate of intersection point of perpendicular line from (x,y) to line with "a" slope  
# and "b" intercept  
calc_dis <- function(a, b, x1, y1){  
  x <- (x1+a*y1-b*a)/(a^2+1)  
  return(c(x=x, y=a*x+b))  
}  
#####  
# data read  
res <- read.table("Path to data_for_script.txt")  
  
res_nd <- subset(res, Path=="nondysplasia")  
res_nd_4 <- subset(res_nd, Size==4)  
res_nd_0.8 <- subset(res_nd, Size==0.8)  
res_nd_0.2 <- subset(res_nd, Size==0.2)  
res_nd_4_l <- subset(res_nd_4, ESCC_risk==0)  
res_nd_4_h <- subset(res_nd_4, ESCC_risk==1)  
res_nd_0.8_l <- subset(res_nd_0.8, ESCC_risk==0)  
res_nd_0.8_h <- subset(res_nd_0.8, ESCC_risk==1)  
res_nd_0.2_l <- subset(res_nd_0.2, ESCC_risk==0)  
res_nd_0.2_h <- subset(res_nd_0.2, ESCC_risk==1)  
#####  
# This is an example of the code to calculate effect of ESCC risks on "mutation number".  
# Modify this code for other metrics.  
#####  
# For 4mm2 sample  
# Set up the regression line of samples without risk  
out_4_l <- lm(res_nd_4_l$mut_num~res_nd_4_l$age-1);out_4_l  
out2_4_l <- summary(out_4_l)  
  
# Calculate the distance from the regression line to each risk (-) sample
```

```

dis_info_4_l <- NULL
for(i in 1:nrow(res_nd_4_l)){
  tmp_age <- res_nd_4_l[i,"age"]
  tmp_num <- res_nd_4_l[i,"mut_num"]
  tmp_res <- calc_dis(out_4_l$coefficients[1], 0,tmp_age, tmp_num)
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
  if(out_4_l$coefficients[1]*tmp_age -tmp_num<0){
    judge <- 1 #above
    dis <- dis
  }else{
    judge <- 0 # below
    dis <- -dis
  }
  tmp_res2 <- c(dis,judge)
  dis_info_4_l <- rbind(dis_info_4_l,tmp_res2)
}
colnames(dis_info_4_l) <- c("dis","judge")
rownames(dis_info_4_l) <- 1:nrow(dis_info_4_l)
dis_info_4_l <- as.data.frame(dis_info_4_l)

# Set up the regression line of samples with risk
out_4_h <- lm(res_nd_4_h$mut_num~res_nd_4_h$age-1);out_4_h
out2_4_h <- summary(out_4_h)

# Calculate the distance from the regression line to each risk (+) sample
dis_info_4_h <- NULL
for(i in 1:nrow(res_nd_4_h)){
  tmp_age <- res_nd_4_h[i,"age"]
  tmp_num <- res_nd_4_h[i,"mut_num"]
  tmp_res <- calc_dis(out_4_l$coefficients[1], 0,tmp_age, tmp_num)
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
  if(out_4_l$coefficients[1]*tmp_age -tmp_num<0){
    judge <- 1 #above
    dis <- dis
  }else{
    judge <- 0 # below

```

```

dis <- -dis
}
tmp_res2 <- c(dis,judge)
dis_info_4_h <- rbind(dis_info_4_h,tmp_res2)
}
colnames(dis_info_4_h) <- c("dis","judge")
rownames(dis_info_4_h) <- 1:nrow(dis_info_4_h)
dis_info_4_h <- as.data.frame(dis_info_4_h)

# Mann-Whitney U test
w.out <- wilcox.test(dis_info_4_h$dis,dis_info_4_l$dis, alternative="greater")

#####

# For 0.8mm2 sample
# Set up the regression line of samples without risk
out_0.8_l <- lm(res_nd_0.8_l$mut_num~res_nd_0.8_l$age-1);out_0.8_l
out2_0.8_l <- summary(out_0.8_l)

