

# Targeted exome-based predictors of patterns of progression of colorectal liver metastasis after percutaneous thermal ablation

Statistical Analysis

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## Contents

<b>1</b>	<b>Setup</b>	<b>3</b>
<b>2</b>	<b>Prepare Data</b>	<b>4</b>
2.1	Download data . . . . .	4
2.2	Load data . . . . .	4
2.3	Filter . . . . .	4
2.4	Preprocess . . . . .	6
2.5	Extract mutations . . . . .	8
<b>3</b>	<b>Population</b>	<b>10</b>
3.1	Baseline characteristics . . . . .	10
3.2	Baseline survival . . . . .	11
3.3	Overview mutations . . . . .	13
3.4	Type of progression . . . . .	14
<b>4</b>	<b>Overall survival</b>	<b>15</b>
4.1	By progression pattern . . . . .	15
<b>5</b>	<b>Intrahepatic progression anywhere</b>	<b>19</b>
5.1	Intrahepatic progression outside the ablation zone . . . . .	23
5.2	Intrahepatic progression at the ablation zone . . . . .	28
<b>6</b>	<b>Salvage LRT</b>	<b>32</b>
6.1	Survival following IHP . . . . .	35
6.2	Survival after IHP - salvage LRT vs. no salvage LRT . . . . .	36

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<b>7</b>	<b>Frequency of mutations</b>	<b>39</b>
7.1	Number of mutations per pathway with predominant gene . . . . .	39
7.2	Number of mutations per pathway . . . . .	42
7.3	Frequency of mutations . . . . .	43
7.4	Mutation per race/ethnicity . . . . .	43
<b>8</b>	<b>Citation</b>	<b>45</b>

# 1 Setup

```
library(tidyverse)
library(config)
library(readr)
library(dplyr)
library(survival)
library(survminer)
library(stringr)
library(parameters)
library(Rfast)
library(ggpubr)
library(tidyr)
library(forestmodel)
library(ggsci)
library(data.table)
library(gtsummary)
library(gt)
source("R/write_params_to_csv.cox.R")
source("R/finegray_model.R")
options(scipen = 999, digits = 3)
```

## 2 Prepare Data

### 2.1 Download data

Download data from RedCAP API. Code omitted.

### 2.2 Load data

```
data_raw <- read_csv("data/data.csv")
```

### 2.3 Filter

Remove excluded cases.

```
data <- data_raw %>% filter(date_of_ablation >= "2011-01-01" & !(exclude_reason %in% c(6, 9)))
```

226 patients with initial CLM ablation between 01/01/2011 - 03/30/2022.

```
data %>% select(exclude_reason) %>%  
  mutate(exclude_reason = factor(exclude_reason,  
                                labels =  
                                c("< 1 Year imaging follow-up",  
                                  "Non-thermal ablation modality (e.g. IRE, Cryo)",  
                                  "Genomic mutation profile missing",  
                                  "< 46 genes tested",  
                                  "Combination with Surgery, TACE or other LRT"))) %>%  
  tbl_summary(label = list(exclude_reason = 'Reason for exclusion'))
```

Characteristic	N = 226
Reason for exclusion	
< 1 Year imaging follow-up	23 (29%)
Non-thermal ablation modality (e.g. IRE, Cryo)	7 (8.8%)
Genomic mutation profile missing	41 (51%)
< 46 genes tested	4 (5.0%)
Combination with Surgery, TACE or other LRT	5 (6.2%)
Unknown	146

```
data <- data %>% filter(exclude_reason %in% c(NA, 3, 4))  
  
data$mutation.analysis <- data$exclude_reason %in% c(NA) & !is.na(data$apc)  
  
data %>% select(mutation.analysis) %>%  
  tbl_summary(label = list(mutation.analysis = 'Included for mutation analysis'))
```

<b>Characteristic</b>	<b>N = 191</b>
Included for mutation analysis	101 (53%)

Total 191 included for progression pattern analysis and 101 for mutation analysis.

## 2.4 Preprocess

```
data <- within(data, {
  deceased = deceased == 1
  gt_46_tested <- data$nr_tested_genes >= 46
  nr_tested_genes = ifelse(is.na(nr_tested_genes), 0, nr_tested_genes)
  a0_ablation = a0_ablation == 1
  n_positive = n_positive == 1
  max_diameter = max_diameter / 10.0
  repeat_lrt = repeat_lrt == 1
  prior_resection = prior_resection == 1
  os_m = overall_survival / 30.25
  metastasis = factor(metastasis_sync,
                      levels = c("sync", "meta"),
                      labels = c("Synchronous", "Metachronous"))
})

data <- within(data, {
  ablation_modality = as.factor(ablation_modality)
  cohort = factor(cohort, labels = c("2005-2015", "2015-2020"))
  year_num = as.numeric(format(data$date_of_ablation, '%Y'))
  year = as.factor(format(data$date_of_ablation, '%Y'))
})

data <- within(data, {
  ltp = ifelse(intrahepatic_recurrence == FALSE, FALSE, local_recurrence == 1)
  ltpfs = ifelse(ltp, time_to_lc, time_to_lfu)
  ihp_outside = ifelse(intrahepatic_recurrence == FALSE, FALSE, new_intrahepatic_lesions == 1)
  ihpfs_outside = ifelse(ihp_outside, time_to_new_lesions, time_to_lfu)
  repeat_lrt = ifelse(intrahepatic_recurrence, repeat_lrt, FALSE)
  ihp_any = ltp | ihp_outside
  ihpfs_any = ifelse(ihp_any, ifelse(ltpfs < ihpfs_outside, ltpfs, ihpfs_outside), time_to_lfu)
})

data$os_ihp = data$overall_survival - data$ihpfs_any
data$os_ihp_m = data$os_ihp / 30.25

data <- within(data, {
  ehp = extrahepatic_recurrence == 1
  ehpf = ifelse(ehp, time_to_ehp, time_to_lfu)
  ehmf = ehmf == 1
})

data$any_progression <- data$ltp | data$ihp_outside | data$ehp

data <- within(data, {
  chemo_yes_no = chemo_yes_no == 1
  chemo_lines = ifelse(chemo_lines == "NAVU", NA, as.numeric(chemo_lines))
})

tmp = list(primary_side = factor(data$primary_location_detail))
```

```

levels(tmp$primary_side) <- list(
  Right = c("ce","ac", "hf"),
  Left = c("dc","r","rs", "sc","sf"),
  Transverse = c("tc")
)
data$primary_side <- tmp$primary_side

data <- within(data, {
  right_side = primary_side == "Right"
  ehpf_before = ifelse(ehp,
    ifelse(ihp_any, ehpf <= ihpf_outside & ehpf <= ltpf, TRUE), FALSE)
  time_to_lfu_m = time_to_lfu / 30.25
  time_to_lfu_y = time_to_lfu / 365
  ehpf_m = ehpf / 30.25
  ihpf_outside_m = ihpf_outside / 30.25
  ltpf = ltpf / 30.25
  ehpf_before = ifelse(ehpf_before, ehpf, ihpf_any)
})

```

## 2.5 Extract mutations

```
cols.p53.logic <- c("mdm2","mdm4", "cdkn2a", "tp53", "atm")
data[cols.p53.logic] <- sapply(data[cols.p53.logic], as.logical)

cols.tgfb.logic <- c("smad4")
data[cols.tgfb.logic] <- sapply(data[cols.tgfb.logic], as.logical)

cols.wnt.logic <- c("rnf43", "apc", "ctnnb1")
data[cols.wnt.logic] <- sapply(data[cols.wnt.logic], as.logical)

cols.ras.logic <- c("egfr", "erbb2", "erbb3", "erbb4", "met", "pdgfra", "fgfr1",
                  "fgfr2", "fgfr3", "fgfr4", "kit", "igf1r", "ret", "ros1", "alk",
                  "flt3", "ntrk1", "jak2", "cbl", "abl1", "nf1", "ptpn11",
                  "kras", "nras", "hras", "araf", "braf", "raf1", "rac1",
                  "mapk1", "map2k1", "map2k2", "ntrk2", "ntrk3", "crebbp")
data[cols.ras.logic] <- sapply(data[cols.ras.logic], as.logical)

cols.pi3k.logic <- c("pik3ca", "pten", "pik3cb", "pik3r1", "akt1", "akt2", "akt3",
                  "ppp2r1a", "stk11", "tsc1", "tsc2", "rheb", "riCTOR", "mtor")
data[cols.pi3k.logic] <- sapply(data[cols.pi3k.logic], as.logical)

cols.wnt.logic <- c("rnf43", "apc", "ctnnb1")
data[cols.wnt.logic] <- sapply(data[cols.wnt.logic], as.logical)

cols.notch.logic <- c("fbxw7", "notch1", "notch2", "notch3")
data[cols.notch.logic] <- sapply(data[cols.notch.logic], as.logical)

cols.cell.logic <- c("cdkn1b", "cdkn2b", "ccne1", "rb1", "ccnd1", "ccnd2", "ccnd3",
                  "cdk2", "cdk4", "cdk6")
data[cols.cell.logic] <- sapply(data[cols.cell.logic], as.logical)
```

### 2.5.1 Grouping KRAS, NRAS, BRAF into one category RAS/BRAF

```
data$rasbraf <- data$kras | data$nras | data$braf
```

### 2.5.2 Grouping pathways

```
data <- data %>%
  rowwise() %>%
  mutate(
    p53 = any(mdm2, mdm4, cdkn2a, tp53, atm, na.rm = TRUE),
    tgfb = any(sm4d, na.rm = TRUE),
    wnt = any(rnf43, apc, ctnnb1, na.rm = TRUE),
    ras = any(egfr, erbb2, erbb3, erbb4, met, pdgfra, fgfr1, fgfr2, fgfr3,
             fgfr4, kit, igf1r, ret, ros1, alk,
             flt3, ntrk1, jak2, cbl, abl1, nf1, ptpn11, kras, hras, nras,
             araf, braf, raf1, rac1, mapk1, map2k1,
```

```
        map2k2, ntrk2, ntrk3, crebbp, na.rm = TRUE),
  pi3k = any(pik3ca, pten, akt1, pik3cb, pik3r1, akt2, akt3, ppp2r1a, stk11,
            tsc1, tsc2, rheb, rictor, mtor, na.rm = TRUE),
  notch = any(fbxw7, notch1, notch2, notch3, na.rm=TRUE),
  cell = any(cdkn1b, cdkn2b, cdkn2a, ccn1, rb1, ccnd1, ccnd2, ccnd3, cdk2,
            cdk4, cdk6, na.rm = TRUE)
)
data$wnt <- ifelse(is.na(data$apc), NA, data$wnt)
data <- within(data, {
  num_altered_paths = (p53 + tgfb + ras + pi3k + notch + cell + wnt)
})
```

## 3 Population

### 3.1 Baseline characteristics

```
data.tbl <- data %>%
  haven::as_factor() %>%
  mutate(chemo_lines = as.factor(chemo_lines)) %>%
  select(age_at_procedure, sex, primary_location, right_side, metastasis, n_positive,
         prior_resection, ehm, nr_of_tumors, max_diameter, time_to_lfu_m,
         chemo_yes_no, chemo_lines, chemo_regimen___5fu1, chemo_regimen___ox,
         chemo_regimen___iri, chemo_regimen___bev) %>%
  tbl_summary(
    label = list(age_at_procedure ~ "Age",
                 sex ~ "Sex",
                 primary_location ~ "Primary location",
                 right_side ~ "Right-sided primary tumor",
                 metastasis ~ "Metastasis",
                 n_positive ~ "Positive lymphnodes",
                 prior_resection ~ "Prior resection",
                 ehm ~ "Extrahepatic metastasis",
                 chemo_yes_no ~ "Pre-ablation chemotherapy",
                 chemo_lines ~ "Lines of pre-ablation chemotherapy",
                 chemo_regimen___5fu1 ~ "Fluorouracil-based regimen",
                 chemo_regimen___ox ~ "Oxaliplatin",
                 chemo_regimen___iri ~ "Irinotecan",
                 chemo_regimen___bev ~ "Use of Bevacizumab",
                 nr_of_tumors ~ "Nr of ablated CLM",
                 max_diameter ~ "Max CLM diameter",
                 time_to_lfu_m ~ "Time to last FU"),
    type = list(nr_of_tumors ~ 'continuous'),
    statistic = list(all_continuous() ~ "{median} [{p25}, {p75}]"),
    digits = list(all_continuous() ~ 2,
                  all_categorical() ~ 1),
    missing_text = "N/A") %>%
  bold_labels()
```

data.tbl

Characteristic	N = 191
Age	56.70 [48.60, 66.55]
Sex	
f	70.0 (36.6%)
m	121.0 (63.4%)
Primary location	
colon	152.0 (79.6%)
rectum	39.0 (20.4%)
Right-sided primary tumor	45.0 (23.6%)
Metastasis	
Synchronous	102.0 (54.3%)
Metachronous	86.0 (45.7%)

