

Code ▾

Rectal IRST_Molinari et al Final Analysis 05302024

```
library(swimplot) library(grid) library(gtable) library(readr) library(mosaic) library(dplyr) library(survival)
library(survminer) library(ggplot2) library(scales) library(coxphf) library(ggthemes) library(tidyverse)
library(gtsummary) library(flextable) library(parameters) library(car) library(ComplexHeatmap) library(tidyverse)
library(readxl) library(survival) library(janitor) library(DT) library(pROC)
```

#ctDNA positivity by stage and window

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```
#Number of Pts at Baseline - percentage positivity by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

total_base <- sum(!is.na(circ_data$ctDNA.Baseline))
print(total_base)
```

```
[1] 33
```

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```
circ_data$ctDNA.Baseline <- as.factor(circ_data$ctDNA.Baseline)
cont_table_base <- table(circ_data$cStage, circ_data$ctDNA.Baseline)
print(cont_table_base)
```

	NEGATIVE	POSITIVE	
	3	0	0
I	0	0	1
II	0	0	6
III	0	3	20

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```
#Number of Pts post NAT - percentage positivity by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

total_postnat <- sum(!is.na(circ_data$ctDNA.postNAC))
print(total_postnat)
```

```
[1] 33
```

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```
circ_data$ctDNA.postNAC <- as.factor(circ_data$ctDNA.postNAC)
cont_table_postnat <- table(circ_data$cStage, circ_data$ctDNA.postNAC)
print(cont_table_postnat)
```

		NEGATIVE	POSITIVE
	3	0	0
I	0	1	0
II	0	5	1
III	0	19	4

Hide

```
#Number of Pts post-surgery - percentage positivity by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

total_mrd <- sum(!is.na(circ_data$ctDNA.postop))
print(total_mrd)
```

```
[1] 33
```

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```
circ_data$ctDNA.postop <- as.factor(circ_data$ctDNA.postop)
cont_table_mrd <- table(circ_data$cStage, circ_data$ctDNA.postop)
print(cont_table_mrd)
```

		NEGATIVE	POSITIVE
	3	0	0
I	0	1	0
II	0	5	1
III	0	21	2

```
#Radiological Recurrence rates of ctDNA positive by stage and window
```

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```
#Number of Pts at Baseline - percentage of radiological recurrence by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.Baseline=="POSITIVE",]
circ_datadf <- as.data.frame(circ_data)

total_base <- sum(!is.na(circ_data$RFS.Event))
print(total_base)
```

```
[1] 27
```

[Hide](#)

```
circ_data$RFS.Event <- as.factor(circ_data$RFS.Event)
cont_table_base <- table(circ_data$cStage, circ_data$RFS.Event)
print(cont_table_base)
```

	FALSE	TRUE
I	1	0
II	5	1
III	14	6

[Hide](#)

```
#Number of Pts post NAT - percentage of radiological recurrence by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.postNAC=="POSITIVE",]
circ_datadf <- as.data.frame(circ_data)

total_postnat <- sum(!is.na(circ_data$RFS.Event))
print(total_postnat)
```

```
[1] 5
```

[Hide](#)

```
circ_data$RFS.Event <- as.factor(circ_data$RFS.Event)
cont_table_postnat <- table(circ_data$cStage, circ_data$RFS.Event)
print(cont_table_postnat)
```

	TRUE
II	1
III	4

Hide

```
#Number of Pts post-surgery - percentage of radiological recurrence by stage
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.postop=="POSITIVE",]
circ_datadf <- as.data.frame(circ_data)

total_mrd <- sum(!is.na(circ_data$RFS.Event))
print(total_mrd)
```

```
[1] 3
```

Hide

```
circ_data$RFS.Event <- as.factor(circ_data$RFS.Event)
cont_table_mrd <- table(circ_data$cStage, circ_data$RFS.Event)
print(cont_table_mrd)
```

	TRUE
II	1
III	2

#Summary Table

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```

rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

circ_data_subset <- circ_data %>%
  select(
    Gender,
    Age,
    Height,
    Weight,
    cStage,
    cStage.TNM,
    NAT.Cycles,
    NAT.Regimen,
    Downstaging,
    postNAT.TNM,
    NAR,
    Surgery.Type,
    pStage,
    pStage.TNM,
    pN,
    pCR,
    Grade,
    MSI.status,
    Path.Response,
    ACT,
    ACT.Regimen,
    RFS.Event,
    Rec.Site,
    OS.Event,
    RFS.months,
    Total.FU.months) %>%
  mutate(
    Gender = factor(Gender, levels = c("Male", "Female")),
    Age = as.numeric(Age),
    Height = as.numeric(Height),
    Weight = as.numeric(Weight),
    cStage = factor(cStage, levels = c("I", "II", "III")),
    cStage.TNM = factor(cStage.TNM, levels = c("T2N0M0", "T2N1M0", "T3/T4N1M0", "T3N0M0", "T3N2M0", "T4N2M0")),
    NAT.Cycles = as.numeric(NAT.Cycles),
    NAT.Regimen = factor(NAT.Regimen, levels = c("FOLFOX", "Capecitabine")),
    Downstaging = factor(Downstaging, levels = c("NR", "R"), labels = c("Non Responder", "Responders")),
    postNAT.TNM = factor(postNAT.TNM, levels = c("T0N0M0", "T1/T2N0M0", "T3N0M0", "T3/T4N1M0", "T1/T2N2M0", "T3N2M0")),
    NAR = as.numeric(NAR),
    Surgery.Type = factor(Surgery.Type, levels = c("RAR", "MILES")),
    pStage = factor(pStage, levels = c("0", "I", "II", "III")),
    pStage.TNM = factor(pStage.TNM, levels = c("T0N0M0", "T1/T2N0M0", "T3N0M0", "T2N1M0", "T3/T4N1M0", "T3N2M0", "T4N2M0")),

```

```
pN = factor(pN, levels = c("0", "1", "2"), labels = c("N0", "N1", "N2")),
pCR = factor(pCR, levels = c("TRUE", "FALSE"), labels = c("Yes", "No")),
Grade = factor(Grade, levels = c("1", "2", "3"), labels = c("G1", "G2", "G3")),
MSI.status = factor(MSI.status, levels = c("MSS", "MSI"), labels = c("MSS", "MSI-High")),
Path.Response = factor(Path.Response, levels = c("Minimal Regression", "Moderate Regression", "Near Complete Regression", "Complete Regression")),
ACT = factor(ACT, levels = c("FALSE", "TRUE"), labels = c("No adjuvant", "Adjuvant")),
ACT.Regimen = factor(ACT.Regimen, levels = c("FOLFOX", "CAPOX", "Capecitabine")),
RFS.Event = factor(RFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence")),
Rec.Site = factor(Rec.Site, levels = c("Lung", "Lung, Liver", "Liver", "Lung, Liver, Bone", "Bone, Adrenal", "Pelvis")),
OS.Event = factor(OS.Event, levels = c("FALSE", "TRUE"), labels = c("Alive", "Deceased")),
RFS.months = as.numeric(RFS.months),
Total.FU.months = as.numeric(Total.FU.months))
table1 <- circ_data_subset %>%
tbl_summary(
  statistic = list(
    all_continuous() ~ "{median} ({min} - {max})",
    all_categorical() ~ "{n} ({p}%)") %>%
bold_labels()
table1
```

Characteristic	N = 30 ¹
Gender	
Male	21 (70%)
Female	9 (30%)
Age	67 (38 - 80)
Height	173 (120 - 184)
Weight	72 (38 - 120)
cStage	
I	1 (3.3%)
II	6 (20%)
III	23 (77%)
cStage.TNM	
T2N0M0	1 (3.3%)
¹ n (%); Median (Range)	

Characteristic	N = 30 ¹
T2N1M0	3 (10%)
T3/T4N1M0	13 (43%)
T3N0M0	6 (20%)
T3N2M0	4 (13%)
T4N2M0	3 (10%)
NAT.Cycles	
3	1 (3.3%)
4	12 (40%)
5	4 (13%)
6	12 (40%)
7	1 (3.3%)
NAT.Regimen	
FOLFOX	13 (43%)
Capecitabine	17 (57%)
Downstaging	
Non Responders	5 (17%)
Responders	25 (83%)
postNAT.TNM	
T0N0M0	1 (3.3%)
T1/T2N0M0	13 (43%)
T3N0M0	3 (10%)
T3/T4N1M0	11 (37%)
T1/T2N2M0	1 (3.3%)
T3N2M0	1 (3.3%)
NAR	
0	1 (3.3%)
¹ n (%); Median (Range)	

Characteristic	N = 30 ¹
0.94	3 (10%)
3.75	5 (17%)
8.43	6 (20%)
14.98	6 (20%)
20.4	3 (10%)
30.07	3 (10%)
50.36	3 (10%)
Surgery.Type	
RAR	26 (87%)
MILES	4 (13%)
pStage	
0	6 (20%)
I	10 (33%)
II	5 (17%)
III	9 (30%)
pStage.TNM	
T0N0M0	6 (20%)
T1/T2N0M0	10 (33%)
T3N0M0	5 (17%)
T2N1M0	1 (3.3%)
T3/T4N1M0	5 (17%)
T3N2M0	2 (6.7%)
T4N2M0	1 (3.3%)
pN	
N0	21 (70%)
N1	6 (20%)
¹ n (%); Median (Range)	

Characteristic	N = 30 ¹
N2	3 (10%)
pCR	6 (20%)
Grade	
G1	6 (30%)
G2	10 (50%)
G3	4 (20%)
Unknown	10
MSI.status	
MSS	8 (80%)
MSI-High	2 (20%)
Unknown	20
Path.Response	
Minimal Regression	4 (13%)
Moderate Regression	16 (53%)
Near Complete Regression	4 (13%)
Complete Regression	6 (20%)
ACT	
No adjuvant	8 (27%)
Adjuvant	22 (73%)
ACT.Regimen	
FOLFOX	12 (55%)
CAPOX	6 (27%)
Capecitabine	4 (18%)
Unknown	8
RFS.Event	
No Recurrence	22 (73%)
¹ n (%); Median (Range)	

Characteristic	N = 30 ¹
Recurrence	8 (27%)
Rec.Site	
Lung	2 (25%)
Lung, Liver	2 (25%)
Liver	1 (13%)
Lung, Liver, Bone	1 (13%)
Bone, Adrenal	1 (13%)
Pelvis	1 (13%)
Unknown	22
OS.Event	
Alive	20 (67%)
Deceased	10 (33%)
RFS.months	54 (4 - 112)
Total.FU.months	60 (10 - 118)
¹ n (%); Median (Range)	

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```
fit1 <- as_flex_table(  
  table1,  
  include = everything(),  
  return_calls = FALSE,  
  strip_md_bold = TRUE)  
fit1
```

Characteristic	N = 30 ¹
Gender	
Male	21 (70%)
Female	9 (30%)
Age	67 (38 - 80)
Height	173 (120 - 184)
Weight	72 (38 - 120)
¹ n (%); Median (Range)	

Characteristic	N = 30 ¹
cStage	
I	1 (3.3%)
II	6 (20%)
III	23 (77%)
cStage.TNM	
T2N0M0	1 (3.3%)
T2N1M0	3 (10%)
T3/T4N1M0	13 (43%)
T3N0M0	6 (20%)
T3N2M0	4 (13%)
T4N2M0	3 (10%)
NAT.Cycles	
3	1 (3.3%)
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Capecitabine	17 (57%)
Downstaging	
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Responders	25 (83%)
postNAT.TNM	
T0N0M0	1 (3.3%)
T1/T2N0M0	13 (43%)
T3N0M0	3 (10%)
T3/T4N1M0	11 (37%)
T1/T2N2M0	1 (3.3%)
T3N2M0	1 (3.3%)

¹n (%); Median (Range)

Characteristic	N = 30 ¹
NAR	
0	1 (3.3%)
0.94	3 (10%)
3.75	5 (17%)
8.43	6 (20%)
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pStage.TNM	
T0N0M0	6 (20%)
T1/T2N0M0	10 (33%)
T3N0M0	5 (17%)
T2N1M0	1 (3.3%)
T3/T4N1M0	5 (17%)
T3N2M0	2 (6.7%)
T4N2M0	1 (3.3%)
pN	
N0	21 (70%)
N1	6 (20%)
N2	3 (10%)
pCR	6 (20%)

¹n (%); Median (Range)

Characteristic	N = 30 ¹
Grade	
G1	6 (30%)
G2	10 (50%)
G3	4 (20%)
Unknown	10
MSI.status	
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MSI-High	2 (20%)
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Adjuvant	22 (73%)
ACT.Regimen	
FOLFOX	12 (55%)
CAPOX	6 (27%)
Capecitabine	4 (18%)
Unknown	8
RFS.Event	
No Recurrence	22 (73%)
Recurrence	8 (27%)
Rec.Site	
Lung	2 (25%)
Lung, Liver	2 (25%)
Liver	1 (13%)
Lung, Liver, Bone	1 (13%)