# Calculate the distance from the regression line to each risk (-) sample
dis_info_0.8_l <- NULL
for(i in 1:nrow(res_nd_0.8_l)){
  tmp_age <- res_nd_0.8_l[i,"age"]
  tmp_num <- res_nd_0.8_l[i,"mut_num"]
  tmp_res <- calc_dis(out_0.8_l$coefficients[1], 0,tmp_age, tmp_num)
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
  if(out_0.8_l$coefficients[1]*tmp_age -tmp_num<0){
    judge <- 1 #above
    dis <- dis
  }else{
    judge <- 0 # below
    dis <- -dis
  }
  tmp_res2 <- c(dis,judge)
  dis_info_0.8_l <- rbind(dis_info_0.8_l,tmp_res2)
}
colnames(dis_info_0.8_l) <- c("dis","judge")

```

```

rownames(dis_info_0.8_l) <- 1:nrow(dis_info_0.8_l)
dis_info_0.8_l <- as.data.frame(dis_info_0.8_l)

# Set up the regression line of samples with risk
out_0.8_h <- lm(res_nd_0.8_h$mut_num~res_nd_0.8_h$age-1);out_0.8_h
out2_0.8_h <- summary(out_0.8_h)

# Calculate the distance from the regression line to each risk (+) sample
dis_info_0.8_h <- NULL
for(i in 1:nrow(res_nd_0.8_h)){
  tmp_age <- res_nd_0.8_h[i,"age"]
  tmp_num <- res_nd_0.8_h[i,"mut_num"]
  tmp_res <- calc_dis(out_0.8_l$coefficients[1], 0,tmp_age, tmp_num)
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
  if(out_0.8_l$coefficients[1]*tmp_age -tmp_num<0){
    judge <- 1 #above
    dis <- dis
  }else{
    judge <- 0 # below
    dis <- -dis
  }
  tmp_res2 <- c(dis,judge)
  dis_info_0.8_h <- rbind(dis_info_0.8_h,tmp_res2)
}
colnames(dis_info_0.8_h) <- c("dis","judge")
rownames(dis_info_0.8_h) <- 1:nrow(dis_info_0.8_h)
dis_info_0.8_h <- as.data.frame(dis_info_0.8_h)

# Mann-Whitney U test
w.out <- wilcox.test(dis_info_0.8_h$dis,dis_info_0.8_l$dis, alternative="greater")

#####

# For 0.2mm2 sample
# Set up the regression line of samples without risk
out_0.2_l <- lm(res_nd_0.2_l$mut_num~res_nd_0.2_l$age-1);out_0.2_l
out2_0.2_l <- summary(out_0.2_l)

```

```
# Calculate the distance from the regression line to each risk (-) sample
```

```
dis_info_0.2_l <- NULL
```

```
for(i in 1:nrow(res_nd_0.2_l)){
```

```
  tmp_age <- res_nd_0.2_l[i,"age"]
```

```
  tmp_num <- res_nd_0.2_l[i,"mut_num"]
```

```
  tmp_res <- calc_dis(out_0.2_l$coefficients[1], 0,tmp_age, tmp_num)
```

```
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
```

```
  if(out_0.2_l$coefficients[1]*tmp_age -tmp_num<0){
```

```
    judge <- 1 #above
```

```
    dis <- dis
```

```
  }else{
```

```
    judge <- 0 # below
```

```
    dis <- -dis
```

```
  }
```

```
  tmp_res2 <- c(dis,judge)
```

```
  dis_info_0.2_l <- rbind(dis_info_0.2_l,tmp_res2)
```

```
}
```

```
colnames(dis_info_0.2_l) <- c("dis","judge")
```

```
rownames(dis_info_0.2_l) <- 1:nrow(dis_info_0.2_l)
```

```
dis_info_0.2_l <- as.data.frame(dis_info_0.2_l)
```

```
# Set up the regression line of samples with risk
```

```
out_0.2_h <- lm(res_nd_0.2_h$mut_num~res_nd_0.2_h$age-1);out_0.2_h
```

```
out2_0.2_h <- summary(out_0.2_h)
```

```
# Calculate the distance from the regression line to each risk (+) sample
```