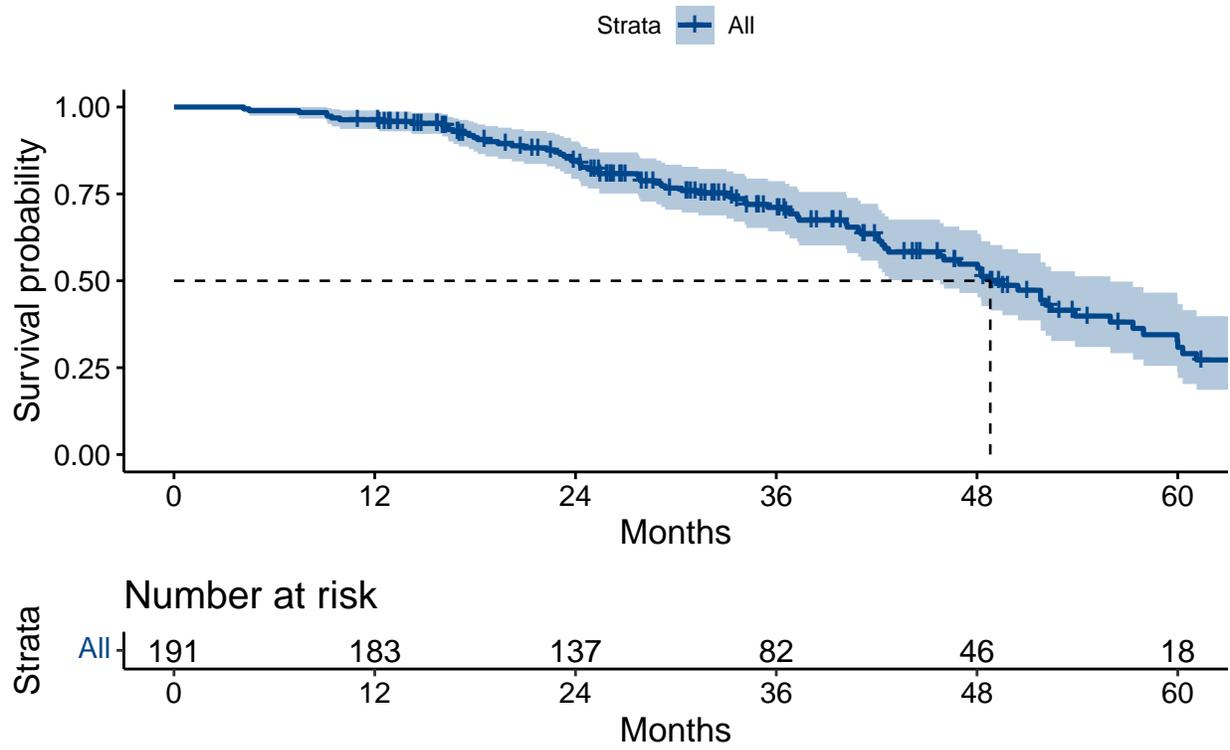
Characteristic	N = 191
N/A	3
Positive lymphnodes	114.0 (59.7%)
Prior resection	117.0 (61.3%)
Extrahepatic metastasis	111.0 (58.1%)
Nr of ablated CLM	1.00 [1.00, 2.00]
Max CLM diameter	1.40 [1.00, 2.00]
Time to last FU	31.50 [18.46, 44.26]
Pre-ablation chemotherapy	142.0 (74.3%)
Lines of pre-ablation chemotherapy	
1	83.0 (43.5%)
2	54.0 (28.3%)
3	5.0 (2.6%)
0	49.0 (25.7%)
Fluorouracil-based regimen	127.0 (66.5%)
Oxaliplatin	74.0 (38.7%)
Irinotecan	70.0 (36.6%)
Use of Bevacizumab	90.0 (47.1%)

```
data.tbl %>% as_gt() %>% gtsave(filename = "out/population.rtf")
```

### 3.2 Baseline survival

```
data$clm <- TRUE

ggsurvplot(survfit(Surv(os_m, deceased) ~ 1, data = data),
  data = data,
  risk.table = TRUE, tables.height = 0.25,
  surv.median.line = 'hv',
  xlim = c(0, 60), break.time.by = 12, xlab = "Months",
  palette = "lancet")
```



### 3.3 Overview mutations

```
mutations.tbl <- data %>%
  filter(mutation.analysis) %>%
  haven::as_factor() %>%
  select(p53, tp53, ras, rasbraf, wnt, apc, tgfb, smad4, pi3k, pik3ca,
         notch, fbxw7, cell, rb1) %>%
  tbl_summary(
    digits = list(all_categorical() ~ 1),
    missing_text = "N/A") %>%
  bold_labels()

mutations.tbl
```

Characteristic	N = 101
<b>p53</b>	75.0 (74.3%)
<b>tp53</b>	75.0 (74.3%)
<b>ras</b>	48.0 (47.5%)
<b>rasbraf</b>	42.0 (41.6%)
<b>wnt</b>	64.0 (63.4%)
<b>apc</b>	61.0 (60.4%)
<b>tgfb</b>	16.0 (15.8%)
<b>smad4</b>	16.0 (15.8%)
<b>pi3k</b>	17.0 (16.8%)
<b>pik3ca</b>	12.0 (11.9%)
<b>notch</b>	5.0 (5.0%)
<b>fbxw7</b>	5.0 (5.0%)
<b>cell</b>	4.0 (4.0%)
<b>rb1</b>	4.0 (4.0%)

Exclude RB1 and Cell Cycle pathway from further analysis

### 3.4 Type of progression

```

data$multiple_progression <- apply(data[,c('ihp_outside', 'ltp', 'ehp')], 1, sum) > 1
data$ihp_both <- data$ltp & data$ihp_outside
data$no_progression <- !(data$ltp | data$ihp_outside | data$ehp)

progression.tbl <- data %>%
  select(ltp, ihp_outside, ihp_both, ihp_any, ehpf, no_progression,
         multiple_progression, repeat_lrt, deceased) %>%
  tbl_summary(
    by = deceased,
    label = list(
      ltp = 'Local tumor progression',
      ihp_outside = "Intrahepatic progression outside the ablation zone",
      ihp_both = "Local and outside intrahepatic progression",
      ihp_any = 'Any kind of intrahepatic progression',
      ehpf = 'Extrahepatic progression',
      no_progression = 'No progression',
      multiple_progression = 'Multiple sites of progression',
      repeat_lrt = 'Repeat LRT'
    ),
    digits = list(all_categorical() ~ 1)
  ) %>%
  add_overall() %>%
  add_p() %>%
  bold_labels()

progression.tbl

```

Characteristic	Overall, N = 191	FALSE, N = 109	TRUE, N = 82	p-value
Local tumor progression	43.0 (22.5%)	19.0 (17.4%)	24.0 (29.3%)	0.053
Intrahepatic progression outside the ablation zone	121.0 (63.4%)	62.0 (56.9%)	59.0 (72.0%)	0.032
Local and outside intrahepatic progression	29.0 (15.2%)	11.0 (10.1%)	18.0 (22.0%)	0.024
Any kind of intrahepatic progression	135.0 (70.7%)	70.0 (64.2%)	65.0 (79.3%)	0.024
Extrahepatic progression	143.0 (74.9%)	72.0 (66.1%)	71.0 (86.6%)	0.001
No progression	26.0 (13.6%)	20.0 (18.3%)	6.0 (7.3%)	0.028
Multiple sites of progression	115.0 (60.2%)	54.0 (49.5%)	61.0 (74.4%)	<0.001
Repeat LRT	71.0 (37.2%)	48.0 (44.0%)	23.0 (28.0%)	0.024

## 4 Overall survival

```
model.format <- forest_model_format_options(point_size = 3)
```

### 4.1 By progression pattern

*Not adjusted*

```
os.multivariable <- coxph(Surv(os_m, deceased) ~ age_at_procedure +  
  ltp + ihp_outside + repeat_lrt + ehp,  
  data = data)  
  
forest_model(model = os.multivariable,  
  exponentiate = TRUE,  
  exclude_infinite_cis = TRUE,  
  format_options = model.format)
```

Variable	N	Hazard ratio		p
age_at_procedure	191		1.01 (0.99, 1.03)	0.24
ltp	191		1.87 (1.14, 3.09)	0.01
ihp_outside	191		3.54 (2.01, 6.22)	<0.001
repeat_lrt	191		0.28 (0.16, 0.48)	<0.001
ehp	191		1.72 (0.87, 3.39)	0.12

```
os.multivariable %>% tbl_regression(  
  exponentiate = TRUE,  
  show_single_row = all_categorical()  
)
```

Characteristic	HR	95% CI	p-value
age_at_procedure	1.01	0.99, 1.03	0.2
ltp	1.87	1.14, 3.09	0.014
ihp_outside	3.54	2.01, 6.22	<0.001
repeat_lrt	0.28	0.16, 0.48	<0.001
ehp	1.72	0.87, 3.39	0.12

```

data$no_progression <- !data$any_progression

os.multivariable <- coxph(Surv(os_m, deceased) ~ age_at_procedure + prior_resection +
  max_diameter + nr_of_tumors + right_side + ehm +
  ltp + ihp_outside + repeat_lrt + ehp,
  data = data)

forest_model(model = os.multivariable,
  exponentiate = TRUE,
  exclude_infinite_cis = TRUE,
  format_options = model.format)

```

Variable	N	Hazard ratio	p
age_at_procedure	191	1.01 (0.99, 1.03)	0.409
prior_resection	191	0.74 (0.45, 1.22)	0.240
max_diameter	191	1.40 (1.07, 1.85)	0.015
nr_of_tumors	191	1.79 (1.18, 2.71)	0.006
right_side	191	1.58 (0.90, 2.77)	0.111
ehm	191	0.94 (0.59, 1.51)	0.804
ltp	191	1.48 (0.85, 2.58)	0.170
ihp_outside	191	3.44 (1.90, 6.24)	<0.001
repeat_lrt	191	0.20 (0.10, 0.40)	<0.001
ehp	191	1.70 (0.86, 3.37)	0.126

0.1 0.2 0.5 1 2 5

```

os.multi.tbl <- os.multivariable %>% tbl_regression(
  label = list(age_at_procedure ~ "Age",
    prior_resection ~ "Prior resection",
    max_diameter ~ "Diameter of largest CLM",
    nr_of_tumors ~ "Number of ablated CLM",
    right_side ~ "Right side primary",
    ehm ~ "Presence of extrahepatic metastasis",

```

```

    ltp ~ "Progression of ablated tumors",
    ihp_outside ~ "Development of new intrahepatic tumors",
    repeat_lrt ~ "Salvage LRT for intrahepatic progression",
    ehpf ~ "Extrahepatic progression"),
  exponentiate = TRUE,
  show_single_row = all_categorical(),
  pvalue_fun = function(x) style_pvalue(x, digits = 2),
  estimate_fun = function(x) style_ratio(x, digits = 2)
)
os.multi.tbl

```

Characteristic	HR	95% CI	p-value
Age	1.01	0.99, 1.03	0.41
Prior resection	0.74	0.45, 1.22	0.24
Diameter of largest CLM	1.40	1.07, 1.85	0.015
Number of ablated CLM	1.79	1.18, 2.71	0.006
Right side primary	1.58	0.90, 2.77	0.11
Presence of extrahepatic metastasis	0.94	0.59, 1.51	0.80
Progression of ablated tumors	1.48	0.85, 2.58	0.17
Development of new intrahepatic tumors	3.44	1.90, 6.24	<0.001
Salvage LRT for intrahepatic progression	0.20	0.10, 0.40	<0.001
Extrahepatic progression	1.70	0.86, 3.37	0.13

```

os.multi.tbl %>% as_gt() %>% gtsave(filename = "out/os.rtf")