¹n (%); Median (Range)

Characteristic	N = 30 ¹
Bone, Adrenal	1 (13%)
Pelvis	1 (13%)
Unknown	22
OS.Event	
Alive	20 (67%)
Deceased	10 (33%)
RFS.months	54 (4 - 112)
Total.FU.months	60 (10 - 118)

¹n (%); Median (Range)

Hide

```
save_as_docx(fit1, path= "~/Downloads/table1.docx")
```

#Summary Table with IQR

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```

rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

circ_data_subset <- circ_data %>%
  select(
    Gender,
    Age,
    Height,
    Weight,
    cStage,
    cStage.TNM,
    NAT.Cycles,
    NAT.Regimen,
    Downstaging,
    postNAT.TNM,
    NAR,
    Surgery.Type,
    pStage,
    pStage.TNM,
    pN,
    pCR,
    Grade,
    MSI.status,
    Path.Response,
    ACT,
    ACT.Regimen,
    RFS.Event,
    Rec.Site,
    OS.Event,
    RFS.months,
    Total.FU.months) %>%
  mutate(
    Gender = factor(Gender, levels = c("Male", "Female")),
    Age = as.numeric(Age),
    Height = as.numeric(Height),
    Weight = as.numeric(Weight),
    cStage = factor(cStage, levels = c("I", "II", "III")),
    cStage.TNM = factor(cStage.TNM, levels = c("T2N0M0", "T2N1M0", "T3/T4N1M0", "T3N0M0", "T3N2M0", "T4N2M0")),
    NAT.Cycles = as.numeric(NAT.Cycles),
    NAT.Regimen = factor(NAT.Regimen, levels = c("FOLFOX", "Capecitabine")),
    Downstaging = factor(Downstaging, levels = c("NR", "R"), labels = c("Non Responder", "Responders")),
    postNAT.TNM = factor(postNAT.TNM, levels = c("T0N0M0", "T1/T2N0M0", "T3N0M0", "T3/T4N1M0", "T1/T2N2M0", "T3N2M0")),
    NAR = as.numeric(NAR),
    Surgery.Type = factor(Surgery.Type, levels = c("RAR", "MILES")),
    pStage = factor(pStage, levels = c("0", "I", "II", "III")),
    pStage.TNM = factor(pStage.TNM, levels = c("T0N0M0", "T1/T2N0M0", "T3N0M0", "T2N1M0", "T3/T4N1M0", "T3N2M0", "T4N2M0")),

```

```
pN = factor(pN, levels = c("0", "1", "2"), labels = c("N0", "N1", "N2")),
pCR = factor(pCR, levels = c("TRUE", "FALSE"), labels = c("Yes", "No")),
Grade = factor(Grade, levels = c("1", "2", "3"), labels = c("G1", "G2", "G3")),
MSI.status = factor(MSI.status, levels = c("MSS", "MSI"), labels = c("MSS", "MSI-High")),
Path.Response = factor(Path.Response, levels = c("Minimal Regression", "Moderate Regression", "Near Complete Regression", "Complete Regression")),
ACT = factor(ACT, levels = c("FALSE", "TRUE"), labels = c("No adjuvant", "Adjuvant")),
ACT.Regimen = factor(ACT.Regimen, levels = c("FOLFOX", "CAPOX", "Capecitabine")),
RFS.Event = factor(RFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence")),
Rec.Site = factor(Rec.Site, levels = c("Lung", "Lung, Liver", "Liver", "Lung, Liver, Bone", "Bone, Adrenal", "Pelvis")),
OS.Event = factor(OS.Event, levels = c("FALSE", "TRUE"), labels = c("Alive", "Deceased")),
RFS.months = as.numeric(RFS.months),
Total.FU.months = as.numeric(Total.FU.months))
table1 <- circ_data_subset %>%
tbl_summary(
  statistic = list(
    all_continuous() ~ "{median} ({p25}, {p75})",
    all_categorical() ~ "{n} ({p}%)") %>%
bold_labels()
table1
```

Characteristic	N = 30 [†]
Gender	
Male	21 (70%)
Female	9 (30%)
Age	67 (63, 76)
Height	173 (161, 178)
Weight	72 (59, 89)
cStage	
I	1 (3.3%)
II	6 (20%)
III	23 (77%)
cStage.TNM	
T2N0M0	1 (3.3%)
[†] n (%); Median (IQR)	

Characteristic	N = 30¹
T2N1M0	3 (10%)
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NAR	
0	1 (3.3%)
¹ n (%); Median (IQR)	

Characteristic	N = 30 ¹
0.94	3 (10%)
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pStage	
0	6 (20%)
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Bone, Adrenal	1 (13%)
Pelvis	1 (13%)
Unknown	22
OS.Event	
Alive	20 (67%)
Deceased	10 (33%)
RFS.months	54 (28, 80)
Total.FU.months	60 (47, 91)
¹ n (%); Median (IQR)	

Hide

```
fit1 <- as_flex_table(  
  table1,  
  include = everything(),  
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  strip_md_bold = TRUE)  
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```

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Characteristic	N = 30 ¹
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T3N2M0	1 (3.3%)

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III	9 (30%)
pStage.TNM	
T0N0M0	6 (20%)
T1/T2N0M0	10 (33%)
T3N0M0	5 (17%)
T2N1M0	1 (3.3%)
T3/T4N1M0	5 (17%)
T3N2M0	2 (6.7%)
T4N2M0	1 (3.3%)
pN	
N0	21 (70%)
N1	6 (20%)
N2	3 (10%)
pCR	6 (20%)

¹n (%); Median (IQR)

Characteristic	N = 30 ¹
Grade	
G1	6 (30%)
G2	10 (50%)
G3	4 (20%)
Unknown	10
MSI.status	
MSS	8 (80%)
MSI-High	2 (20%)
Unknown	20
Path.Response	
Minimal Regression	4 (13%)
Moderate Regression	16 (53%)
Near Complete Regression	4 (13%)
Complete Regression	6 (20%)
ACT	
No adjuvant	8 (27%)
Adjuvant	22 (73%)
ACT.Regimen	
FOLFOX	12 (55%)
CAPOX	6 (27%)
Capecitabine	4 (18%)
Unknown	8
RFS.Event	
No Recurrence	22 (73%)
Recurrence	8 (27%)
Rec.Site	
Lung	2 (25%)
Lung, Liver	2 (25%)
Liver	1 (13%)
Lung, Liver, Bone	1 (13%)

¹n (%); Median (IQR)

Characteristic	N = 30 ¹
Bone, Adrenal	1 (13%)
Pelvis	1 (13%)
Unknown	22
OS.Event	
Alive	20 (67%)
Deceased	10 (33%)
RFS.months	54 (28, 80)
Total.FU.months	60 (47, 91)

¹n (%); Median (IQR)

Hide

```
save_as_docx(fit1, path= "~/Downloads/table1.docx")
```

#DFS in Complete Cohort

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

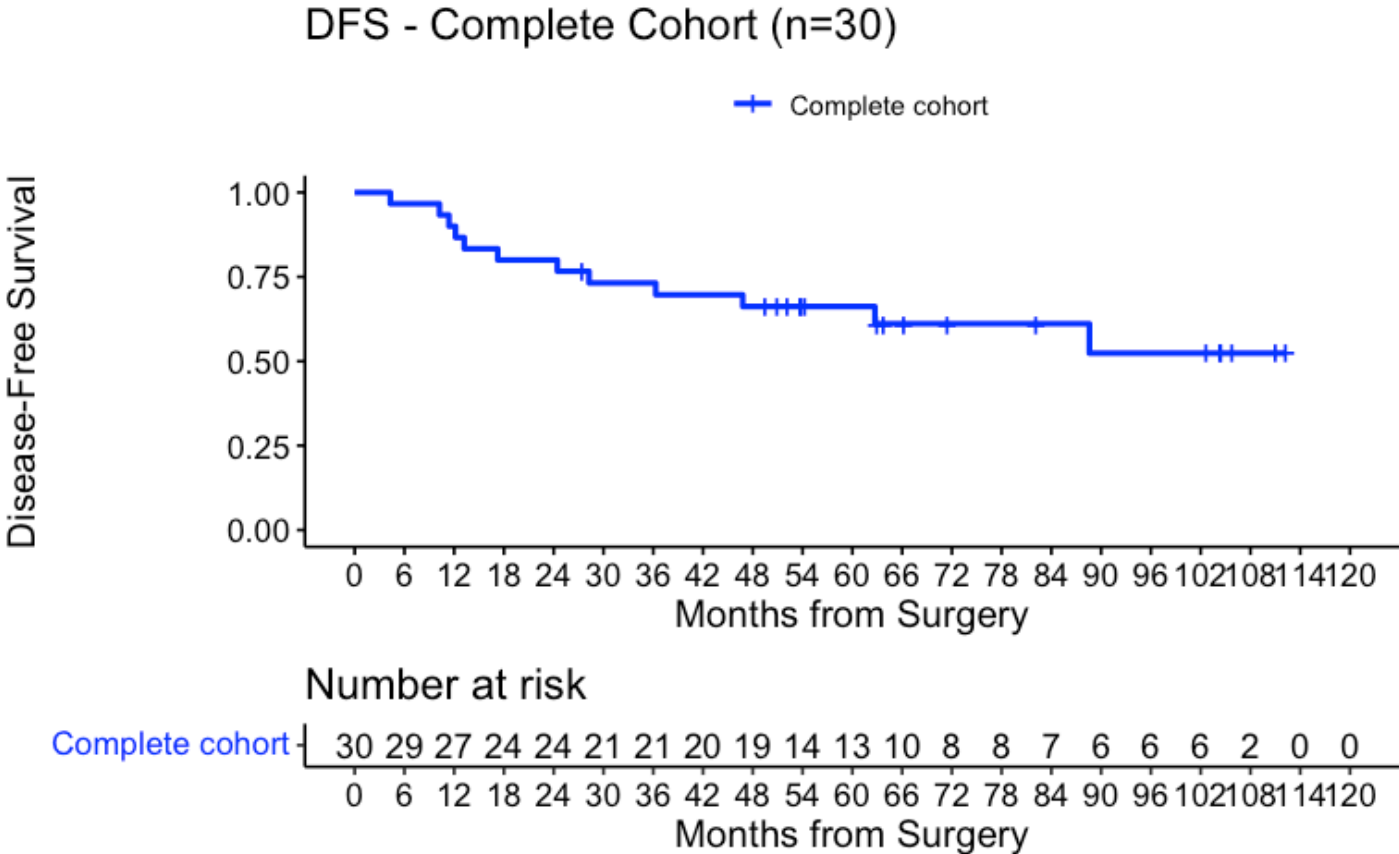
survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~Rectal.Rome, data
= circ_data)
```

```
Call: survfit(formula = Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event) ~
Rectal.Rome, data = circ_data)

      n events median 0.95LCL 0.95UCL
[1,] 30      12    NA    62.7     NA
```

Hide

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ Rectal.Rome, data = circ_data,conf.int=0.95,conf.type
="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRU
E, break.time.by=6, palette=c("blue"), title="DFS – Complete Cohort (n=30)", ylab= "Dise
ase-Free Survival", xlab="Months from Surgery", legend.labs=c("Complete cohort"), legen
d.title="")
```