```
dis_info_0.2_h <- NULL
```

```
for(i in 1:nrow(res_nd_0.2_h)){
```

```
  tmp_age <- res_nd_0.2_h[i,"age"]
```

```
  tmp_num <- res_nd_0.2_h[i,"mut_num"]
```

```
  tmp_res <- calc_dis(out_0.2_h$coefficients[1], 0,tmp_age, tmp_num)
```

```
  dis <- sqrt((tmp_age- tmp_res[1])^2+(tmp_num- tmp_res[2])^2)
```

```
  if(out_0.2_h$coefficients[1]*tmp_age -tmp_num<0){
```

```
    judge <- 1 #above
```

```
    dis <- dis
```

```

}else{
  judge <- 0 # below
  dis <- -dis
}
tmp_res2 <- c(dis,judge)
dis_info_0.2_h <- rbind(dis_info_0.2_h,tmp_res2)
}
colnames(dis_info_0.2_h) <- c("dis","judge")
rownames(dis_info_0.2_h) <- 1:nrow(dis_info_0.2_h)
dis_info_0.2_h <- as.data.frame(dis_info_0.2_h)

# Mann-Whitney U test
w.out <- wilcox.test(dis_info_0.2_h$dis,dis_info_0.2_l$dis, alternative="greater")

# Apply the same calculation to 0.8mm2 and 0.2 mm2 samples

# Meta analysis for ESCC risk
out.m <- rbind(c(mean(dis_info_0.2_h$dis),sd(dis_info_0.2_h$dis),length(dis_info_0.2_h$dis),
  mean(dis_info_0.2_l$dis),sd(dis_info_0.2_l$dis),length(dis_info_0.2_l$dis)),
  c(mean(dis_info_0.8_h$dis),sd(dis_info_0.8_h$dis),length(dis_info_0.8_h$dis),
  mean(dis_info_0.8_l$dis),sd(dis_info_0.8_l$dis),length(dis_info_0.8_l$dis)),
  c(mean(dis_info_4_h$dis),sd(dis_info_4_h$dis),length(dis_info_4_h$dis),
  mean(dis_info_4_l$dis),sd(dis_info_4_l$dis),length(dis_info_4_l$dis)))
colnames(out.m) <- c("mean_h","sd_h","num_h","mean_l","sd_l","num_l")
out.m <- as.data.frame(out.m)

# Calculate effect sizes and outcome measures
# "SMDH" for the standardized mean difference with heteroscedastic population variances in the two groups (Bonett,
2008, 2009).
out.m2 <- escalc(measure="SMDH", m1i=mean_h,m2i=mean_l,
  sd1i=sd_h,sd2i=sd_l,n1i=num_h,n2i=num_l, data=out.m)

# Random-effects model
out.m3=rma(yi, vi, data=out.m2)

# Plot forest plot

```

```

forest(out.m3,showweights=T,alim=c(-4,4),at=(-4:4),
      slab=c("0.2 mm2", "0.8 mm2", "4 mm2"),cex=1.2,cex.lab=1,cex.axis=1,
      main="Number of mutations and ESCC risk")
text(1,5,paste("P=",signif(out.m3$pval,3)))

#####

# Meta analysis for age
out.m <- rbind(c(sqrt(out2_0.2_h$r.squared),length(dis_info_0.2_h$dis),
      sqrt(out2_0.2_l$r.squared),length(dis_info_0.2_l$dis)),
      c(sqrt(out2_0.8_h$r.squared),length(dis_info_0.8_h$dis),
      sqrt(out2_0.8_l$r.squared),length(dis_info_0.8_l$dis)),
      c(sqrt(out2_4_h$r.squared),length(dis_info_4_h$dis),
      sqrt(out2_4_l$r.squared),length(dis_info_4_l$dis)))
colnames(out.m) <- c("coef_h","num_h","coef_l","num_l")
out.m <- as.data.frame(out.m)

# Calculate effect sizes and outcome measures
# "ZCOR" for the Fisher's r-to-z transformed correlation coefficient (Fisher, 1921).
out.m2 <- escalc(measure="ZCOR", ri=coef_l,ni=num_l,data=out.m)

# Random-effects model
out.m3=rma(yi, vi, data=out.m2)

# Plot forest plot
forest(out.m3,showweights=T,alim=c(-2,2),at=(0:20-10)/10,
      slab=c("0.2 mm2", "0.8 mm2", "4 mm2"),cex=1.2,cex.lab=1,cex.axis=1,
      main="Number of mutations and age in risk (-)")
text(1,5,paste("P=",signif(out.m3$pval,3)))

```

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