```

```

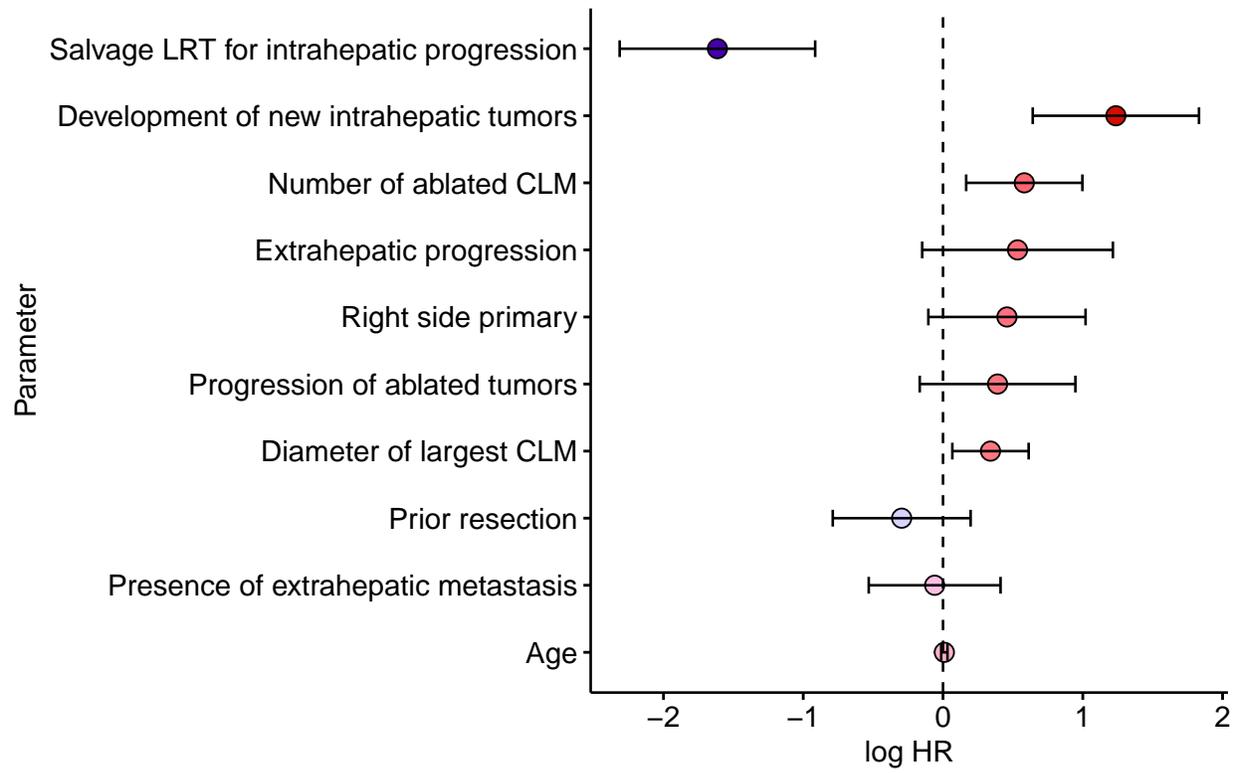
source("R/my_forest_plot.R")

```

```

os.multi.tbl$table_body %>%
  select(var_label, estimate, conf.low, conf.high) %>%
  mutate(estimate = log(estimate),
         conf.low = log(conf.low),
         conf.high = log(conf.high)) %>%
  arrange(abs(estimate)) %>%
  ggdotplot(., x = 'var_label', y = 'estimate',
            fill = "estimate",
            orientation = 'horizontal',
            xlab = "Parameter", ylab = "log HR") +
  scale_fill_gsea() +
  guides(fill="none") +
  geom_errorbar(aes(group = var_label, ymax = conf.high, ymin = conf.low),
               position = position_dodge(width = 0.8), width = 0.25) +
  geom_hline(yintercept = 0, linetype = "dashed")

```



## 5 Intrahepatic progression anywhere

```
ihpfs.any <- coxph(Surv(ihpfs_any, ihp_any) ~ age_at_procedure + prior_resection + a0_ablation +
  max_diameter + nr_of_tumors + ehm + right_side +
  p53 + ras + wnt + tgfb + pi3k + notch + cell,
  data = data)

forest_model(model = ihpfs.any,
  exponentiate = TRUE,
  format_options = model.format)
```

Variable	N	Hazard ratio		p
age_at_procedure	101		1.00 (0.97, 1.02)	0.935
prior_resection	101		1.09 (0.66, 1.80)	0.740
a0_ablation	101		0.42 (0.22, 0.82)	0.010
max_diameter	101		1.20 (0.91, 1.57)	0.192
nr_of_tumors	101		1.42 (0.96, 2.09)	0.077
ehm	101		0.75 (0.43, 1.31)	0.316
right_side	101		0.90 (0.48, 1.66)	0.725
p53	101		0.97 (0.54, 1.75)	0.923
ras	101		0.68 (0.38, 1.20)	0.179
wnt	101		0.56 (0.33, 0.94)	0.030
tgfb	101		2.74 (1.42, 5.27)	0.003
pi3k	101		1.00 (0.51, 1.98)	0.996
notch	101		0.63 (0.17, 2.29)	0.482
cell	101		2.53 (0.78, 8.20)	0.121

```
ihpfs.any.tbl <- ihpfs.any %>% tbl_regression(
  label = list(age_at_procedure ~ "Age",
    prior_resection ~ "Prior resection",
    a0_ablation ~ "A0 ablation",
    max_diameter ~ "Diameter of largest CLM",
    nr_of_tumors ~ "Number of ablated CLM",
    right_side ~ "Right side primary",
    ehm ~ "Presence of extrahepatic metastasis",
    p53 ~ "p53",
    ras ~ "RTK-RAS",
    wnt ~ "Wnt",
    tgfb ~ "TGFb",
    pi3k ~ "PI3K",
    notch ~ "Notch",
    cell ~ "Cell"),
```

```

exponentiate = TRUE,
show_single_row = all_categorical(),
pvalue_fun = function(x) style_pvalue(x, digits = 2),
estimate_fun = function(x) style_ratio(x, digits = 2)
)
ihpfs.any.tbl

```

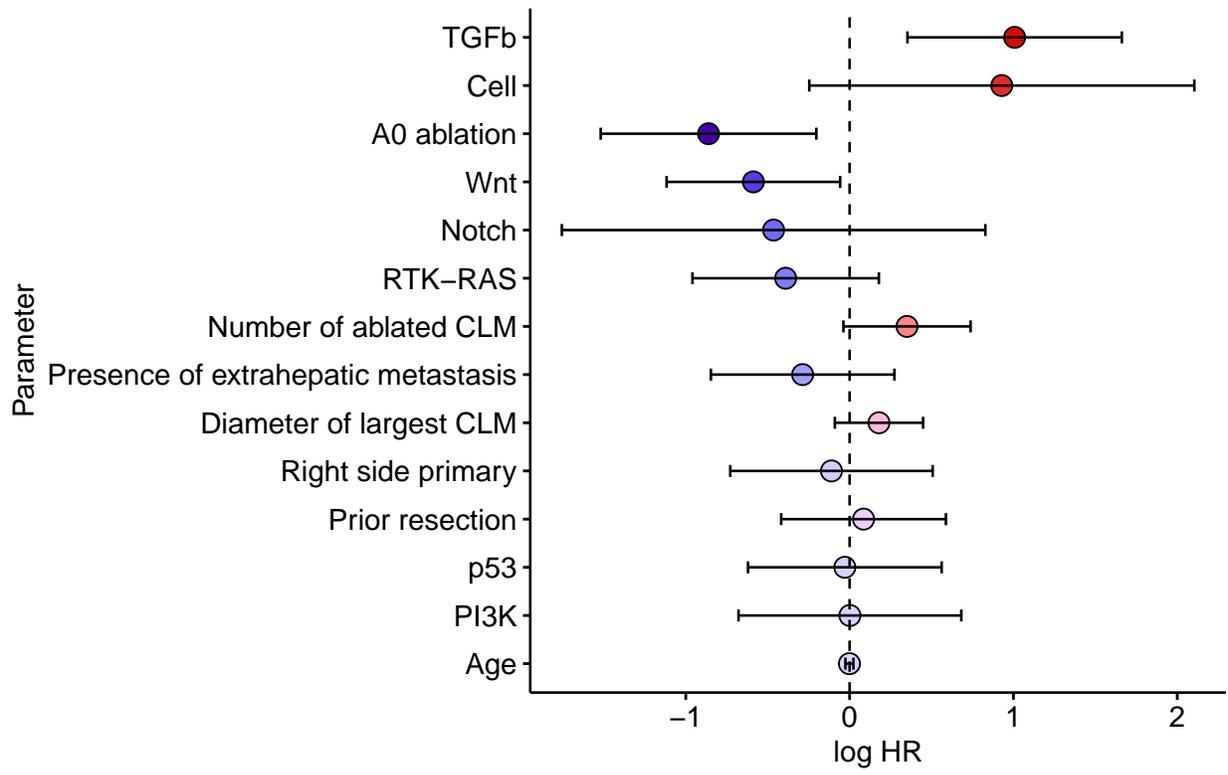
Characteristic	HR	95% CI	p-value
Age	1.00	0.97, 1.02	0.94
Prior resection	1.09	0.66, 1.80	0.74
A0 ablation	0.42	0.22, 0.82	0.010
Diameter of largest CLM	1.20	0.91, 1.57	0.19
Number of ablated CLM	1.42	0.96, 2.09	0.077
Presence of extrahepatic metastasis	0.75	0.43, 1.31	0.32
Right side primary	0.90	0.48, 1.66	0.73
p53	0.97	0.54, 1.75	0.92
RTK-RAS	0.68	0.38, 1.20	0.18
Wnt	0.56	0.33, 0.94	0.030
TGFb	2.74	1.42, 5.27	0.003
PI3K	1.00	0.51, 1.98	>0.99
Notch	0.63	0.17, 2.29	0.48
Cell	2.53	0.78, 8.20	0.12

```
ihpfs.any.tbl %>% as_gt() %>% gtsave(filename = "out/ihpfs_any.rtf")
```

```

ihpfs.any.tbl$table_body %>%
  select(var_label, estimate, conf.low, conf.high) %>%
  mutate(estimate = log(estimate),
         conf.low = log(conf.low),
         conf.high = log(conf.high)) %>%
  arrange(abs(estimate)) %>%
  ggdotplot(., x = 'var_label', y = 'estimate',
            fill = "estimate",
            orientation = 'horizontal',
            xlab = "Parameter", ylab = "log HR") +
  scale_fill_gsea() +
  guides(fill="none") +
  geom_errorbar(aes(group = var_label, ymax = conf.high, ymin = conf.low),
               position = position_dodge(width = 0.8), width = 0.25) +
  geom_hline(yintercept = 0, linetype = "dashed")

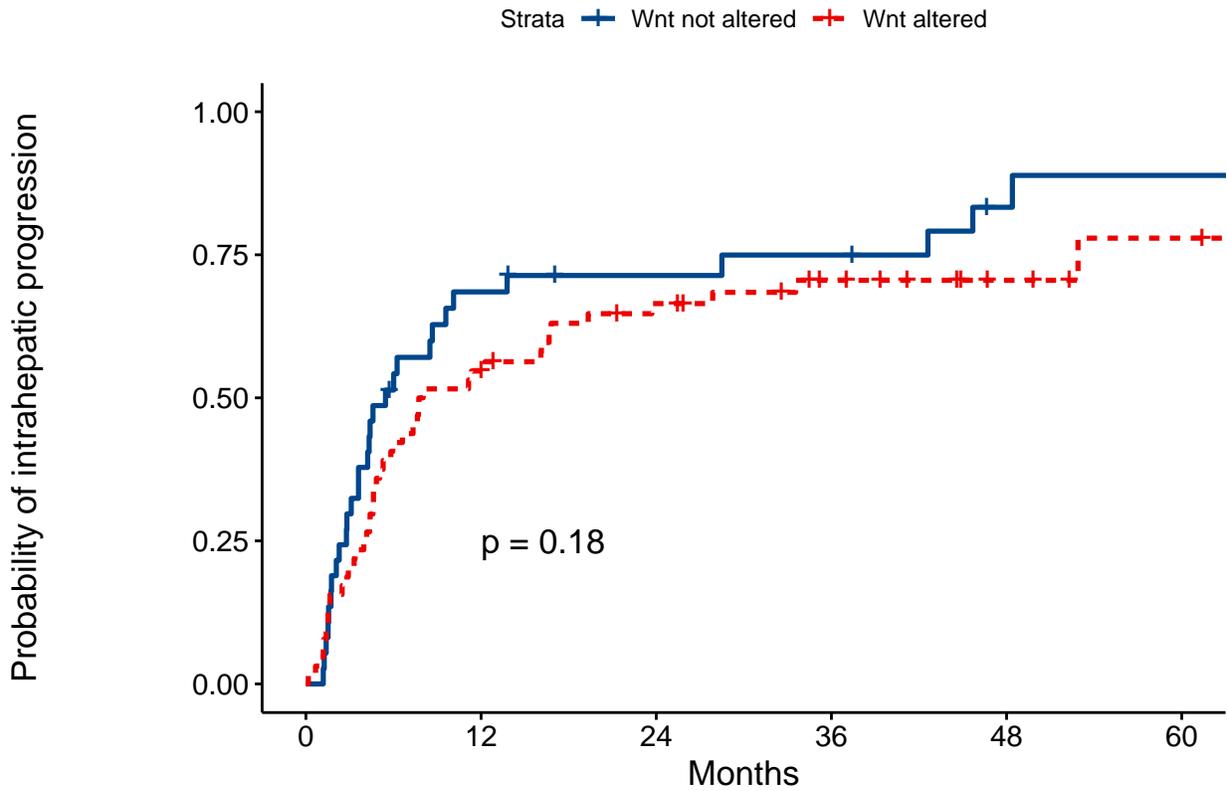
```



```

ggsurvplot(survfit(Surv(ihpfs_any / 30.25, ihp_any) ~ wnt,
                  data = data[data$mutation.analysis,]),
           pval = TRUE, risk.table = TRUE, conf.int = FALSE,
           pval.coord = c(12, 0.25),
           data = data[data$mutation.analysis,],
           legend.labs = c("Wnt not altered", "Wnt altered"),
           xlim = c(0, 60), break.time.by = 12, xlab = "Months",
           ylim = c(0, 1),
           ylab = "Probability of intrahepatic progression",
           palette = "lancet", fun = "event",
           linetype = c("strata"))