Hide

```
summary(KM_curve, times= c(18, 24, 36))
```

```
Call: survfit(formula = surv_object ~ Rectal.Rome, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI
18	24	6	0.800	0.0730	0.608	0.905
24	24	0	0.800	0.0730	0.608	0.905
36	21	2	0.732	0.0812	0.534	0.856

#Total Follow-up in Complete Cohort

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

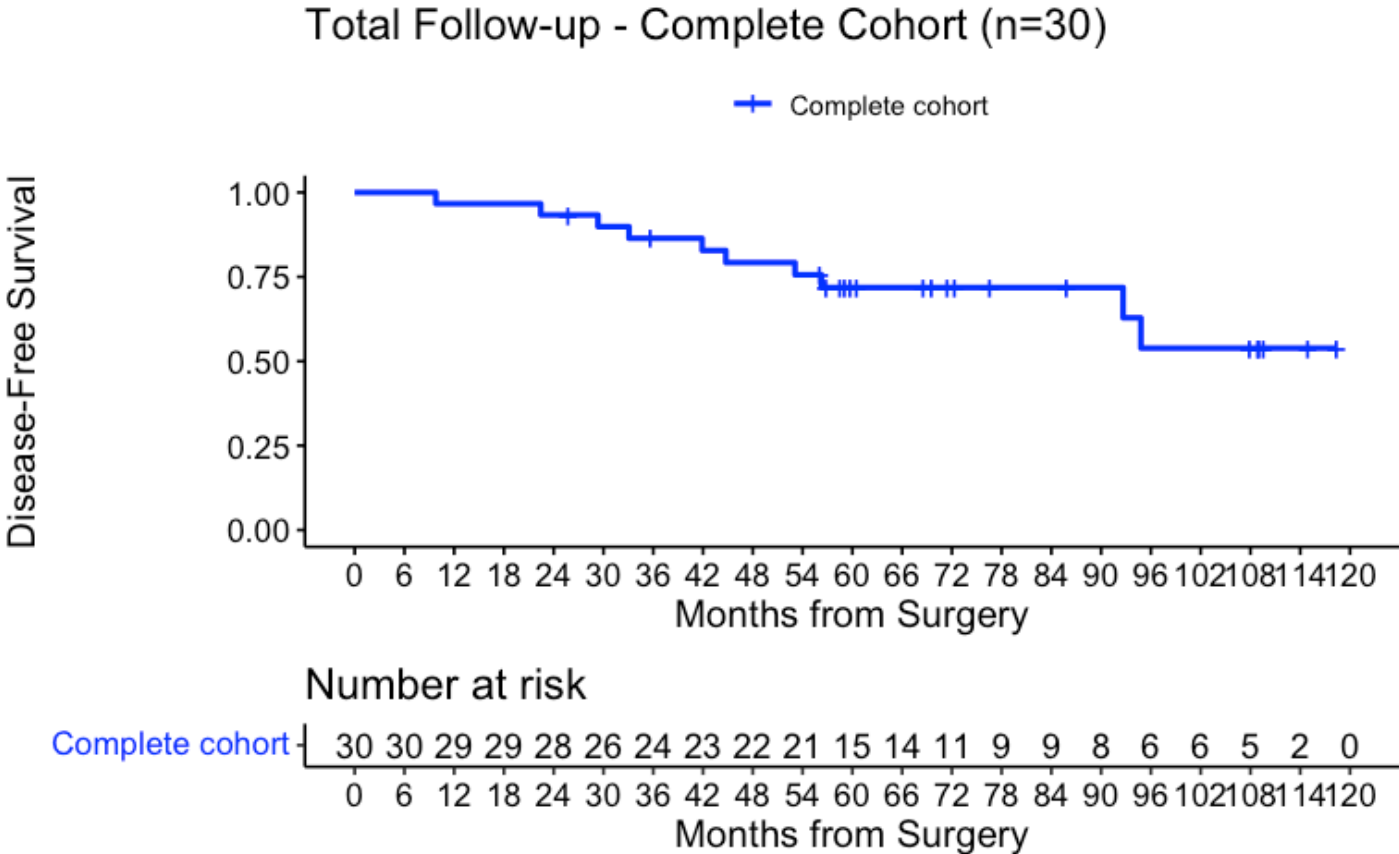
survfit(Surv(time = circ_data$Total.FU.months, event = circ_data$OS.Event)~Rectal.Rome,
data = circ_data)
```

```
Call: survfit(formula = Surv(time = circ_data$Total.FU.months, event = circ_data$OS.Event) ~
  Rectal.Rome, data = circ_data)
```

	n	events	median	0.95LCL	0.95UCL
[1,]	30	10	NA	92.7	NA

Hide

```
surv_object <-Surv(time = circ_data$Total.FU.months, event = circ_data$OS.Event)
KM_curve <- survfit(surv_object ~ Rectal.Rome, data = circ_data,conf.int=0.95,conf.type
="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE,
break.time.by=6, palette=c("blue"), title="Total Follow-up - Complete Cohort (n=30)",
ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("Complete cohort"),
legend.title="")
```



Hide

```
summary(KM_curve, times= c(18, 24, 36))
```

```
Call: survfit(formula = surv_object ~ Rectal.Rome, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI
18	29	1	0.967	0.0328	0.786	0.995
24	28	1	0.933	0.0455	0.759	0.983
36	24	2	0.864	0.0632	0.677	0.947

#Heatmap for the clinical factors

Hide

```

rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)
circ_data <- circ_data %>% arrange(cStage)
circ_datadf <- as.data.frame(circ_data)

ha <- HeatmapAnnotation(
  cStage = circ_data$cStage,
  Gender = circ_data$Gender,
  NAT.Regimen = circ_data$NAT.Regimen,
  Downstaging = circ_data$Downstaging,
  NAR.Score = circ_data$NAR.Score,
  pStage = circ_data$pStage,
  Grade = circ_data$Grade,
  ACT = circ_data$ACT,
  ACT.Regimen = circ_data$ACT.Regimen,
  ctDNA.Baseline = circ_data$ctDNA.Baseline,
  ctDNA.postNAC = circ_data$ctDNA.postNAC,
  ctDNA.postop = circ_data$ctDNA.postop,
  ctDNA.anytime = circ_data$ctDNA.anytime,
  RFS.Event = circ_data$RFS.Event,
  OS.Event = circ_data$OS.Event,

  col = list(cStage = c("I" = "seagreen1", "II" = "khaki", "III" = "orange"),
    Gender = c("Female" = "goldenrod", "Male" = "blue4"),
    NAT.Regimen = c("FOLF0X" = "coral", "Capecitabine" = "darkgreen"),
    Downstaging = c("R" = "yellow", "NR" = "brown"),
    NAR.Score = c("Low" = "lightblue", "Mid/High" = "orange"),
    pStage = c("0" = "khaki", "I" = "seagreen2", "II" = "cornflowerblue", "III" = "darkmagenta"),
    Grade = c("1" = "yellow3", "2" = "darkgreen", "3" = "brown2"),
    ACT = c("TRUE" = "darkmagenta", "FALSE" = "gray"),
    ACT.Regimen = c("FOLF0X" = "lightblue", "CAPOX" = "orange2", "Capecitabine" = "khaki"),
    ctDNA.Baseline = c("POSITIVE" = "red3", "NEGATIVE" = "blue"),
    ctDNA.postNAC = c("POSITIVE" = "red3", "NEGATIVE" = "blue"),
    ctDNA.postop = c("POSITIVE" = "red3", "NEGATIVE" = "blue"),
    ctDNA.anytime = c("POSITIVE" = "red3", "NEGATIVE" = "blue"),
    RFS.Event = c("TRUE" = "red3", "FALSE" = "blue"),
    OS.Event = c("TRUE" = "black", "FALSE" = "grey")
  )
)
ht <- Heatmap(matrix(nrow = 0, ncol = length(circ_data$cStage)), show_row_names = FALSE,
  cluster_rows = F, cluster_columns = FALSE, top_annotation = ha)
pdf("heatmap.pdf", width = 7, height = 7)
draw(ht, annotation_legend_side = "bottom")
dev.off()

```

```

null device
      1

```

#Overview plot

Hide

```

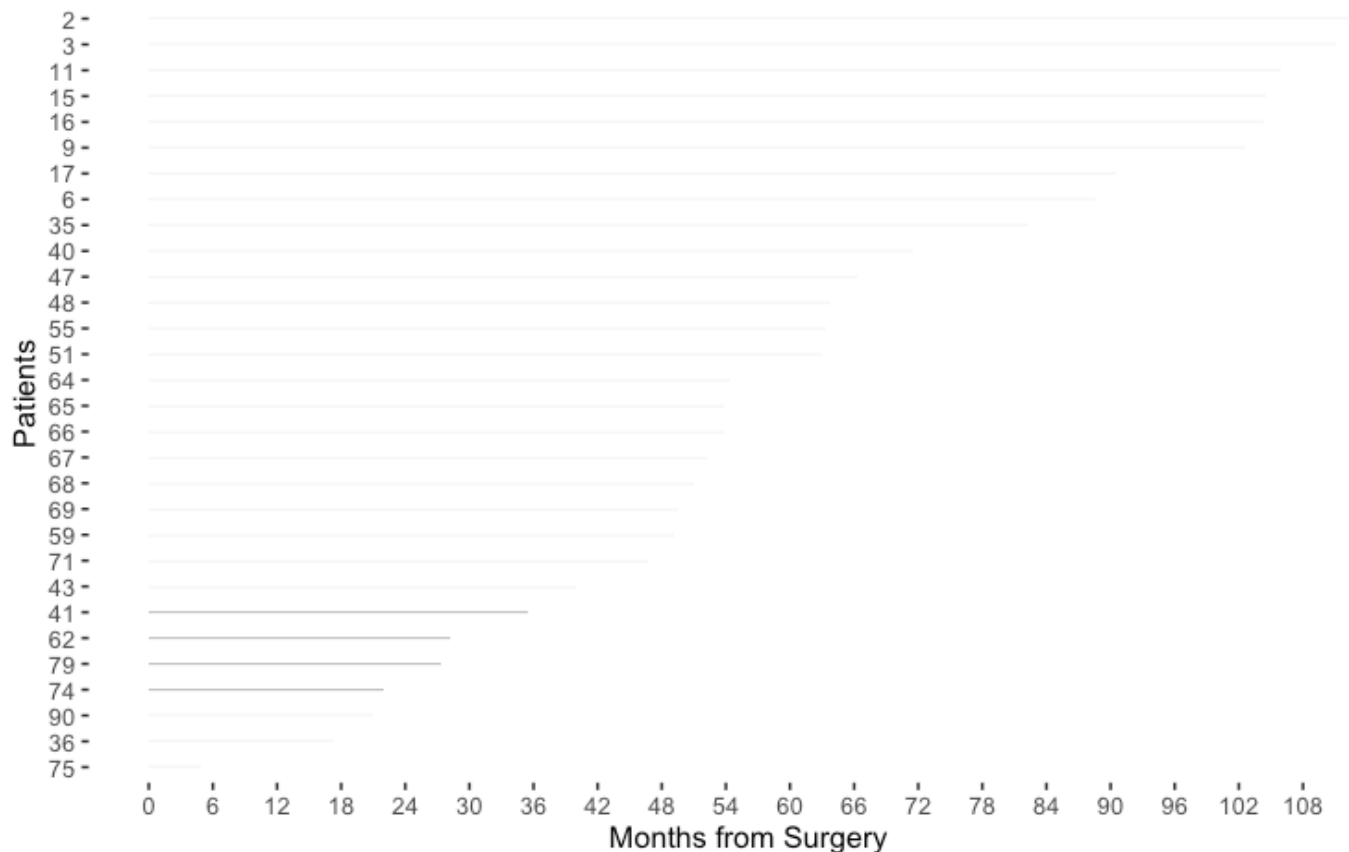
library(swimplot)
library(ggplot2)
library(grid)
library(gtable)

##Overview plot
setwd("~/Downloads")
clinstage<- read.csv("IRST_Rectal_OP.csv")
clinstage_df<- as.data.frame(clinstage)

#Display the swimmer plot with the label box
oplot<-swimmer_plot(df=clinstage_df,
                    id='PatientName',
                    end='fu.diff.months',
                    fill='gray',
                    width=.01,)

oplot <- oplot + theme(panel.border = element_blank())
oplot <- oplot + scale_y_continuous(breaks = seq(-12, 108, by = 6))
oplot <- oplot + labs(x ="Patients" , y="Months from Surgery")
oplot

```

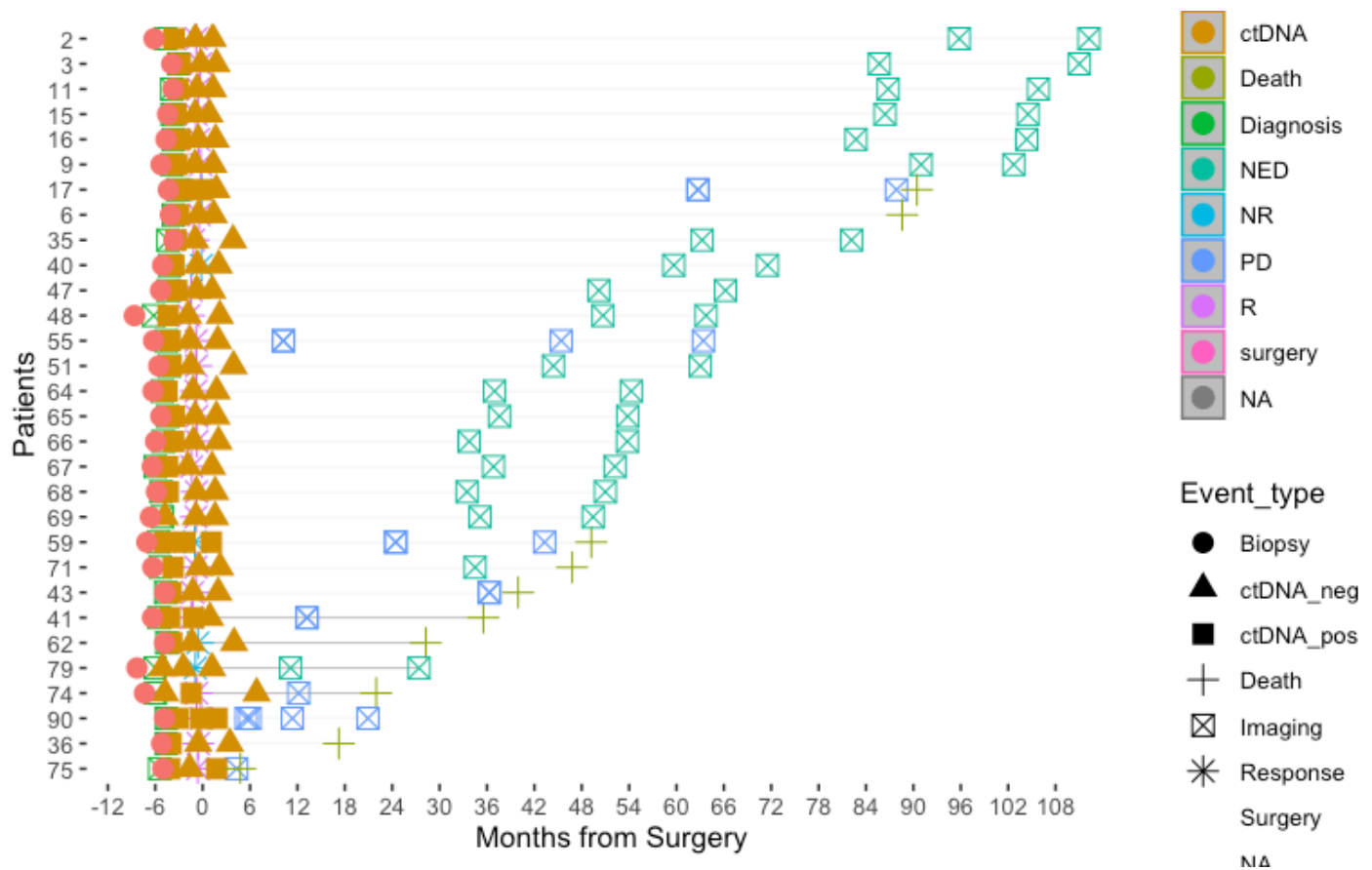


Hide

```
##plot events
oplot_ev1 <- oplot + swimmer_points(df_points=clinstage_df,
                                   id='PatientName',
                                   time='date.diff.months',
                                   name_shape = 'Event_type',
                                   name_col = 'Event',
                                   size=3.5,fill='black',
                                   #col='darkgreen'
                                   )
oplot_ev1
```

Warning: The shape palette can deal with a maximum of 6 discrete values because more than 6 becomes difficult to discriminate; you have 7. Consider specifying shapes manually if you must have them.

Warning: Removed 82 rows containing missing values (`geom_point()`).



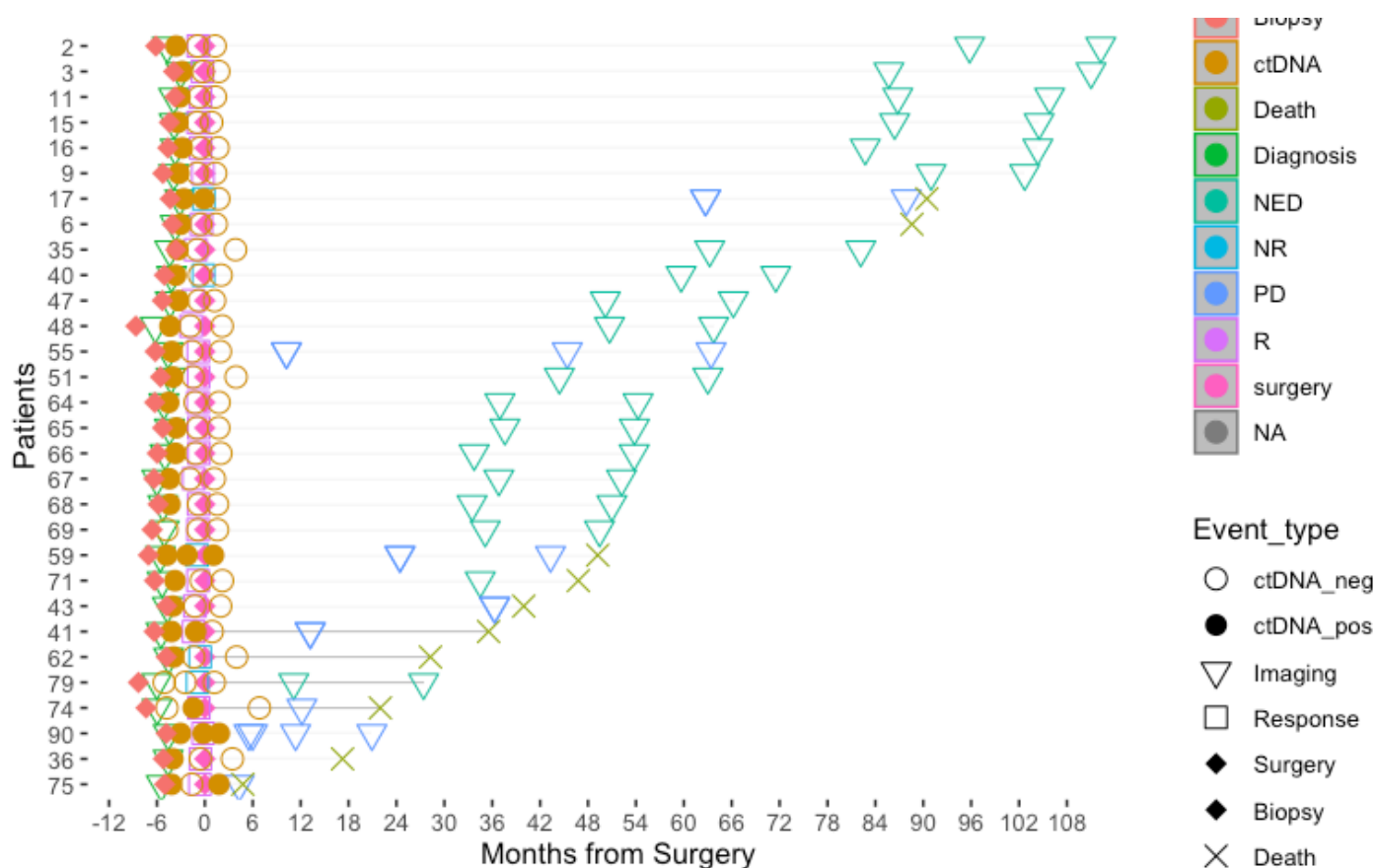
Hide

```
#Shape customization to Event_type
```

```
oplot_ev1.1 <- oplot_ev1 + ggplot2::scale_shape_manual(name="Event_type",values=c(1,16,
6,0,18,18,4),breaks=c('ctDNA_neg','ctDNA_pos','Imaging','Response','Surgery','Biopsy',
'Death'))
```

```
oplot_ev1.1
```

Warning: Removed 52 rows containing missing values (`geom_point()`).



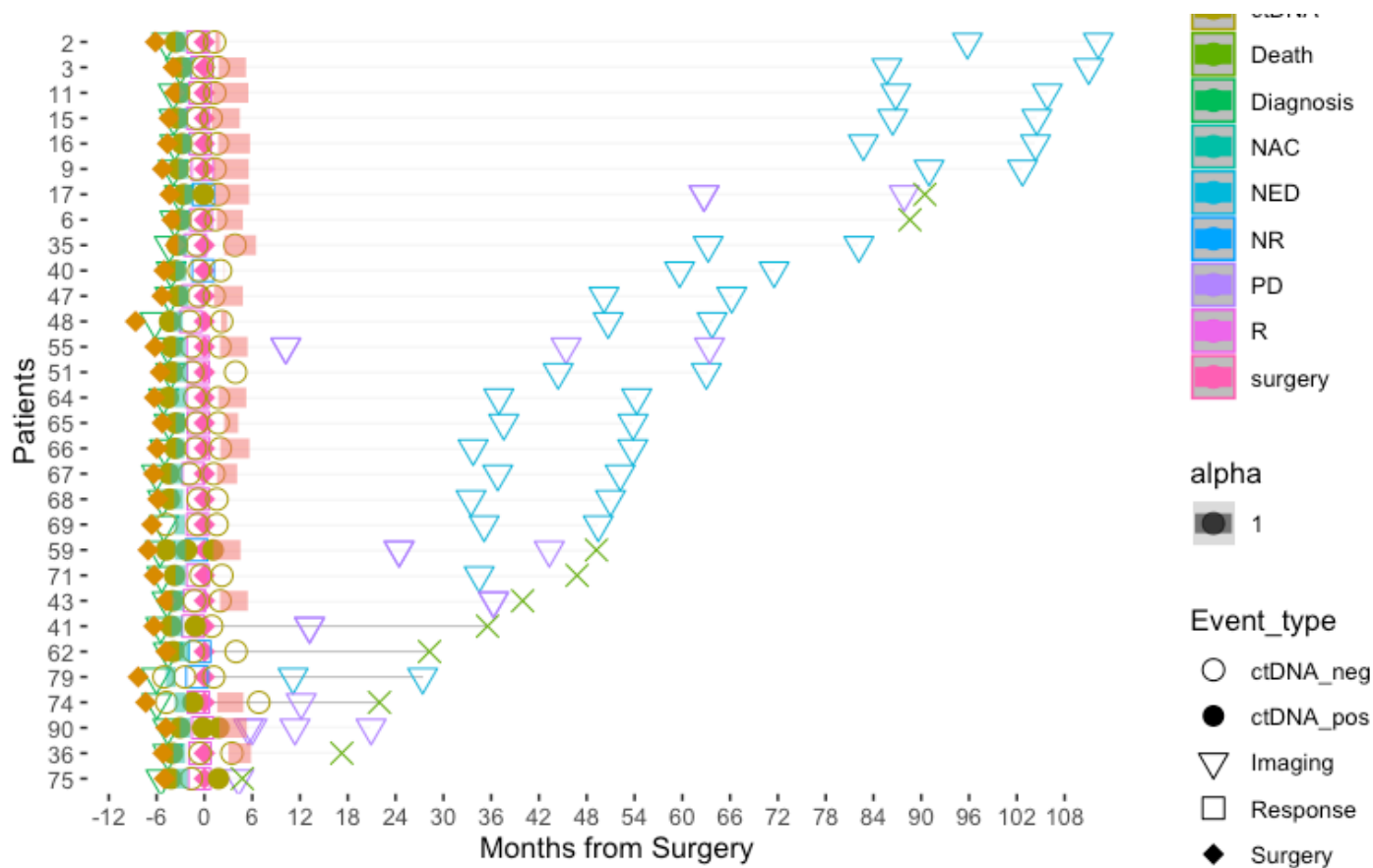
Hide