```



Number at risk

Strata	0	12	24	36	48	60
Wnt not altered	37	11	8	7	3	2
Wnt altered	64	29	19	12	6	3

Months

## 5.1 Intrahepatic progression outside the ablation zone

```
data.ihp <- data %>% filter(ihp_any)
```

```
ihpfs.out <- coxph(Surv(ihpfs_outside, ihp_outside) ~ age_at_procedure + prior_resection +
  max_diameter + nr_of_tumors + ehm + a0_ablation + right_side +
  p53 + ras + wnt + tgfb + pi3k + notch + cell,
  data = data)
```

```
forest_model(model = ihpfs.out,
  exponentiate = TRUE,
  format_options = model.format)
```

Variable	N	Hazard ratio		p
age_at_procedure	101		1.00 (0.97, 1.02)	0.863
prior_resection	101		1.55 (0.90, 2.69)	0.116
max_diameter	101		1.24 (0.93, 1.66)	0.148
nr_of_tumors	101		1.77 (1.20, 2.63)	0.004
ehm	101		0.78 (0.43, 1.42)	0.416
a0_ablation	101		0.61 (0.31, 1.21)	0.154
right_side	101		1.24 (0.65, 2.39)	0.514
p53	101		0.73 (0.39, 1.37)	0.330
ras	101		0.65 (0.35, 1.20)	0.165
wnt	101		0.60 (0.35, 1.03)	0.062
tgfb	101		2.75 (1.39, 5.45)	0.004
pi3k	101		0.82 (0.37, 1.80)	0.620
notch	101		1.10 (0.30, 4.11)	0.882
cell	101		2.12 (0.65, 6.92)	0.212

```
ihpfs.tbl <- ihpfs.out %>% tbl_regression(
  label = list(age_at_procedure ~ "Age",
    prior_resection ~ "Prior resection",
    a0_ablation ~ "A0 ablation",
    max_diameter ~ "Diameter of largest CLM",
    nr_of_tumors ~ "Number of ablated CLM",
    right_side ~ "Right side primary",
    ehm ~ "Presence of extrahepatic metastasis",
    p53 ~ "p53",
    ras ~ "RTK-RAS",
    wnt ~ "Wnt",
    tgfb ~ "TGFb",
    pi3k ~ "PI3K",
```

```

    notch ~ "Notch",
    cell ~ "Cell"),
  exponentiate = TRUE,
  show_single_row = all_categorical(),
  pvalue_fun = function(x) style_pvalue(x, digits = 2),
  estimate_fun = function(x) style_ratio(x, digits = 2)
)
ihpfs.tbl

```

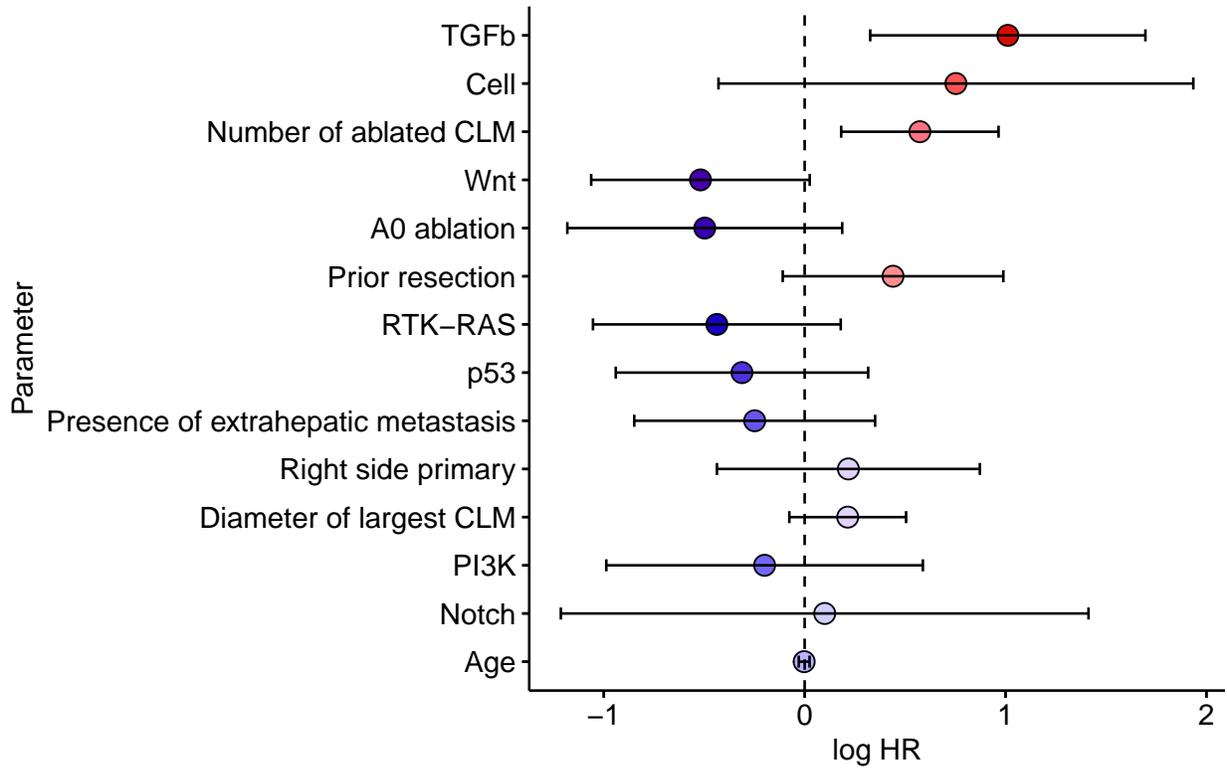
Characteristic	HR	95% CI	p-value
Age	1.00	0.97, 1.02	0.86
Prior resection	1.55	0.90, 2.69	0.12
Diameter of largest CLM	1.24	0.93, 1.66	0.15
Number of ablated CLM	1.77	1.20, 2.63	0.004
Presence of extrahepatic metastasis	0.78	0.43, 1.42	0.42
A0 ablation	0.61	0.31, 1.21	0.15
Right side primary	1.24	0.65, 2.39	0.51
p53	0.73	0.39, 1.37	0.33
RTK-RAS	0.65	0.35, 1.20	0.16
Wnt	0.60	0.35, 1.03	0.062
TGFb	2.75	1.39, 5.45	0.004
PI3K	0.82	0.37, 1.80	0.62
Notch	1.10	0.30, 4.11	0.88
Cell	2.12	0.65, 6.92	0.21

```
ihpfs.tbl %>% as_gt() %>% gtsave(filename = "out/ihpfs.rtf")
```

```

ihpfs.tbl$table_body %>%
  select(var_label, estimate, conf.low, conf.high) %>%
  mutate(estimate = log(estimate),
         conf.low = log(conf.low),
         conf.high = log(conf.high)) %>%
  arrange(abs(estimate)) %>%
  ggdotplot(., x = 'var_label', y = 'estimate',
            fill = "estimate",
            orientation = 'horizontal',
            xlab = "Parameter", ylab = "log HR") +
  scale_fill_gsea() +
  guides(fill="none") +
  geom_errorbar(aes(group = var_label, ymax = conf.high, ymin = conf.low),
               position = position_dodge(width = 0.8), width = 0.25) +
  geom_hline(yintercept = 0, linetype = "dashed")

```

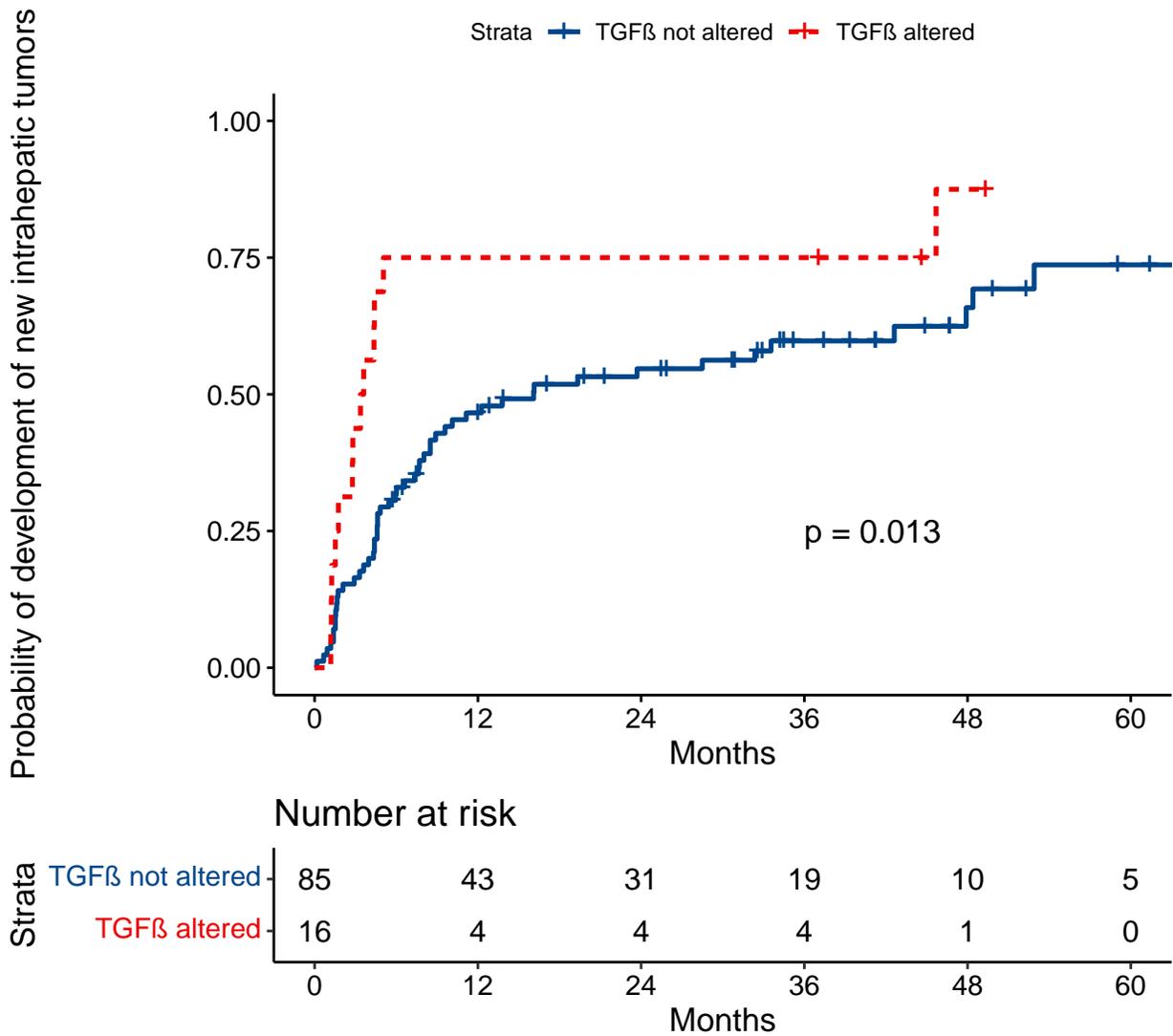


### 5.1.1 Survival curve based on TGFb

```

ggsurvplot(survfit(Surv(ihpfs_outside / 30.25, ihp_outside) ~ tgfb,
                  data = data[data$mutation.analysis, ]),
           pval = TRUE, risk.table = TRUE, conf.int = FALSE,
           pval.coord = c(36, 0.25),
           data = data[data$mutation.analysis, ],
           legend.labs = c("TGF\u03B2 not altered", "TGF\u03B2 altered"),
           xlim = c(0, 60), break.time.by = 12, xlab = "Months",
           ylim = c(0, 1),
           ylab = "Probability of development of new intrahepatic tumors",
           palette = "lancet", fun = "event",
           linetype = c("strata"))