```
#plot treatment
```

```
oplot_ev2 <- oplot_ev1.1 + swimmer_lines(df_lines=clinstage_df,
                                         id='PatientName',
                                         start='Tx_start.months',
                                         end='Tx_end.months',
                                         name_col='Tx_type',
                                         size=3.5,
                                         name_alpha = 1.0)
oplot_ev2 <- oplot_ev2 + guides(linetype = guide_legend(override.aes = list(size = 5, co
lor = "black")))
oplot_ev2
```


Warning: Removed 52 rows containing missing values (`geom_point()`).

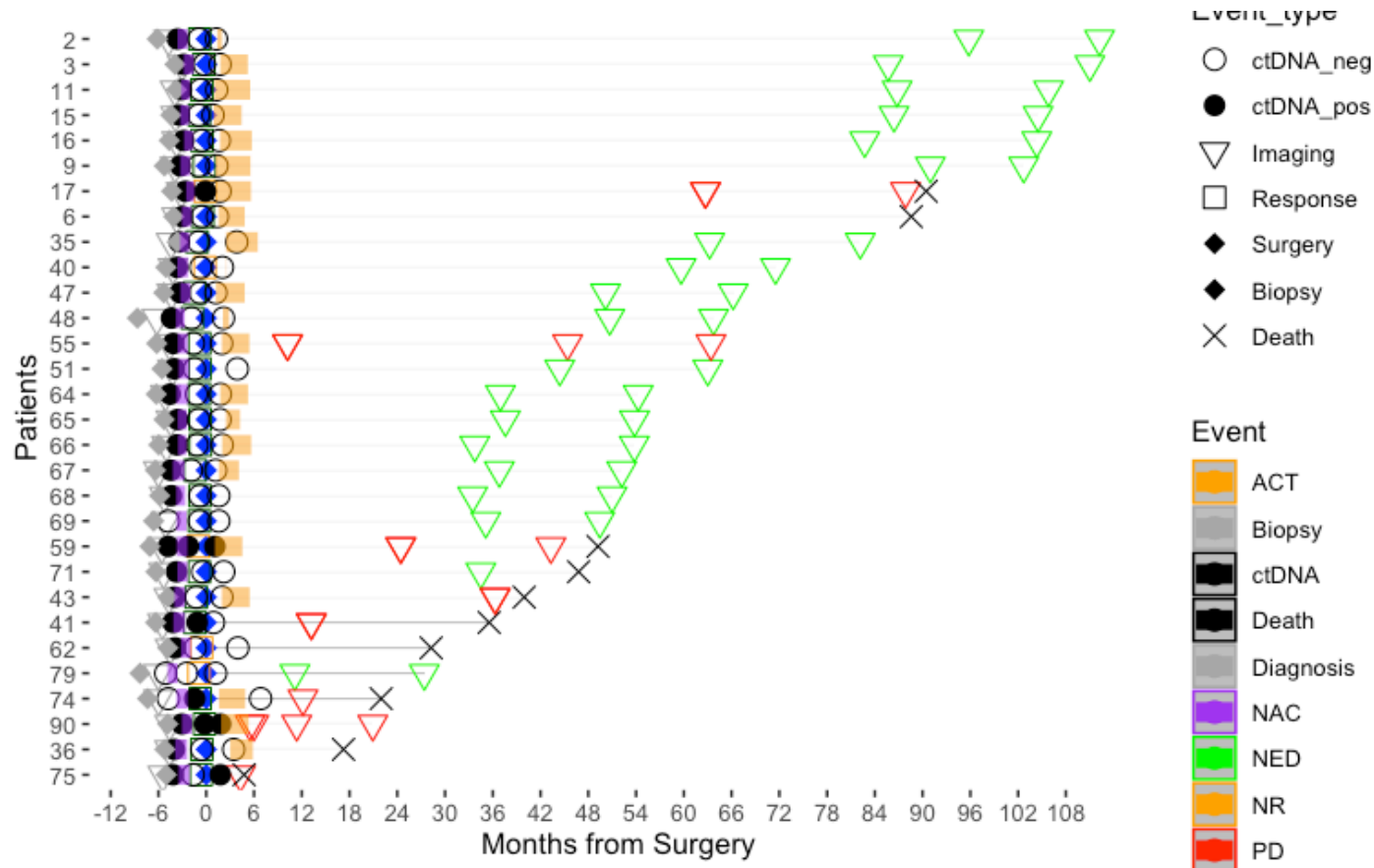
Warning: Removed 277 rows containing missing values (`geom_segment()`).


[Hide](#)

```
#colour customization
##Dark Gray=MRI diagnosis, purple=ACT, Black=Death, Red=PD, ctDNA negative=white, ctDNA
positive=black, surgery=blue, Resp=Dark Green,
oplot_ev2.2 <- oplot_ev2 + ggplot2::scale_color_manual(name="Event",values=c( "orang
e","darkgray","black", "black", "darkgray", "purple", "green", "orange", "red", "darkgre
en", "blue"))
oplot_ev2.2
```

Warning: Removed 52 rows containing missing values (`geom_point()`).

Warning: Removed 277 rows containing missing values (`geom_segment()`).



#DFS by ctDNA post-NAT

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.postNAC, data = circ_data)
```

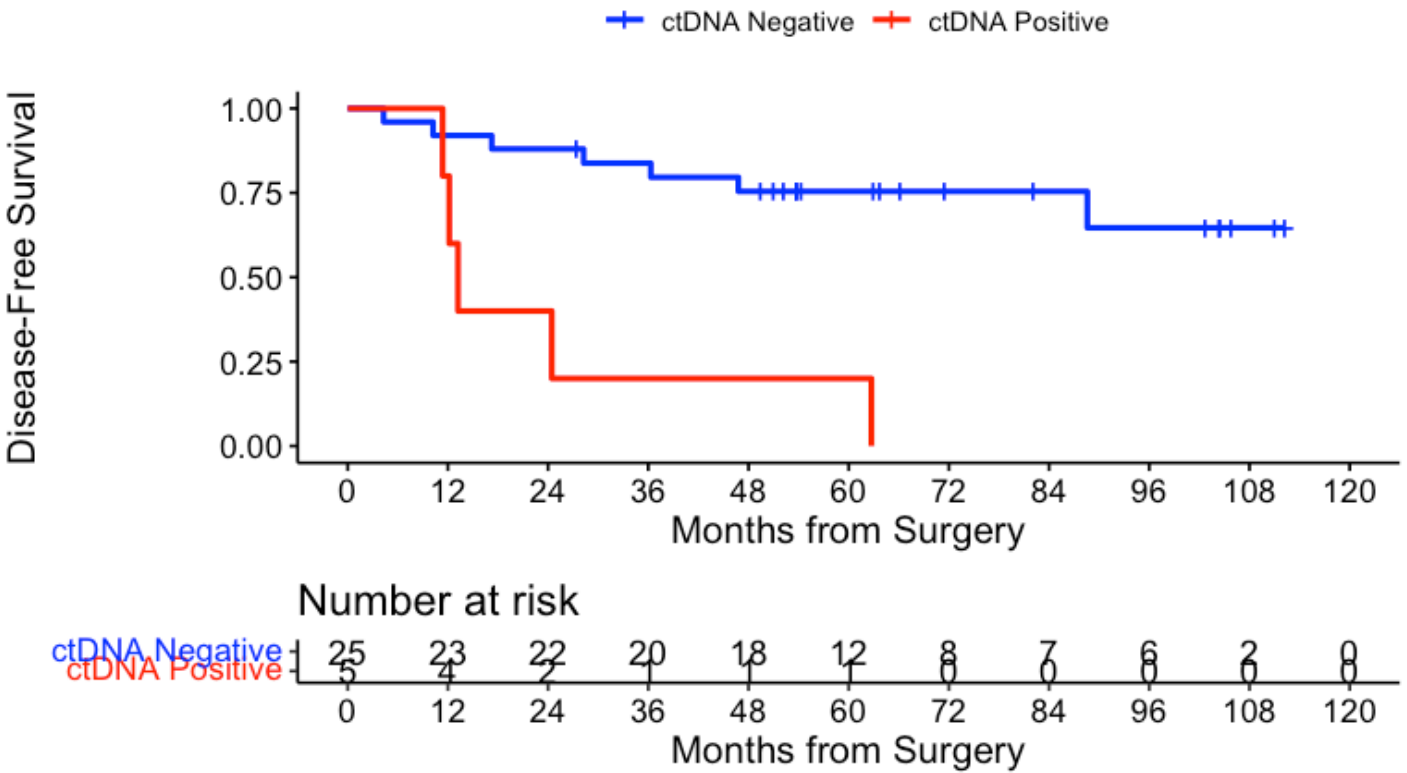
Call: `survfit(formula = Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event) ~ ctDNA.postNAC, data = circ_data)`

	n	events	median	0.95LCL	0.95UCL
ctDNA.postNAC=NEGATIVE	25	7	NA	88.6	NA
ctDNA.postNAC=POSITIVE	5	5	13.2	12.2	NA

Hide

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.postNAC, data = circ_data, conf.int=0.95, conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue","red"), title="DFS - ctDNA post-NAT", ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("ctDNA Negative", "ctDNA Positive"), legend.title="")
```

DFS - ctDNA post-NAT



Hide

```
summary(KM_curve, times= c(12))
```

```
Call: survfit(formula = surv_object ~ ctDNA.postNAC, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

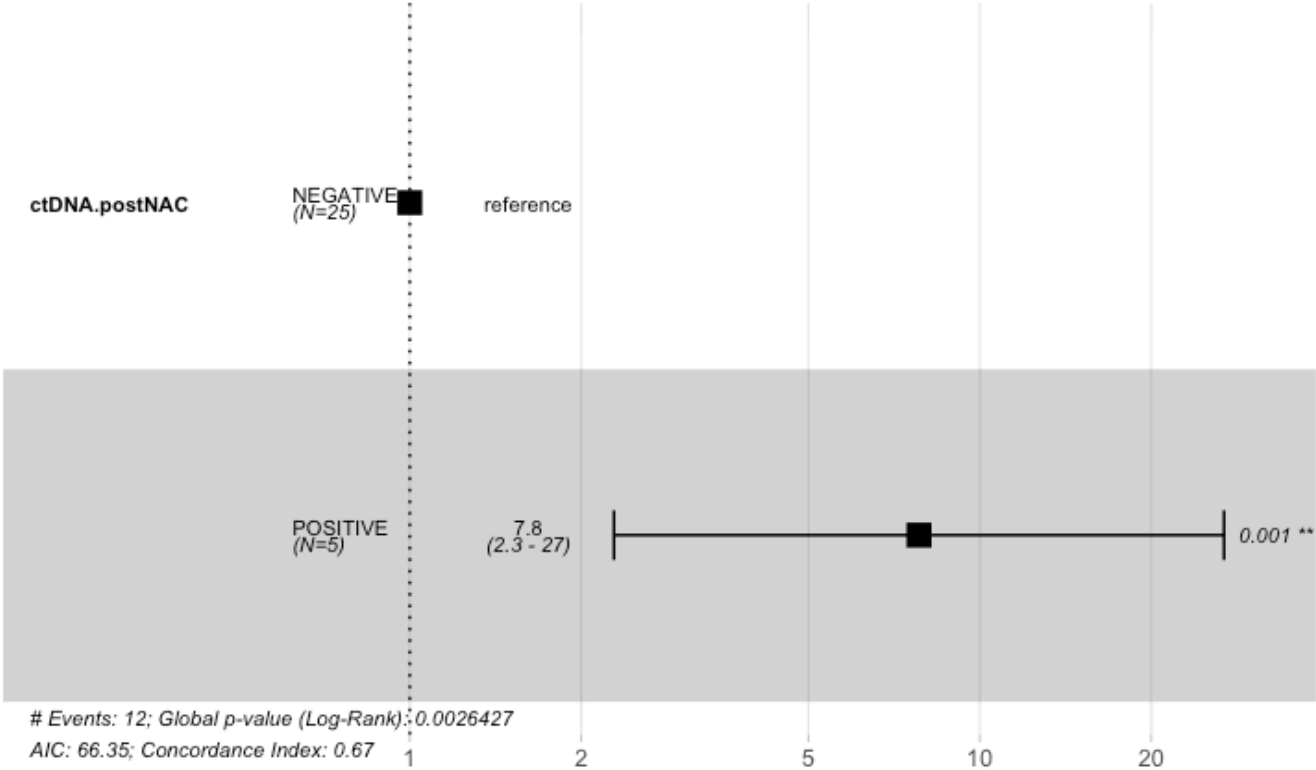
ctDNA.postNAC=NEGATIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
94	12.0000	23.0000	2.0000	0.9200	0.0543	0.7164	0.97

ctDNA.postNAC=POSITIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
69	12.000	4.000	1.000	0.800	0.179	0.204	0.9

Hide

```
circ_data$ctDNA.postNAC <- factor(circ_data$ctDNA.postNAC, levels=c("NEGATIVE","POSITIV  
E"))  
cox_fit <- coxph(surv_object ~ ctDNA.postNAC, data=circ_data)  
ggforest(cox_fit,data = circ_data)
```

Hazard ratio



Hide

```
summary(cox_fit)
```

Call:
coxph(formula = surv_object ~ ctDNA.postNAC, data = circ_data)

n= 30, number of events= 12

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.postNACPOSITIVE	2.0570	7.8221	0.6287	3.272	0.00107 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.postNACPOSITIVE	7.822	0.1278	2.281	26.82

Concordance= 0.669 (se = 0.064)
Likelihood ratio test= 9.04 on 1 df, p=0.003
Wald test = 10.7 on 1 df, p=0.001
Score (logrank) test = 14.51 on 1 df, p=1e-04

Hide

```
cox_fit_summary <- summary(cox_fit)

#Extract values for HR, 95% CI, and p-value
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 7.82 (2.28-26.82); p = 0.001"
```

[Hide](#)

```
#Fisher plot for ctDNA post-NAT and Rec Status
circ_data$ctDNA.postNAC <- factor(circ_data$ctDNA.postNAC, levels = c("NEGATIVE", "POSITIVE"), labels = c("ctDNA(-)", "ctDNA(+)))
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$ctDNA.postNAC, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

```
Warning in stats::chisq.test(x, y, ...) :
  Chi-squared approximation may be incorrect
```

[Hide](#)

```
print(chi_square_test)
```

Pearson's Chi-squared test with Yates' continuity correction

```
data: contingency_table
X-squared = 6.25, df = 1, p-value = 0.01242
```

[Hide](#)

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.005558
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 1.742654      Inf
sample estimates:
odds ratio
      Inf
```