```



```
ihpfs.tgfb <- survfit(Surv(ihpfs_outside / 30.25, ihp_outside) ~ tgfb,  
                    data = data[data$mutation.analysis,])  
print(ihpfs.tgfb)
```

```
## Call: survfit(formula = Surv(ihpfs_outside/30.25, ihp_outside) ~ tgfb,  
##      data = data[data$mutation.analysis, ])  
##  
##           n events median 0.95LCL 0.95UCL  
## tgfb=FALSE 85     52  16.10    8.50   47.9  
## tgfb=TRUE  16     13   3.49    1.75    NA
```

## 5.2 Intrahepatic progression at the ablation zone

*Notch excluded because it leads to infinite CIs*

```
ltp.multi <- coxph(Surv(ltpfs, ltp) ~ age_at_procedure + prior_resection + right_side +
  max_diameter + nr_of_tumors + ehm + a0_ablation +
  p53 + ras + wnt + tgfb + pi3k + cell,
  data = data)

forest_model(model = ltp.multi,
  exponentiate = TRUE,
  format_options = model.format)
```

Variable	N	Hazard ratio		p
age_at_procedure	101		0.97 (0.94, 1.01)	0.20
prior_resection	101		0.36 (0.16, 0.83)	0.02
right_side	101		0.82 (0.28, 2.34)	0.71
max_diameter	101		1.59 (1.04, 2.42)	0.03
nr_of_tumors	101		1.04 (0.55, 1.98)	0.90
ehm	101		0.56 (0.25, 1.28)	0.17
a0_ablation	101		0.17 (0.05, 0.66)	0.01
p53	101		1.42 (0.50, 4.05)	0.51
ras	101		0.75 (0.32, 1.77)	0.52
wnt	101		0.83 (0.36, 1.93)	0.66
tgfb	101		1.46 (0.47, 4.54)	0.51
pi3k	101		1.03 (0.36, 2.99)	0.95
cell	101		1.04 (0.12, 9.25)	0.97

0.05 0.1 0.2 0.5 1 2 5

```
ltp.tbl <- ltp.multi %>% tbl_regression(
  label = list(age_at_procedure ~ "Age",
    prior_resection ~ "Prior resection",
    a0_ablation ~ "A0 ablation",
    max_diameter ~ "Diameter of largest CLM",
    nr_of_tumors ~ "Number of ablated CLM",
    right_side ~ "Right side primary",
    ehm ~ "Presence of extrahepatic metastasis",
    p53 ~ "p53",
    ras ~ "RTK-RAS",
    wnt ~ "Wnt",
    tgfb ~ "TGFb",
    pi3k ~ "PI3K",
    cell ~ "Cell"),
```

```

exponentiate = TRUE,
show_single_row = all_categorical(),
pvalue_fun = function(x) style_pvalue(x, digits = 2),
estimate_fun = function(x) style_ratio(x, digits = 2)
)

ltp.tbl

```

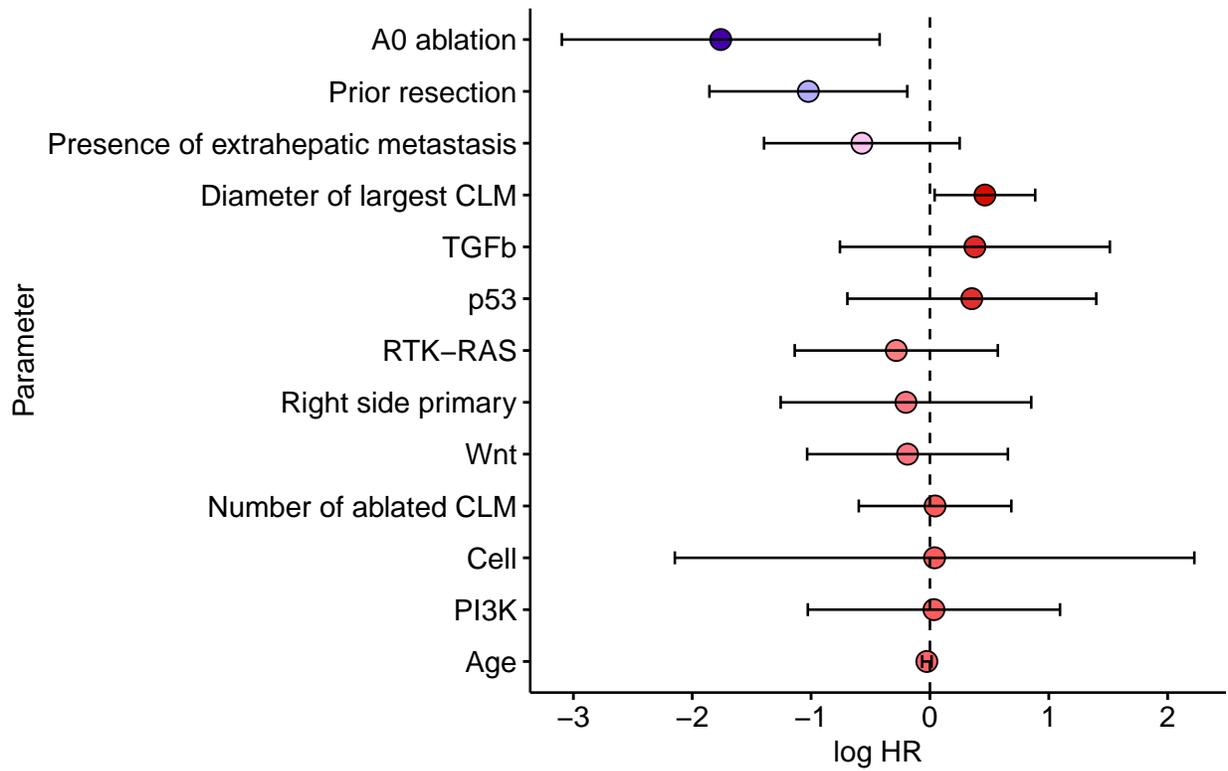
Characteristic	HR	95% CI	p-value
Age	0.97	0.94, 1.01	0.20
Prior resection	0.36	0.16, 0.83	0.016
Right side primary	0.82	0.28, 2.34	0.71
Diameter of largest CLM	1.59	1.04, 2.42	0.032
Number of ablated CLM	1.04	0.55, 1.98	0.90
Presence of extrahepatic metastasis	0.56	0.25, 1.28	0.17
A0 ablation	0.17	0.05, 0.66	0.010
p53	1.42	0.50, 4.05	0.51
RTK-RAS	0.75	0.32, 1.77	0.52
Wnt	0.83	0.36, 1.93	0.66
TGFb	1.46	0.47, 4.54	0.51
PI3K	1.03	0.36, 2.99	0.95
Cell	1.04	0.12, 9.25	0.97

```
ltp.tbl %>% as_gt() %>% gtsave("out/ltp.rtf")
```

```

ltp.tbl$table_body %>%
  select(var_label, estimate, conf.low, conf.high) %>%
  mutate(estimate = log(estimate),
         conf.low = log(conf.low),
         conf.high = log(conf.high)) %>%
  arrange(abs(estimate)) %>%
  ggdotplot(., x = 'var_label', y = 'estimate',
            fill = "estimate",
            orientation = 'horizontal',
            xlab = "Parameter", ylab = "log HR") +
  scale_fill_gsea() +
  guides(fill="none") +
  geom_errorbar(aes(group = var_label, ymax = conf.high, ymin = conf.low),
               position = position_dodge(width = 0.8), width = 0.25) +
  geom_hline(yintercept = 0, linetype = "dashed")

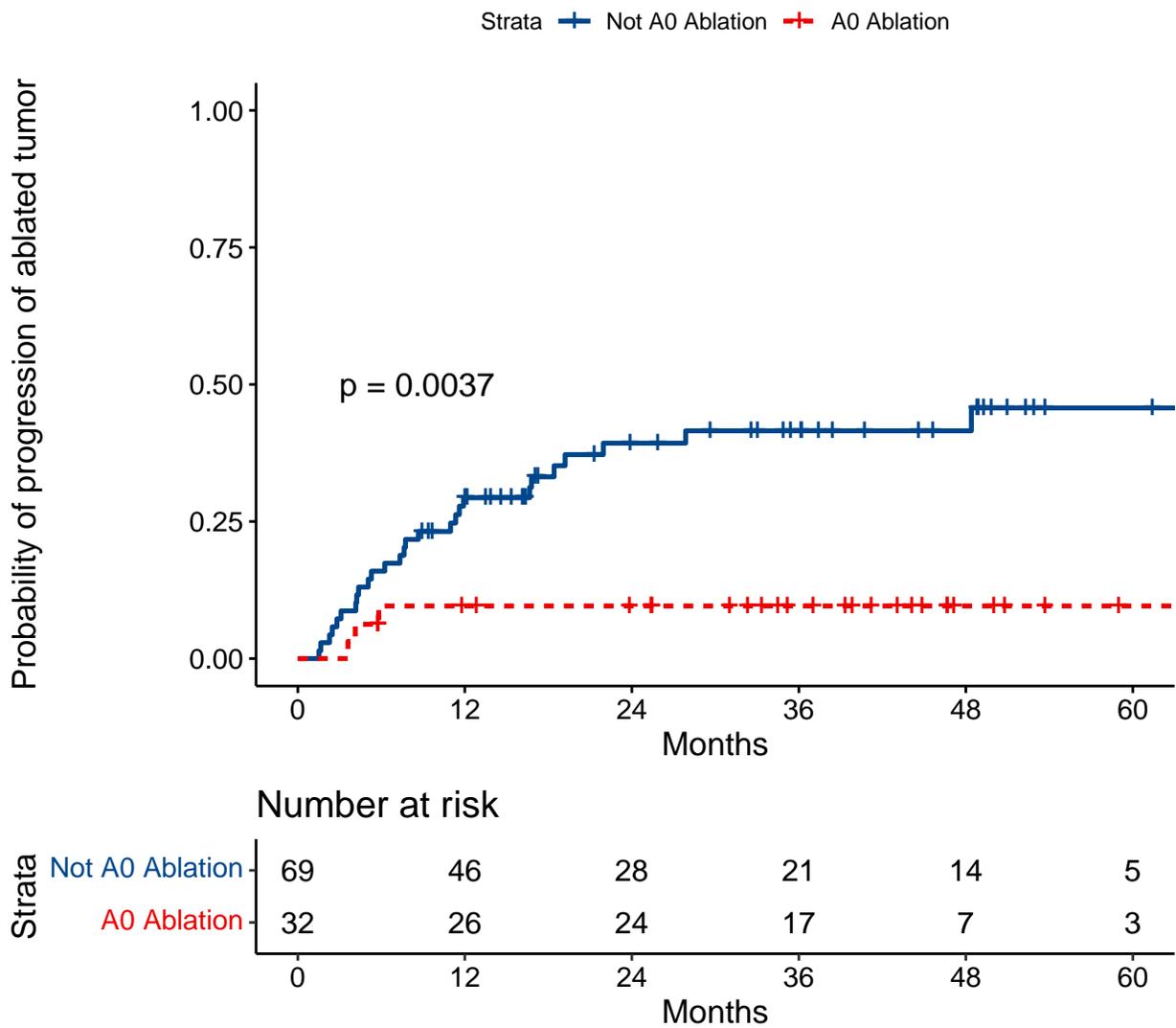
```



```

ggsurvplot(survfit(Surv(ltpfs, ltp) ~ a0_ablation,
                  data = data[data$mutation.analysis,]),
           pval = TRUE, risk.table = TRUE, conf.int = FALSE,
           pval.coord = c(3, 0.5),
           data = data[data$mutation.analysis,],
           legend.labs = c("Not A0 Ablation", "A0 Ablation"),
           xlim = c(0, 60), ylim = c(0, 1), break.time.by = 12, xlab = "Months",
           ylab = "Probability of progression of ablated tumor",
           palette = "lancet", fun = "event",
           linetype = c("strata"))

```



## 6 Salvage LRT

```
m.repeat.lrt <- glm(repeat_lrt ~ age_at_procedure + prior_resection +
                    ehm + right_side +
                    p53 + ras + wnt + tgfb + pi3k + notch + cell,
                    data = data.ihp,
                    family = 'binomial')

forest_model(model = m.repeat.lrt,
             exponentiate = TRUE,
             format_options = model.format)
```

Variable	N	Odds ratio		p
age_at_procedure	75		1.03 (0.98, 1.09)	0.256
prior_resection	75		2.71 (0.86, 9.35)	0.097
ehm	75		0.67 (0.21, 2.10)	0.496
right_side	75		1.02 (0.26, 4.13)	0.975
p53	75		0.48 (0.12, 1.71)	0.274
ras	75		0.38 (0.11, 1.20)	0.106
wnt	75		5.80 (1.94, 19.52)	0.003
tgfb	75		0.62 (0.13, 2.83)	0.531
pi3k	75		1.54 (0.31, 8.90)	0.605
notch	75		0.17 (0.01, 2.18)	0.186
cell	75		0.49 (0.04, 5.86)	0.554

0.01 0.05 0.1 0.5 1 5 10

```
repeat.lrt.tbl <- m.repeat.lrt %>% tbl_regression(
  label = list(age_at_procedure ~ "Age",
               prior_resection ~ "Prior resection",
               right_side ~ "Right side primary",
               ehm ~ "Presence of extrahepatic metastasis",
               p53 ~ "p53",
               ras ~ "RTK-RAS",
               wnt ~ "Wnt",
               tgfb ~ "TGFb",
               pi3k ~ "PI3K",
               notch ~ "Notch",
               cell ~ "Cell"),
  exponentiate = TRUE,
  show_single_row = all_categorical(),
```

```

    pvalue_fun = function(x) style_pvalue(x, digits = 2),
    estimate_fun = function(x) style_ratio(x, digits = 2)
  )
repeat.lrt.tbl

```

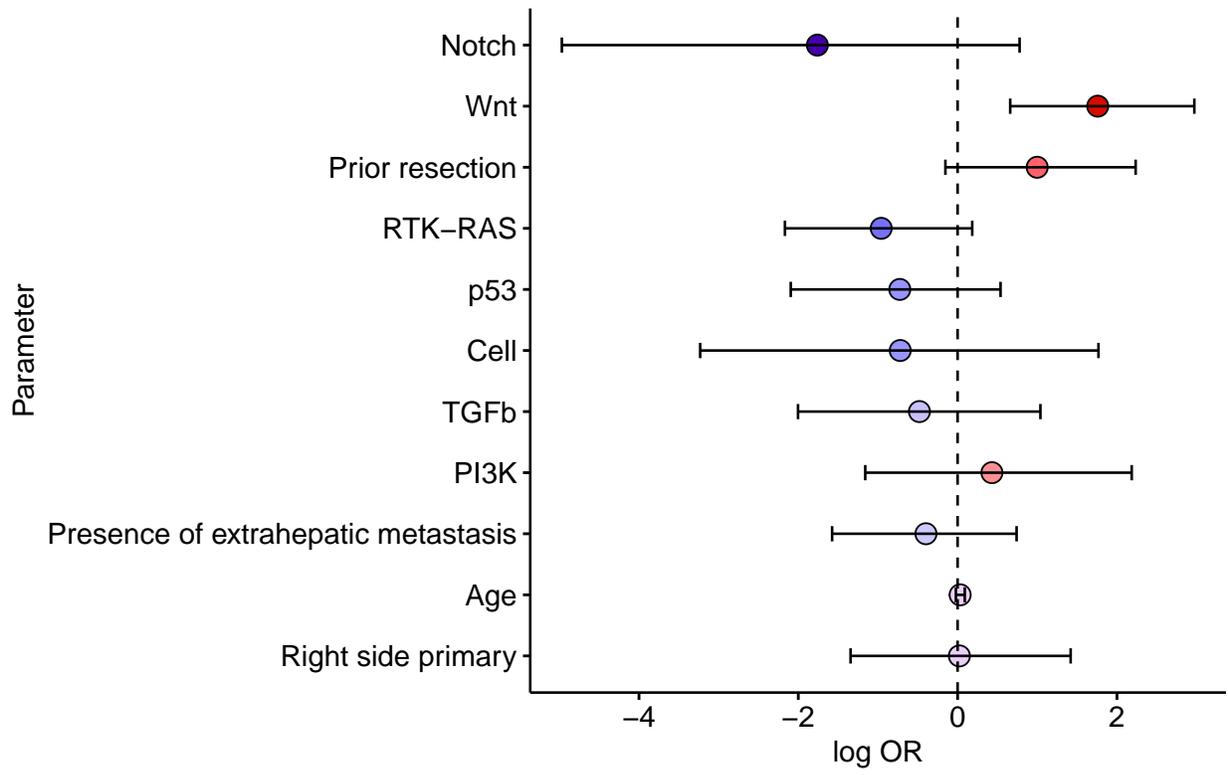
Characteristic	OR	95% CI	p-value
Age	1.03	0.98, 1.09	0.26
Prior resection	2.71	0.86, 9.35	0.10
Presence of extrahepatic metastasis	0.67	0.21, 2.10	0.50
Right side primary	1.02	0.26, 4.13	0.97
p53	0.48	0.12, 1.71	0.27
RTK-RAS	0.38	0.11, 1.20	0.11
Wnt	5.80	1.94, 19.5	0.003
TGFb	0.62	0.13, 2.83	0.53
PI3K	1.54	0.31, 8.90	0.61
Notch	0.17	0.01, 2.18	0.19
Cell	0.49	0.04, 5.86	0.55

```
repeat.lrt.tbl %>% as_gt() %>% gtsave("out/repeat.lrt.rtf")
```

```

repeat.lrt.tbl$table_body %>%
  select(var_label, estimate, conf.low, conf.high) %>%
  mutate(estimate = log(estimate),
         conf.low = log(conf.low),
         conf.high = log(conf.high)) %>%
  arrange(abs(estimate)) %>%
  ggdotplot(., x = 'var_label', y = 'estimate',
            fill = "estimate",
            orientation = 'horizontal',
            xlab = "Parameter", ylab = "log OR") +
  scale_fill_gsea() +
  guides(fill="none") +
  geom_errorbar(aes(group = var_label, ymax = conf.high, ymin = conf.low),
               position = position_dodge(width = 0.8), width = 0.25) +
  geom_hline(yintercept = 0, linetype = "dashed")

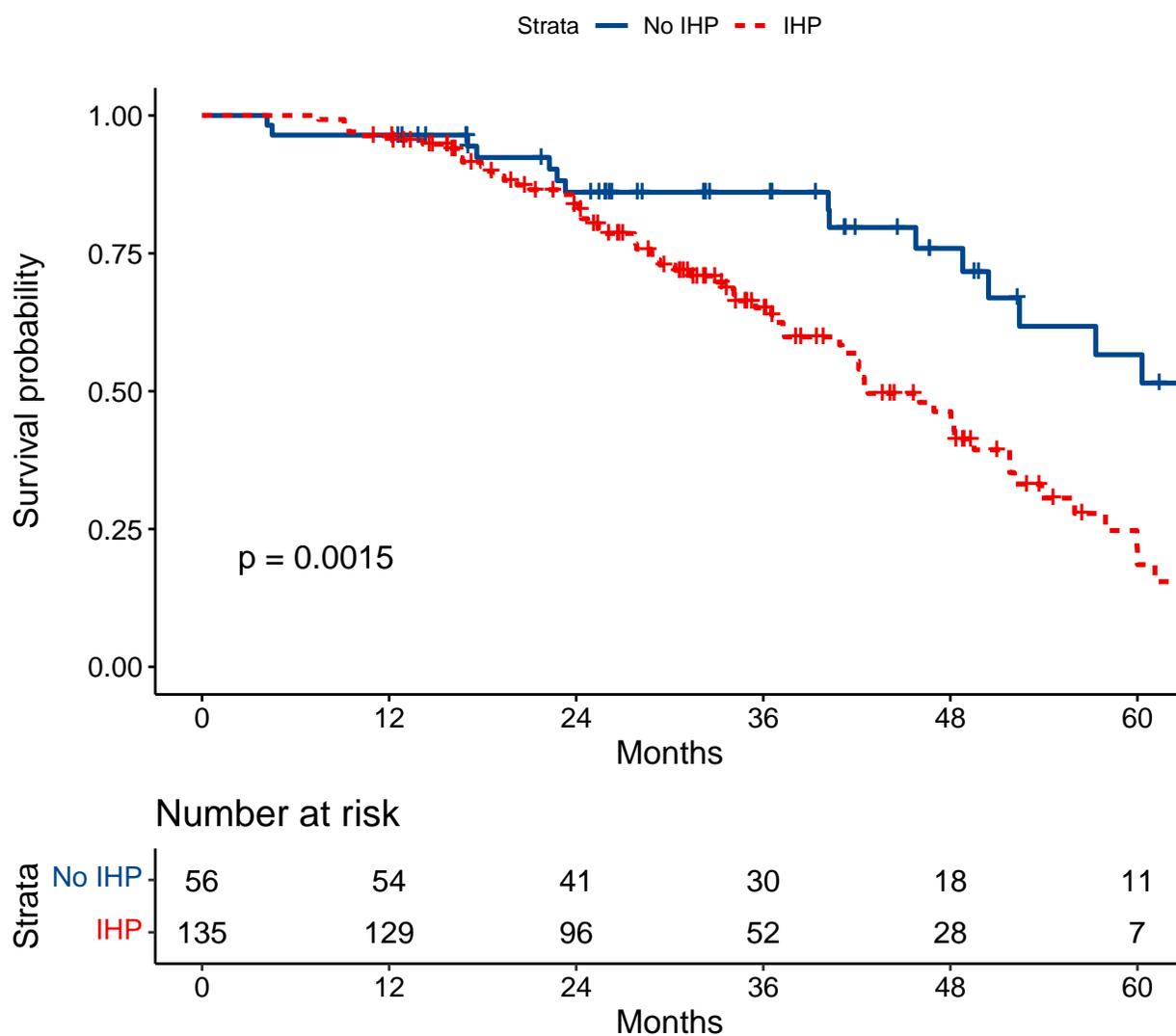
```



## 6.1 Survival following IHP

```
os.plot <- ggsurvplot(survfit(Surv(os_m, deceased) ~ ihp_any, data = data),
  pval = TRUE, risk.table = TRUE, conf.int = FALSE,
  data = data,
  legend.labs = c("No IHP", "IHP"),
  xlim = c(0, 60), break.time.by = 12, xlab = "Months",
  palette = "lancet",
  linetype = c("strata"))+
  guides(color = guide_legend(override.aes = list(shape = NA)))
```

```
os.plot
```