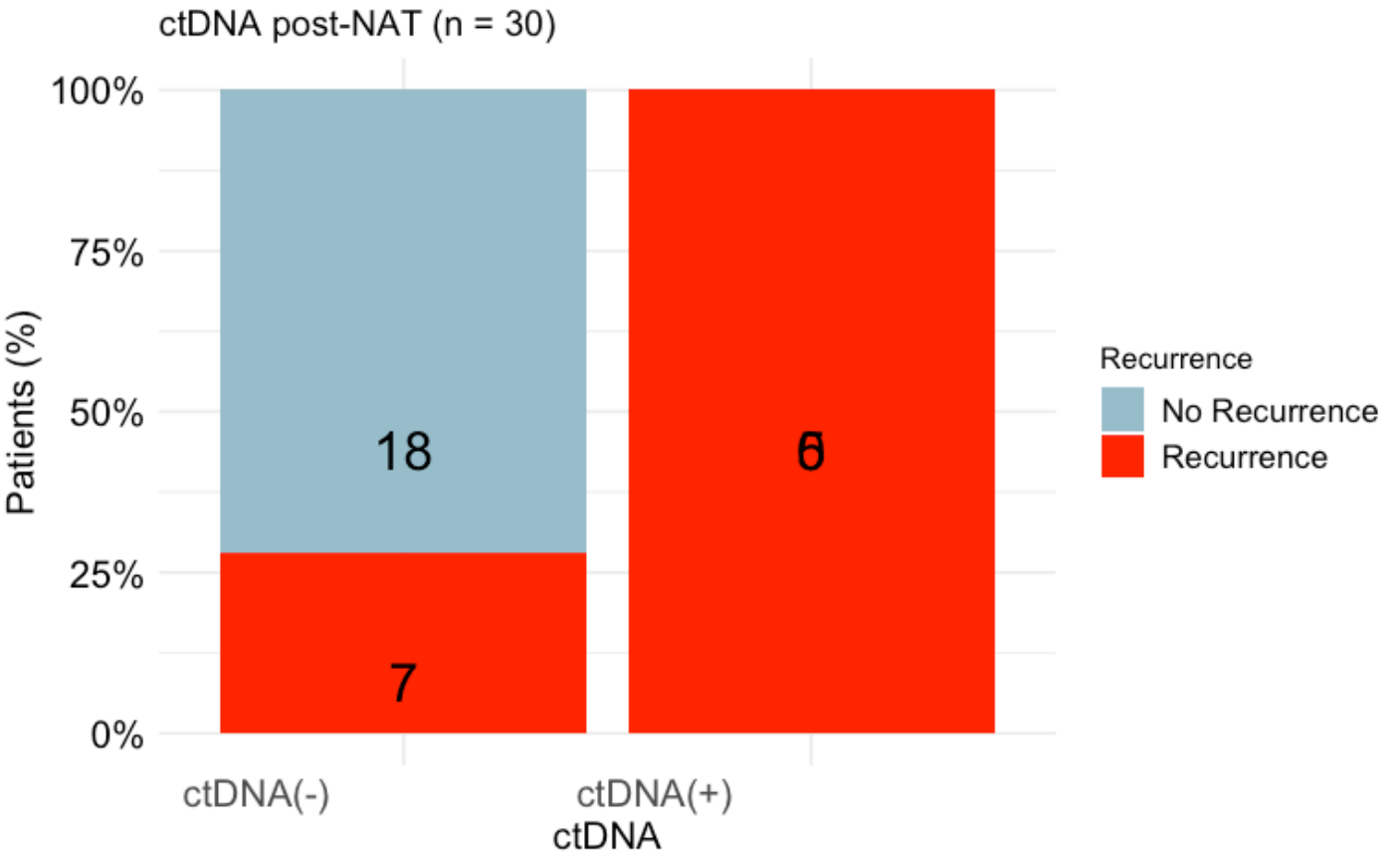
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
ctDNA(-)	18	7
ctDNA(+)	0	5

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black", vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA post-NAT (n = 30)", x = "ctDNA", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
        axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
        axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
        axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
        legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#DFS by ctDNA clearance post-NAT

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.NAC.Clearance, data = circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$DFS.months, event = circ_data\$DFS.Event) ~ ctDNA.NAC.Clearance, data = circ_data)

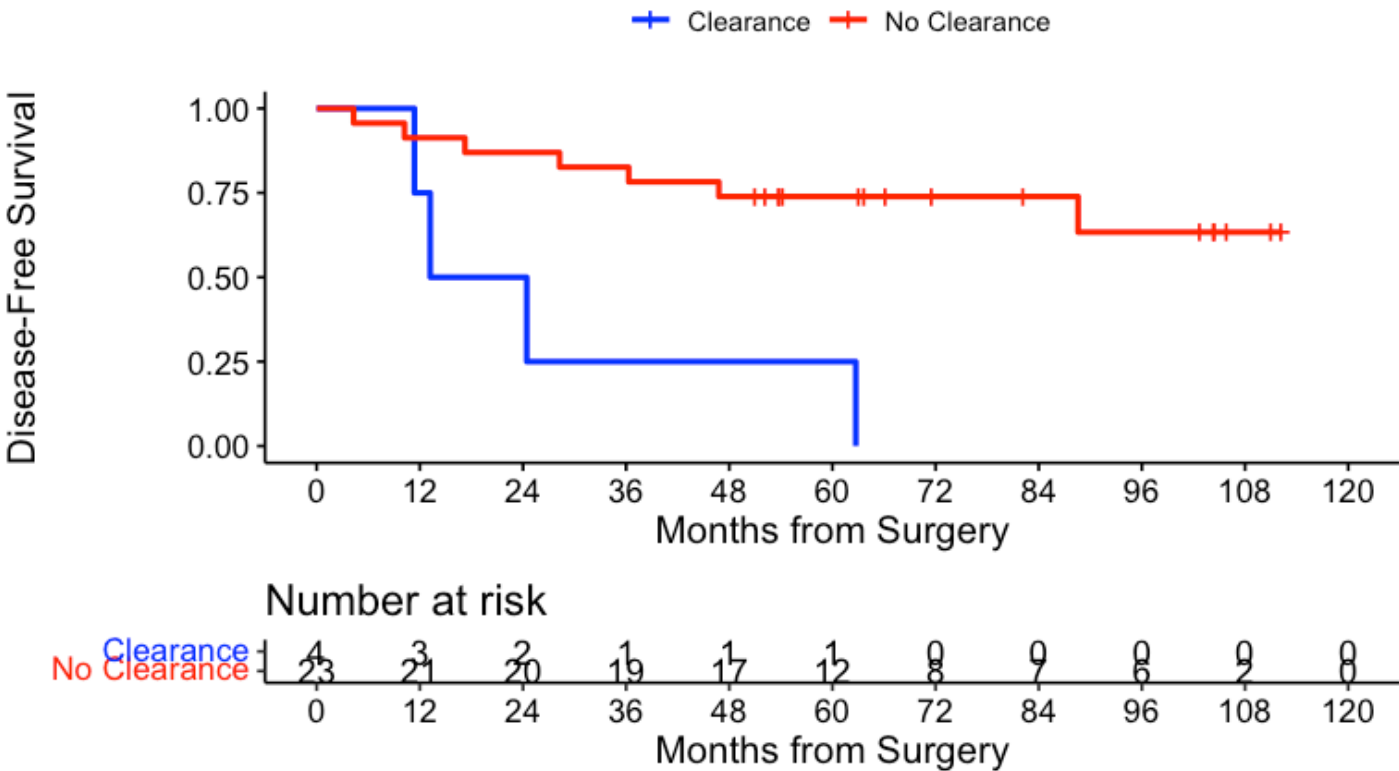
3 observations deleted due to missingness

	n	events	median	0.95LCL	0.95UCL
ctDNA.NAC.Clearance=FALSE	4	4	18.8	11.4	NA
ctDNA.NAC.Clearance=TRUE	23	7	NA	88.6	NA

Hide


```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.NAC.Clearance, data = circ_data,conf.int=0.95,conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue","red"), title="DFS - ctDNA Clearance post-NAT", ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("Clearance", "No Clearance"), legend.title="")
```

DFS - ctDNA Clearance post-NAT



Hide

```
summary(KM_curve, times= c(12))
```

```
Call: survfit(formula = surv_object ~ ctDNA.NAC.Clearance, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

3 observations deleted due to missingness

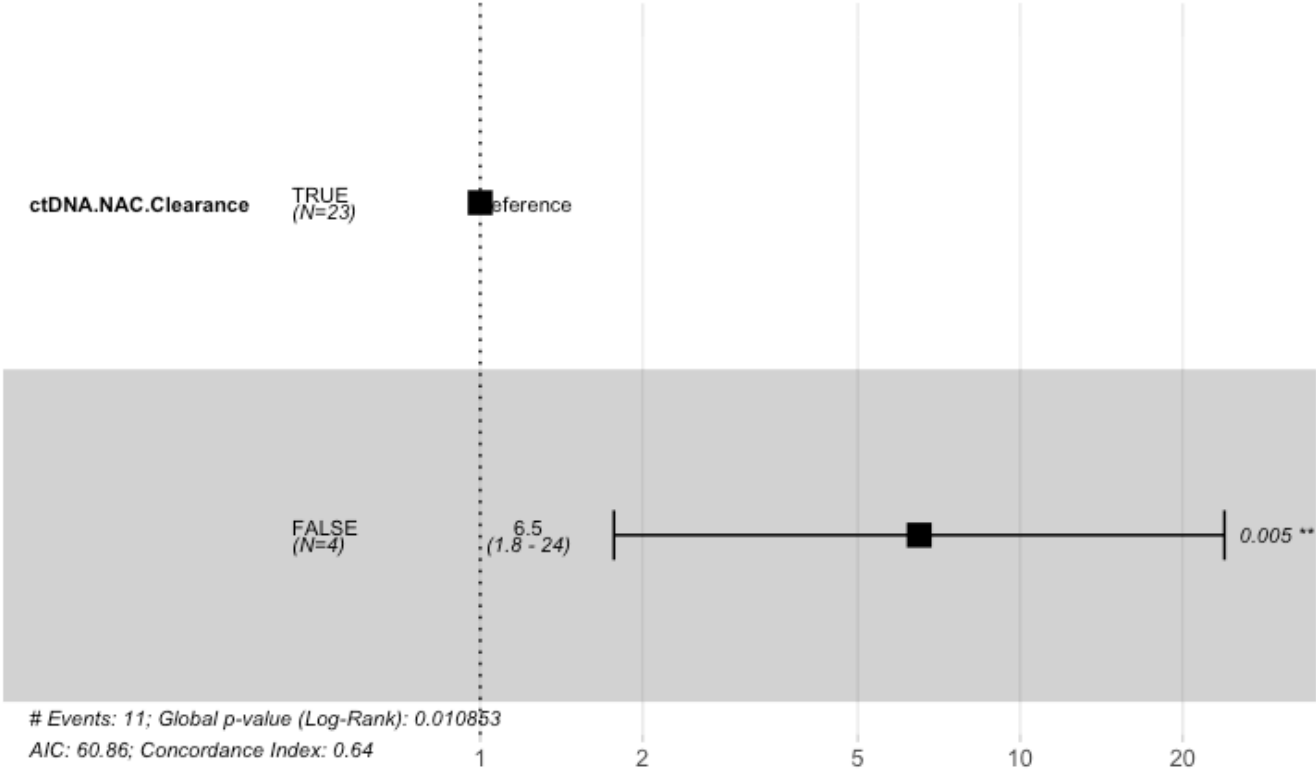
ctDNA.NAC.Clearance=FALSE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
61	12.000	3.000	1.000	0.750	0.217	0.128	0.9

ctDNA.NAC.Clearance=TRUE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
75	12.0000	21.0000	2.0000	0.9130	0.0588	0.6949	0.97

Hide

```
circ_data$ctDNA.NAC.Clearance <- factor(circ_data$ctDNA.NAC.Clearance, levels=c("TRUE", "FALSE"))
cox_fit <- coxph(surv_object ~ ctDNA.NAC.Clearance, data=circ_data)
ggforest(cox_fit, data = circ_data)
```

Hazard ratio



Hide

```
summary(cox_fit)
```

Call:
coxph(formula = surv_object ~ ctDNA.NAC.Clearance, data = circ_data)

n= 27, number of events= 11
(3 observations deleted due to missingness)

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.NAC.ClearanceFALSE	1.8718	6.4998	0.6642	2.818	0.00483 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.NAC.ClearanceFALSE	6.5	0.1539	1.768	23.89

Concordance= 0.641 (se = 0.064)
Likelihood ratio test= 6.49 on 1 df, p=0.01
Wald test = 7.94 on 1 df, p=0.005
Score (logrank) test = 10.33 on 1 df, p=0.001

Hide

```
cox_fit_summary <- summary(cox_fit)

#Extract values for HR, 95% CI, and p-value
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 6.5 (1.77-23.89); p = 0.005"
```

Hide