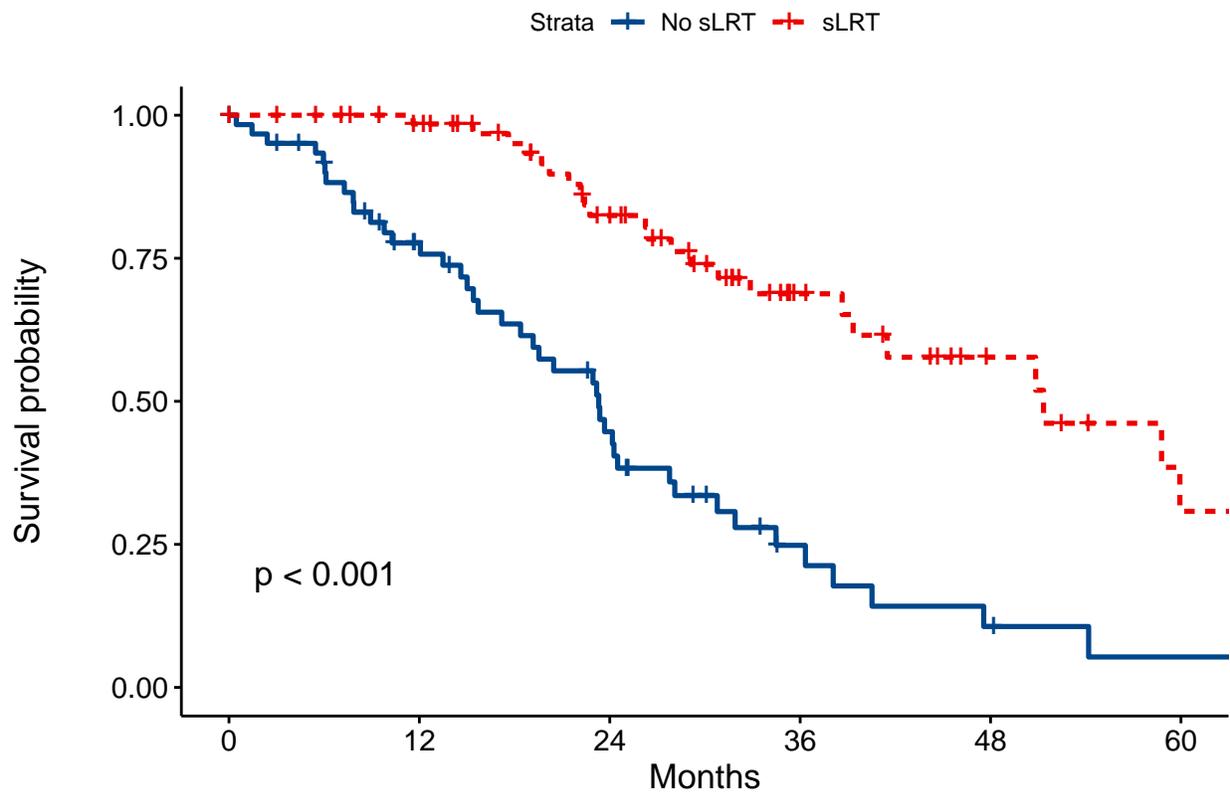
```
os_m.surv <- survfit(Surv(os_m, deceased) ~ ihp_any, data = data)
print(os_m.surv)
```

```
## Call: survfit(formula = Surv(os_m, deceased) ~ ihp_any, data = data)
##
##           n events median 0.95LCL 0.95UCL
## ihp_any=FALSE  56     17   70.3    52.4     NA
## ihp_any=TRUE  135     65   42.7    40.9    51.8
```

## 6.2 Survival after IHP - salvage LRT vs. no salvage LRT

```
os2.plot <- ggsurvplot(survfit(Surv(os_ihp_m, deceased) ~ repeat_lrt, data = data.ihp),
  pval = "p < 0.001", risk.table = TRUE, conf.int = FALSE,
  data = data.ihp,
  legend.labs = c("No sLRT", "sLRT"),
  xlim = c(0, 60), break.time.by = 12, xlab = "Months",
  palette = "lancet",
  linetype = c("strata"))
```

os2.plot



Number at risk

Strata	0	12	24	36	48	60
No sLRT	64	40	21	7	3	1
sLRT	71	63	44	20	10	4

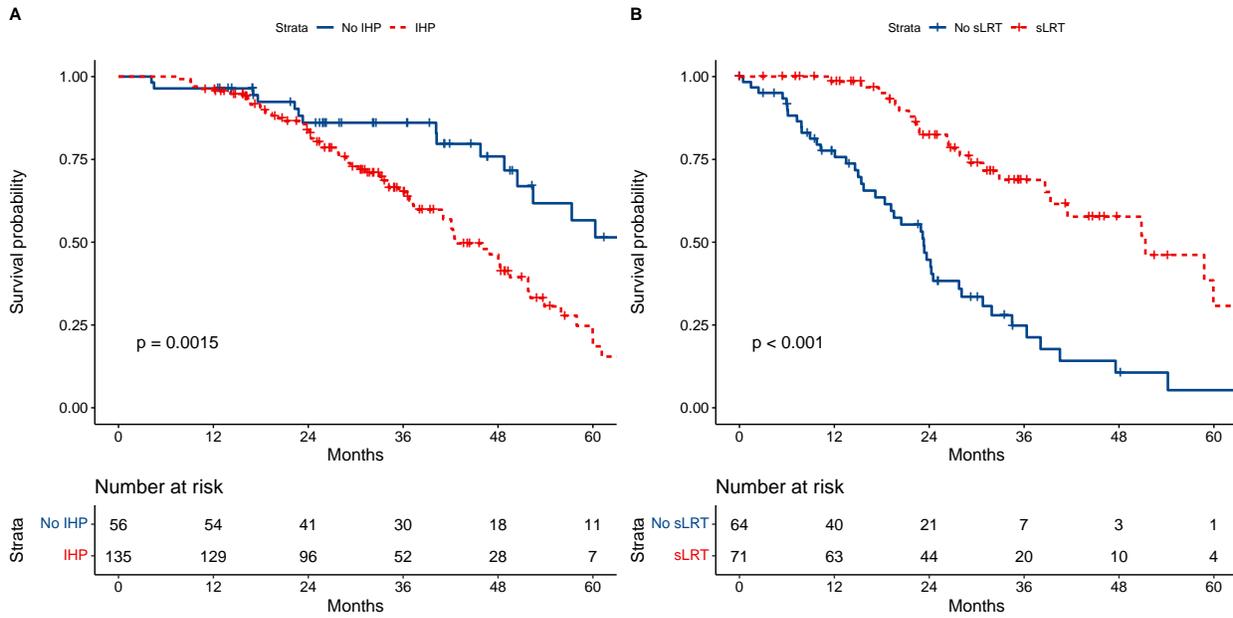
Months

```
os2_m.surv <- survfit(Surv(os_ihp_m, deceased) ~ repeat_lrt, data = data.ihp)
print(os2_m.surv)
```

```
## Call: survfit(formula = Surv(os_ihp_m, deceased) ~ repeat_lrt, data = data.ihp)
##
##              n events median 0.95LCL 0.95UCL
## repeat_lrt=FALSE 64     42  23.3   18.4   28.1
## repeat_lrt=TRUE  71     23  51.3   39.3   NA
```

## 6.2.1 Combined plot

```
ggarrange(os.plot$plot, os2.plot$plot, os.plot$table, os2.plot$table,
  labels = c("A", "B"),
  ncol = 2, nrow = 2, heights = c(2.5, 0.8),
  align = "v")
```



## 7 Frequency of mutations

```
data.grouped <- data %>% filter(mutation.analysis) %>%
  group_by(race) %>%
  summarise(
    n = n(),
    n_tot = n(),
    p53 = sum(p53),
    Wnt = sum(wnt),
    "RTK-RAS" = sum(ras),
    PI3K = sum(pi3k),
    TGFb = sum(tgfb),
    Notch = sum(notch),
    "Cell Cycle" = sum(cell),
  )
data.grouped
```

race	n	n_tot	p53	Wnt	RTK-RAS	PI3K	TGFb	Notch	Cell Cycle
af	10	10	8	6	6	4	1	0	0
as	5	5	4	2	2	1	0	0	0
cauc	77	77	57	51	35	10	13	4	3
his	6	6	6	3	2	1	1	1	1
NA	3	3	0	2	3	1	1	0	0

```
data.grouped.wide <- gather(data.grouped, key = type_of_mutation, nr_mutated,
  p53,Wnt,"RTK-RAS",PI3K,TGFb,Notch,"Cell Cycle",
  factor_key=TRUE, na.rm = TRUE)

data.grouped.wide <- within(data.grouped.wide, {
  nr_mutated_perc <- (nr_mutated / n) * 100.0

  race <- ifelse(race == "his", "Hi", race)
  race <- ifelse(race == "cauc", "Ca", race)
  race <- ifelse(race == "as", "As", race)
  race <- ifelse(race == "af", "Af", race)
  race <- factor(race, levels = c("Ca", "Hi", "Af", "As"))
})
```

### 7.1 Number of mutations per pathway with predominant gene

```
data.mut <- data %>% filter(mutation.analysis)

data.plot <- data.frame(pathway = c("p53", "p53", "Wnt", "Wnt", "RTK-RAS", "RTK-RAS",
  "PI3K", "PI3K", "TGFb", "TGFb", "Notch", "Notch",
  "Cell Cycle", "Cell Cycle"),
  mutation = c("p53", "TP53", "Wnt", "APC", "RTK-RAS", "RAS/BRAF",
  "PI3K", "PIK3CA", "TGFb", "SMAD4", "Notch", "FBXW7",
  "Cell Cycle", "RB1"),
```

```

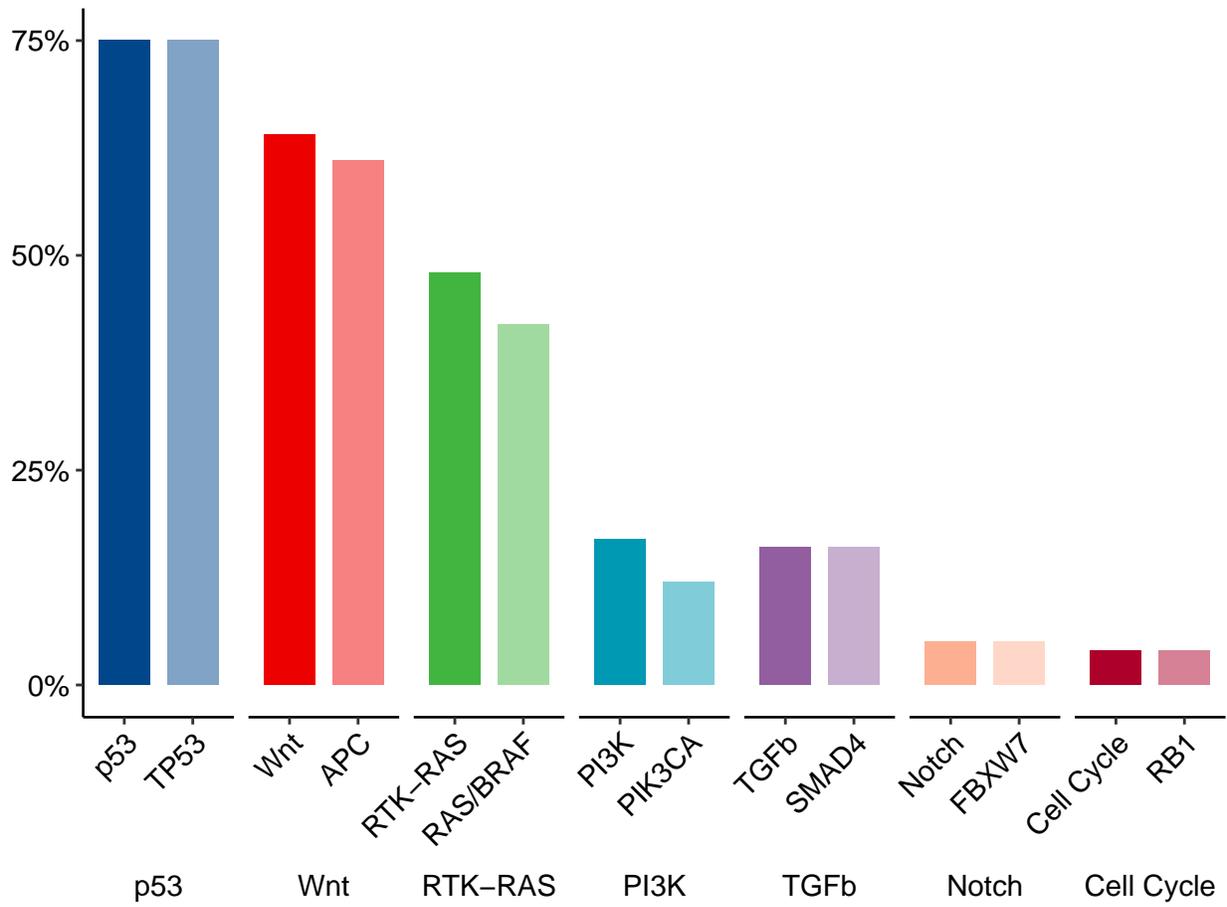
        nr_mutated = c(sum(data.mut$p53), sum(data.mut$tp53),
                      sum(data.mut$wnt), sum(data.mut$apc),
                      sum(data.mut$ras), sum(data.mut$rasbraf),
                      sum(data.mut$pi3k), sum(data.mut$pik3ca),
                      sum(data.mut$tgfb), sum(data.mut$smad4),
                      sum(data.mut$notch), sum(data.mut$fbxw7),
                      sum(data.mut$cell), sum(data.mut$rb1))) %>%
mutate(nr_mutated_perc = nr_mutated / sum(data.mut$mutation.analysis))

data.plot$mutation <- factor(data.plot$mutation,
                            levels = c("p53", "TP53", "Wnt", "APC", "RTK-RAS", "RAS/BRAF",
                                         "PI3K", "PIK3CA", "TGFb", "SMAD4", "Notch", "FBXW7",
                                         "Cell Cycle", "RB1"))

data.plot$pathway <- factor(data.plot$pathway,
                           levels = c("p53", "Wnt", "RTK-RAS",
                                       "PI3K", "TGFb", "Notch", "Cell Cycle"))

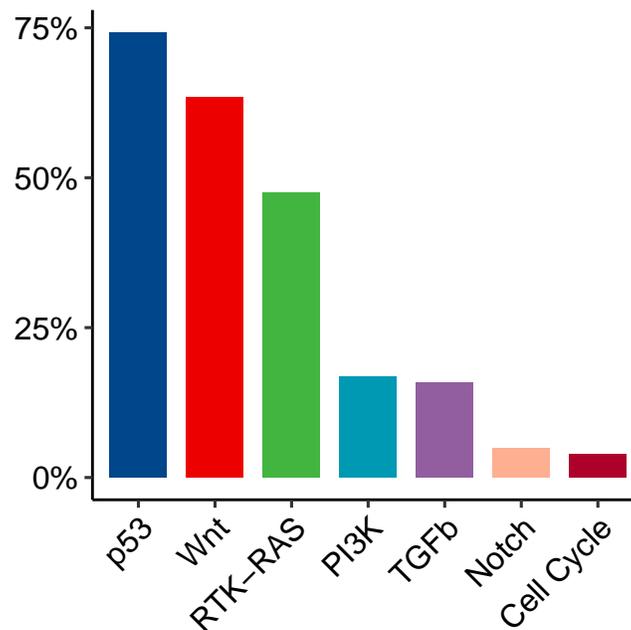
ggplot(data = data.plot, aes(x = mutation, y = nr_mutated,
                            fill = mutation)) +
  geom_bar(stat="identity", width = 0.75, position = position_dodge(width=1)) +
  facet_grid(~pathway, scales = "free_x", space = "free_x", switch = "x") +
  scale_fill_manual(values = rbind(pal_lancet()(7), pal_lancet(alpha = 0.5)(7))) +
  scale_y_continuous(breaks = seq(0, 100, 25), labels = paste(seq(0, 100, 25), "%", sep="")) +
  theme_pubr() +
  theme(strip.placement = "outside",
        axis.text.x = element_text(size = 12, angle = 45, hjust = 1),
        axis.title = element_blank()) +
  theme(strip.text.x = element_text(size = 12, color = "black"),
        strip.background = element_blank()) +
  theme(legend.position = "None")