```
#Fisher plot for ctDNA clearance post-NAT and Rec Status
circ_data$ctDNA.NAC.Clearance <- factor(circ_data$ctDNA.NAC.Clearance, levels = c("TRUE", "FALSE"), labels = c("Clearance", "No Clearance"))
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$ctDNA.NAC.Clearance, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

```
Warning in stats::chisq.test(x, y, ...) :
  Chi-squared approximation may be incorrect
```

Hide

```
print(chi_square_test)
```

Pearson's Chi-squared test with Yates' continuity correction

```
data: contingency_table
X-squared = 4.2525, df = 1, p-value = 0.03919
```

Hide

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.0188
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 1.130759      Inf
sample estimates:
odds ratio
      Inf
```

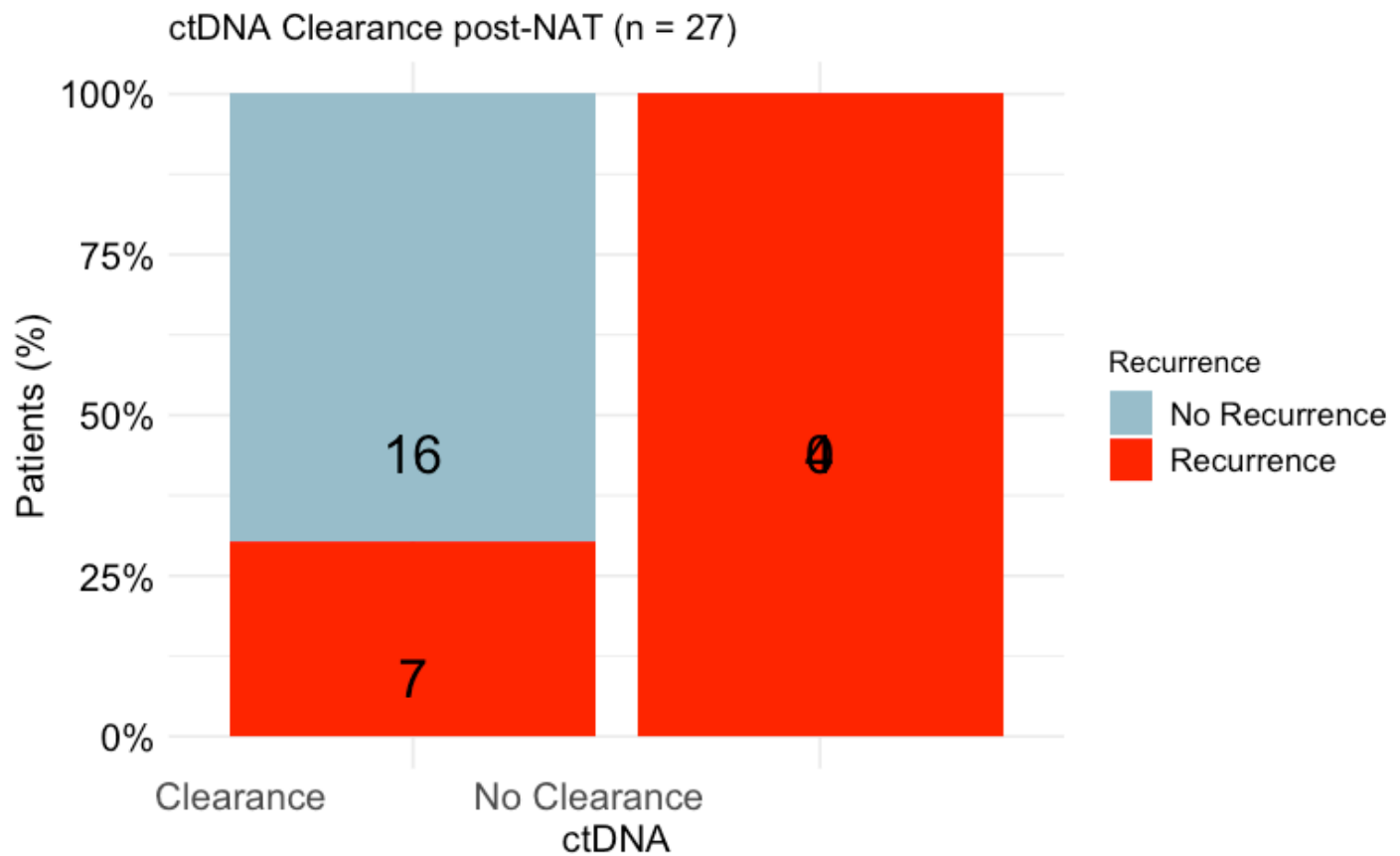
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
Clearance	16	7
No Clearance	0	4

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA Clearance post-NAT (n = 27)", x = "ctDNA", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#Logistic regression for association between ctDNA Clearance & Radiological Recurrence

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

#Fisher plot for ctDNA clearance post-NAT and Rec Status
circ_data$ctDNA.NAC.Clearance <- factor(circ_data$ctDNA.NAC.Clearance, levels = c("TRUE", "FALSE"), labels = c("Clearance", "No Clearance"))
circ_data$Downstaging <- factor(circ_data$Downstaging, levels = c("R", "NR"), labels = c("Responders", "Non-Responders"))
contingency_table <- table(circ_data$ctDNA.NAC.Clearance, circ_data$Downstaging)
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.09179
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
  0.4442762 204.4797745
sample estimates:
odds ratio
  9.059876
```

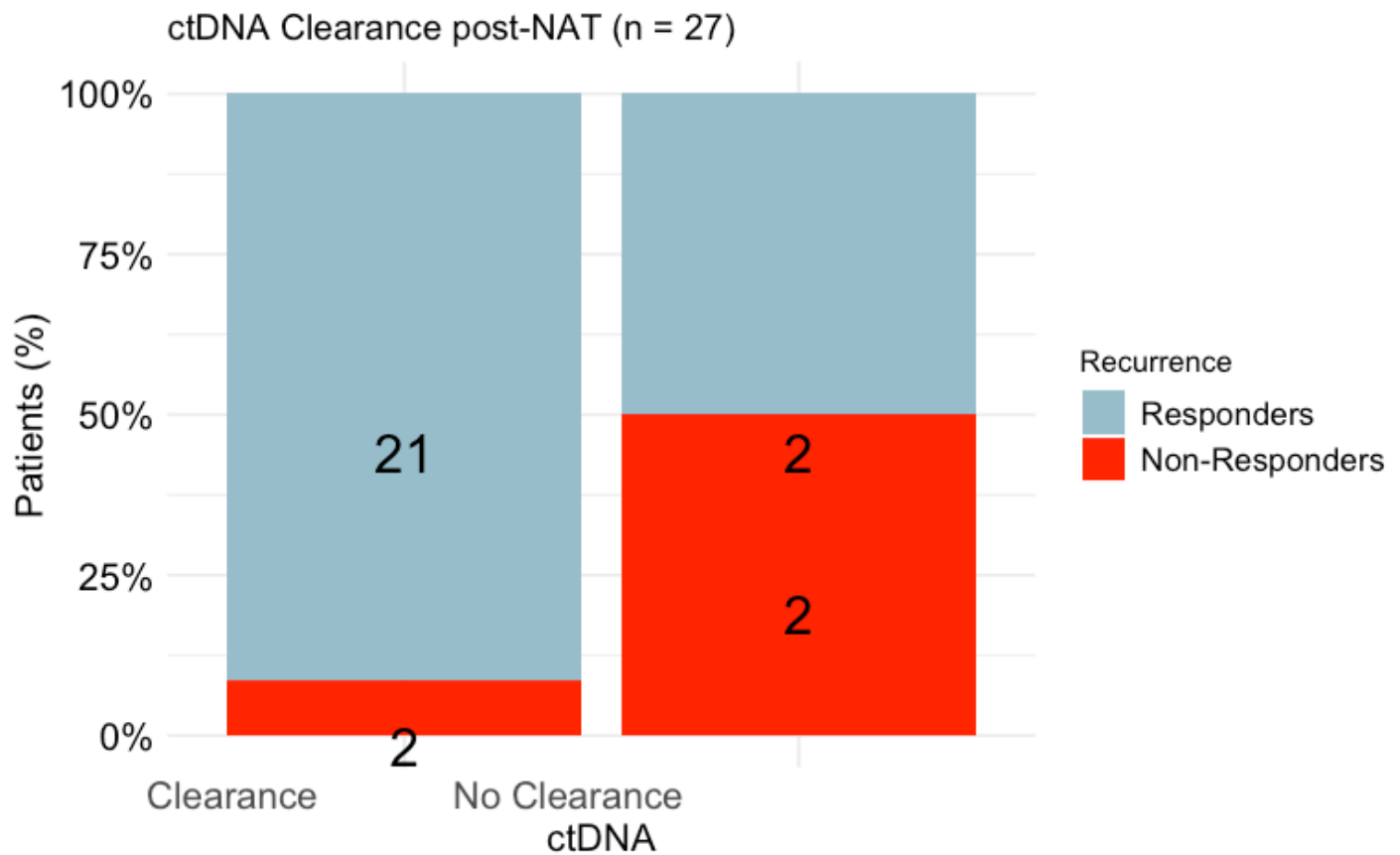
Hide

```
print(contingency_table)
```

	Responders	Non-Responders
Clearance	21	2
No Clearance	2	2

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA Clearance post-NAT (n = 27)", x = "ctDNA", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("Responders" = "lightblue3", "Non-Responders" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```


[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

#Vertical Fisher plot for ctDNA clearance post-NAT and Rec Status
circ_data$ctDNA.NAC.Clearance <- factor(circ_data$ctDNA.NAC.Clearance, levels = c("TRUE", "FALSE"), labels = c("Clearance", "No Clearance"))
circ_data$Downstaging <- factor(circ_data$Downstaging, levels = c("R", "NR"), labels = c("Responders", "Non-Responders"))
contingency_table <- table(circ_data$ctDNA.NAC.Clearance, circ_data$Downstaging)
fisher_exact_test <- fisher.test(contingency_table)
chi_square_test <- chisq.test(contingency_table)
```

Warning in stats::chisq.test(x, y, ...) :
Chi-squared approximation may be incorrect

[Hide](#)

```
print(chi_square_test)
```


Pearson's Chi-squared test with Yates' continuity correction

```
data: contingency_table  
X-squared = 1.9148, df = 1, p-value = 0.1664
```

Hide

```
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table  
p-value = 0.09179  
alternative hypothesis: true odds ratio is not equal to 1  
95 percent confidence interval:  
  0.4442762 204.4797745  
sample estimates:  
odds ratio  
  9.059876
```

Hide

```
print(contingency_table)
```

	Responders	Non-Responders
Clearance	21	2
No Clearance	2	2

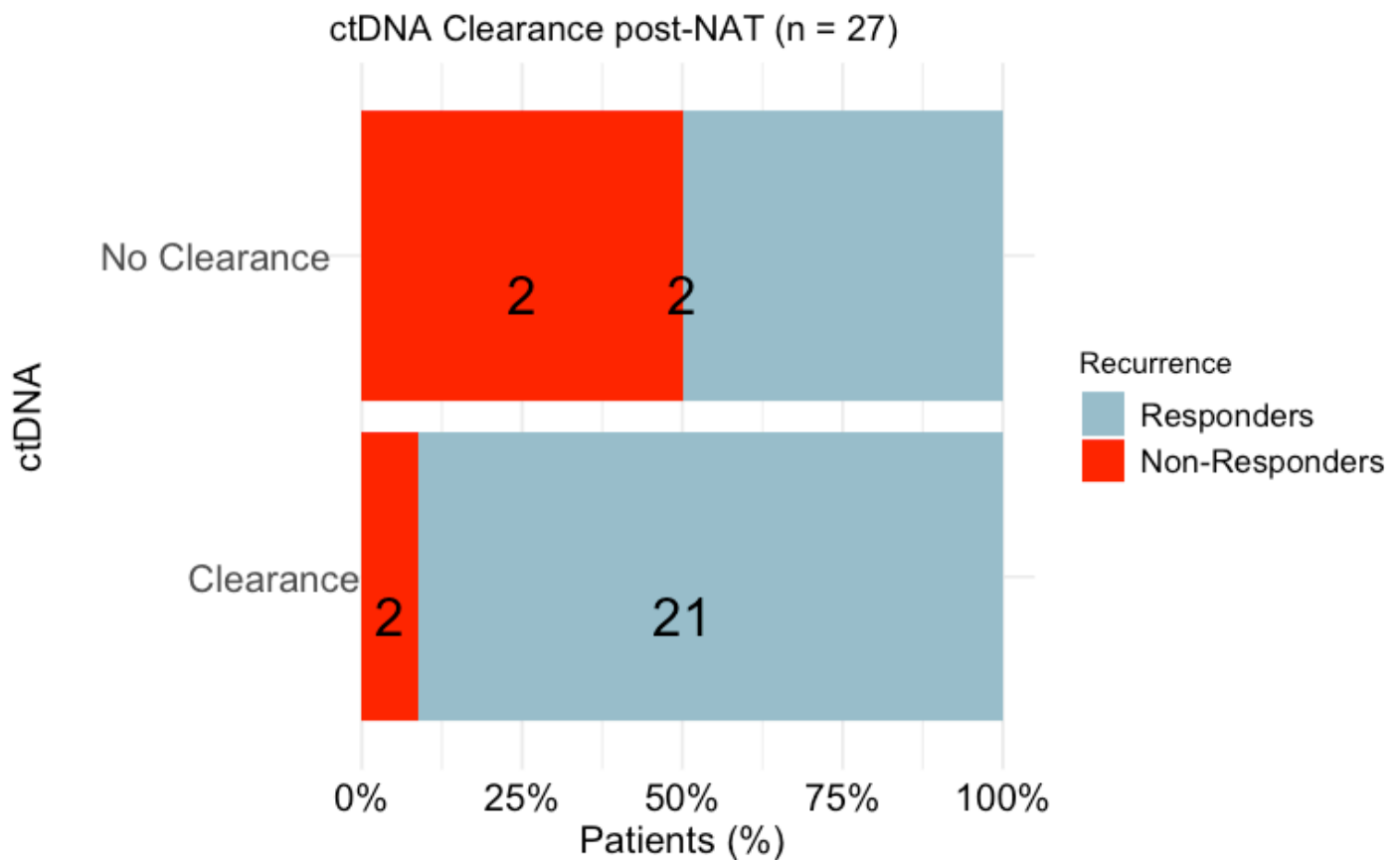
Hide