```



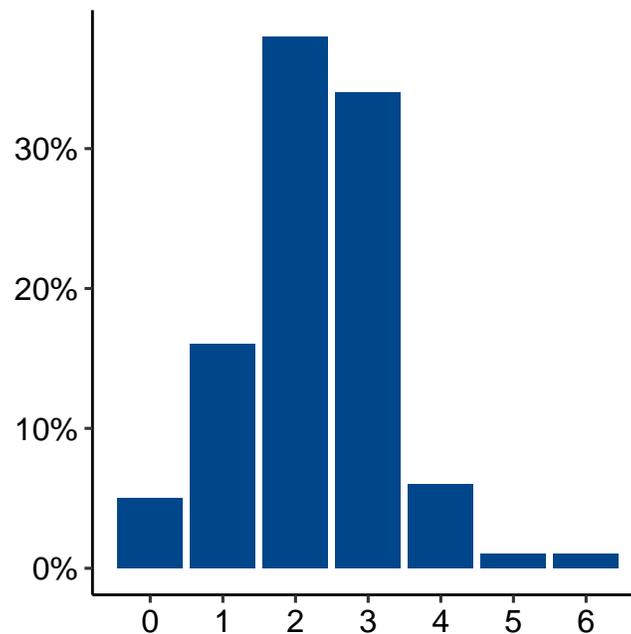
## 7.2 Number of mutations per pathway

```
data.plot <- data.grouped.wide %>%
  group_by(type_of_mutation) %>%
  summarise(
    n = n(),
    n_tot = sum(n_tot),
    nr_mutated = sum(nr_mutated)
  )
data.plot$nr_mutated_perc <- (data.plot$nr_mutated / data.plot$n_tot) * 100
ggplot(data = data.plot, aes(x = type_of_mutation, y = nr_mutated_perc,
                             fill = type_of_mutation)) +
  geom_bar(stat="identity", width = 0.75, position = position_dodge(width=1)) +
  scale_fill_lancet() +
  scale_y_continuous(breaks = seq(0, 100, 25), labels = paste(seq(0, 100, 25), "%", sep="")) +
  theme_pubr() +
  theme(strip.placement = "outside",
        axis.text.x = element_text(size = 12, angle = 45, hjust = 1),
        axis.title = element_blank()) +
  theme(strip.text.x = element_text(size = 12, color = "black"),
        strip.background = element_blank()) +
  theme(legend.position = "None")
```



### 7.3 Frequency of mutations

```
ggplot(data = data, aes(x = num_altered_paths, fill = "num_altered_paths")) +  
  geom_histogram(stat="count") +  
  scale_fill_lancet() +  
  scale_y_continuous(breaks = seq(0, 50, 10), labels = paste(seq(0, 50, 10), "%", sep="")) +  
  scale_x_continuous(breaks = seq(0, 6, 1)) +  
  xlab("Nr of altered pathways") +  
  theme_pubr() +  
  theme(strip.placement = "outside",  
        axis.text.x = element_text(size = 12),  
        axis.title = element_blank()) +  
  theme(strip.text.x = element_text(size = 12, color = "black"),  
        strip.background = element_blank()) +  
  theme(legend.position = "None")
```



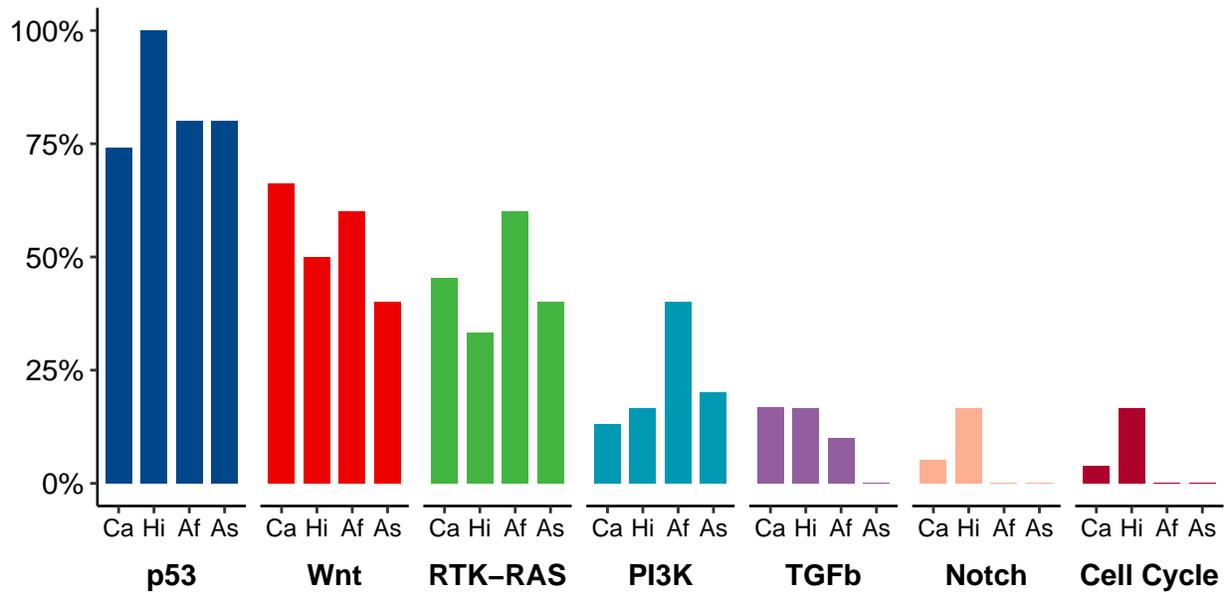
### 7.4 Mutation per race/ethnicity

```
data.grouped.wide <- data.grouped.wide %>% filter(!is.na(race))  
data.grouped.wide$race <- as.factor(data.grouped.wide$race)  
  
ggplot(data = data.grouped.wide, aes(x = race, y = nr_mutated_perc,  
                                     fill = type_of_mutation)) +  
  geom_bar(stat="identity", width = 0.75, position = position_dodge(width=1)) +  
  facet_grid(~type_of_mutation, scales = "free_x", space = "free_x", switch = "x") +  
  scale_fill_lancet() +  
  scale_y_continuous(breaks = seq(0, 100, 25), labels = paste(seq(0, 100, 25), "%", sep="")) +  
  theme_pubr() +
```

```

theme(strip.placement = "outside",
      axis.text.x = element_text(size = 10),
      axis.title = element_blank() +
theme(strip.text.x = element_text( size = 12, color = "black", face = "bold"),
      strip.background = element_blank()) +
theme(legend.position = "None")

```



## 8 Citation

```
citation('ggplot2')
```

```
##  
## To cite ggplot2 in publications, please use:  
##  
## H. Wickham. ggplot2: Elegant Graphics for Data Analysis.  
## Springer-Verlag New York, 2016.  
##  
## A BibTeX entry for LaTeX users is  
##  
## @Book{,  
##   author = {Hadley Wickham},  
##   title = {ggplot2: Elegant Graphics for Data Analysis},  
##   publisher = {Springer-Verlag New York},  
##   year = {2016},  
##   isbn = {978-3-319-24277-4},  
##   url = {https://ggplot2.tidyverse.org},  
## }
```

```
citation('ggpubr')
```

```
##  
## To cite package 'ggpubr' in publications use:  
##  
## Alboukadel Kassambara (2020). ggpubr: 'ggplot2' Based Publication  
## Ready Plots. R package version 0.4.0.  
## https://CRAN.R-project.org/package=ggpubr  
##  
## A BibTeX entry for LaTeX users is  
##  
## @Manual{,  
##   title = {ggpubr: 'ggplot2' Based Publication Ready Plots},  
##   author = {Alboukadel Kassambara},  
##   year = {2020},  
##   note = {R package version 0.4.0},  
##   url = {https://CRAN.R-project.org/package=ggpubr},  
## }
```

```
citation('tidyverse')
```

```
##  
## Wickham et al., (2019). Welcome to the tidyverse. Journal of Open  
## Source Software, 4(43), 1686, https://doi.org/10.21105/joss.01686  
##  
## A BibTeX entry for LaTeX users is  
##  
## @Article{,  
##   title = {Welcome to the {tidyverse}},  
##   author = {Hadley Wickham and Mara Averick and Jennifer Bryan and Winston Chang and Lucy D'Agostini
```

```
##   year = {2019},
##   journal = {Journal of Open Source Software},
##   volume = {4},
##   number = {43},
##   pages = {1686},
##   doi = {10.21105/joss.01686},
## }
```

```
citation()
```

```
##
## To cite R in publications use:
##
##   R Core Team (2021). R: A language and environment for statistical
##   computing. R Foundation for Statistical Computing, Vienna, Austria.
##   URL https://www.R-project.org/.
##
## A BibTeX entry for LaTeX users is
##
##   @Manual{,
##     title = {R: A Language and Environment for Statistical Computing},
##     author = {{R Core Team}},
##     organization = {R Foundation for Statistical Computing},
##     address = {Vienna, Austria},
##     year = {2021},
##     url = {https://www.R-project.org/},
##   }
##
## We have invested a lot of time and effort in creating R, please cite it
## when using it for data analysis. See also 'citation("pkgname")' for
## citing R packages.
```