```

table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2

# Swapping x and y in ggplot function to make bar plot vertical
ggplot(table_df, aes(y = Var1, x = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(x = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA Clearance post-NAT (n = 27)", y = "ctDNA", x = "Patients (%)", fill = "Recurrence") +
  scale_x_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("Responders" = "lightblue3", "Non-Responders" = "red")) +
  # define custom colors
  theme(axis.text.y = element_text(angle = 0, hjust = 1.5, size = 14), # increase y-axis text size
    axis.text.x = element_text(size = 14, color = "black"), # increase x-axis text size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size

```



Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

circ_data$ctDNA.NAC.Clearance <- factor(circ_data$ctDNA.NAC.Clearance, levels = c("FALSE", "TRUE"))
circ_data$Downstaging <- factor(circ_data$Downstaging, levels = c("NR", "R"))

logit_model <- glm(Downstaging ~ ctDNA.NAC.Clearance, data = circ_data, family = "binomial")
summary(logit_model)
```

Call:
 glm(formula = Downstaging ~ ctDNA.NAC.Clearance, family = "binomial",
 data = circ_data)

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.2101	0.4265	0.4265	0.4265	1.1774

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.585e-15	1.000e+00	0.00	1.0000
ctDNA.NAC.ClearanceTRUE	2.351e+00	1.244e+00	1.89	0.0587 .

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 22.652 on 26 degrees of freedom
 Residual deviance: 19.135 on 25 degrees of freedom
 (3 observations deleted due to missingness)
 AIC: 23.135

Number of Fisher Scoring iterations: 5

Hide

```
OR <- exp(coef(logit_model))
CI <- exp(confint(logit_model))
```

Waiting for profiling to be done...

Hide

```
P_values <- summary(logit_model)$coefficients[,4]
result <- data.frame(
  OddsRatio = OR,
  Lower95CI = CI[, 1],
  Upper95CI = CI[, 2],
  Pvalue = P_values)
print(result)
```

	OddsRatio <dbl>	Lower95CI <dbl>	Upper95CI <dbl>	Pvalue <dbl>
(Intercept)	1.0	0.1199776	8.334891	1.00000000
ctDNA.NAC.ClearanceTRUE	10.5	0.8923342	148.452227	0.05874159

2 rows

#DFS by ctDNA at MRD time point

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.MRD!="",]
circ_data$DFS.months=circ_data$DFS.months-3
circ_data <- circ_data[circ_data$DFS.months>=0,]
circ_datadf <- as.data.frame(circ_data)

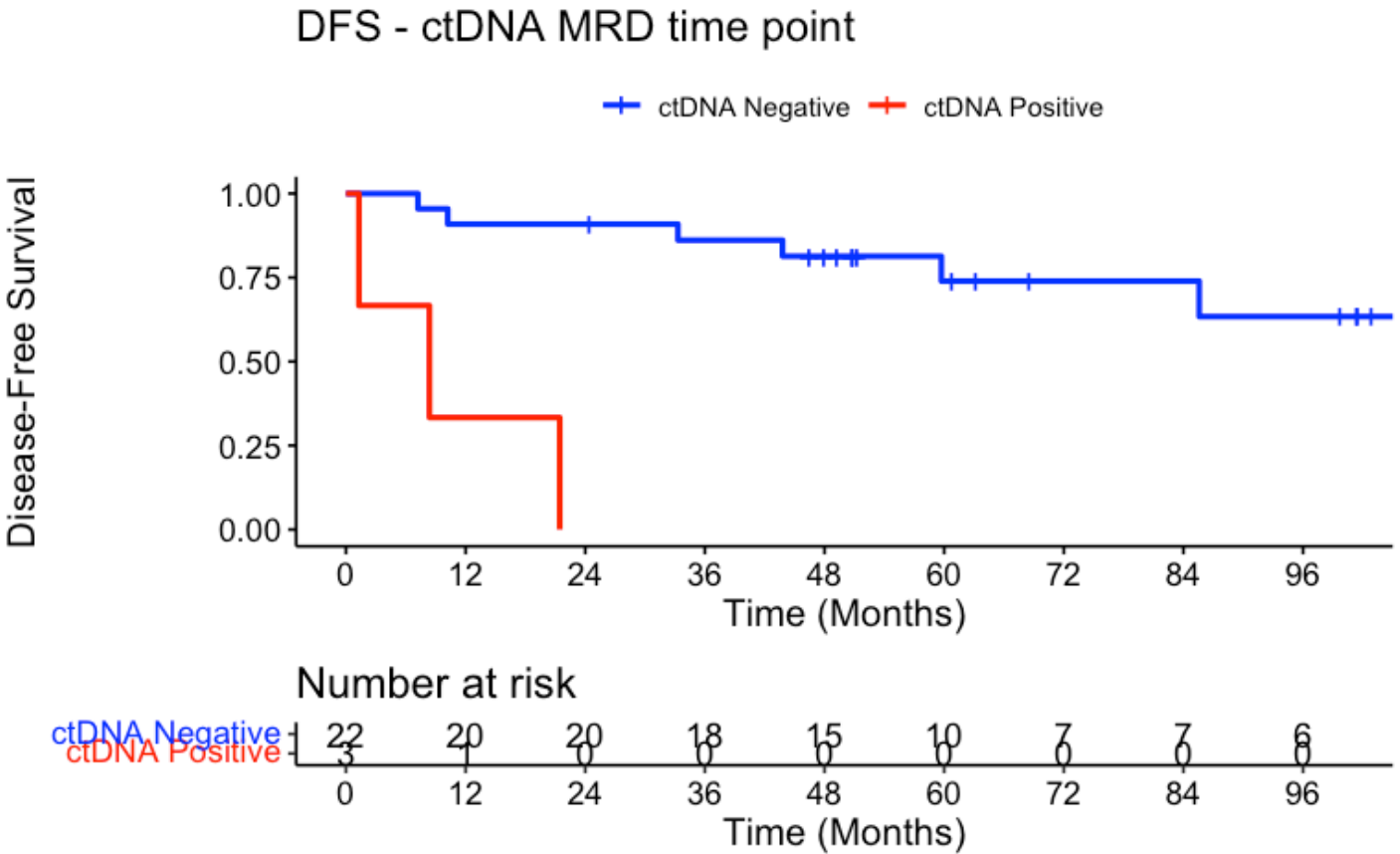
survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.MRD, data =
circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$DFS.months, event = circ_data\$DFS.Event) ~
ctDNA.MRD, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.MRD=NEGATIVE	22	6	NA	85.6	NA
ctDNA.MRD=POSITIVE	3	3	8.37	1.3	NA

Hide

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.MRD, data = circ_data,conf.int=0.95,conf.type="l
og-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRU
E, break.time.by=12, palette=c("blue","red"), title="DFS - ctDNA MRD time point", ylab=
"Disease-Free Survival", xlab="Time (Months)", legend.labs=c("ctDNA Negative", "ctDNA Po
sitive"), legend.title="")
```



Hide

```
summary(KM_curve, times= c(12))
```

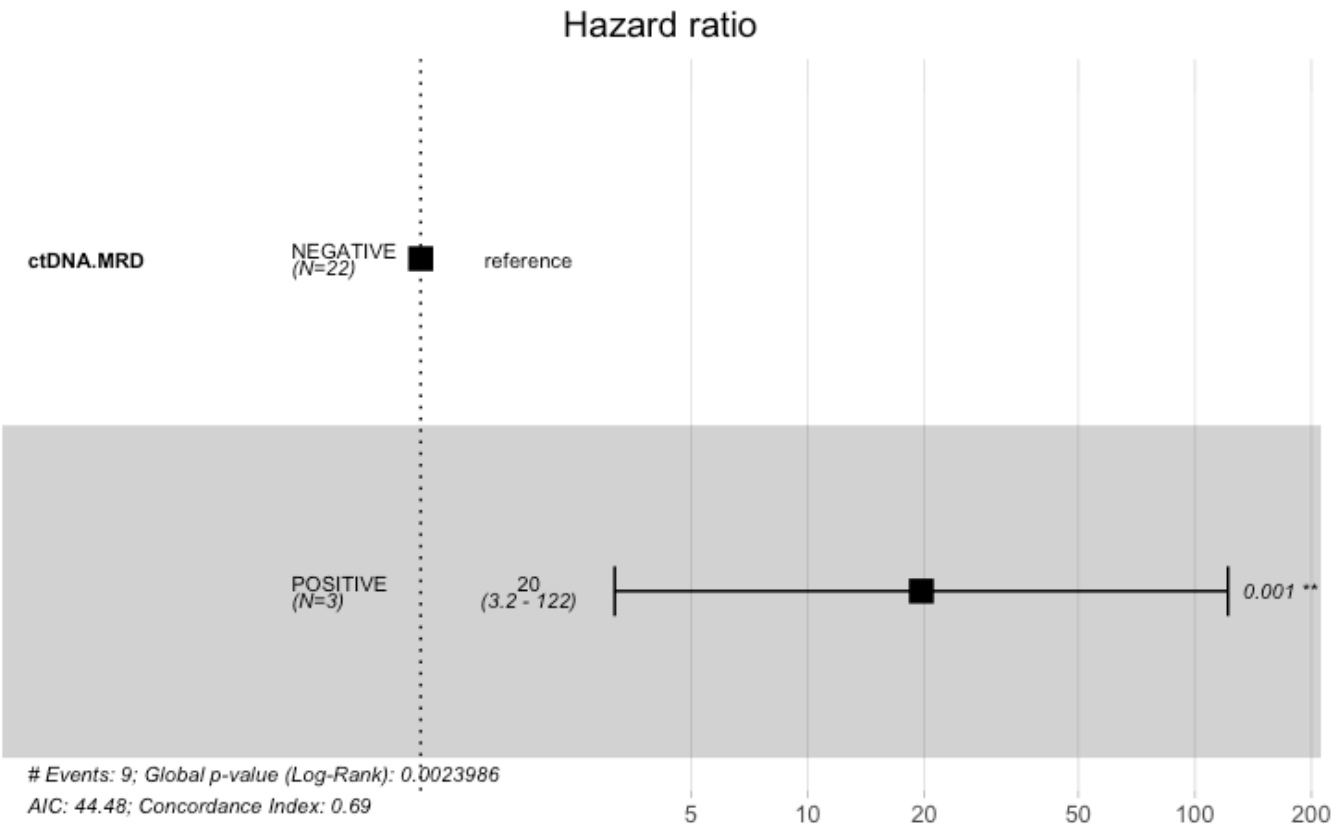
```
Call: survfit(formula = surv_object ~ ctDNA.MRD, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

ctDNA.MRD=NEGATIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
65	12.0000	20.0000	2.0000	0.9091	0.0613	0.6830	0.97

ctDNA.MRD=POSITIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
15	12.00000	1.00000	2.00000	0.33333	0.27217	0.00896	0.774

Hide

```
circ_data$ctDNA.MRD <- factor(circ_data$ctDNA.MRD, levels=c("NEGATIVE","POSITIVE"))
cox_fit <- coxph(surv_object ~ ctDNA.MRD, data=circ_data)
ggforest(cox_fit,data = circ_data)
```



Hide

```
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ ctDNA.MRD, data = circ_data)
```

```
n= 25, number of events= 9
```

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.MRDPOSITIVE	2.9780	19.6490	0.9306	3.2	0.00137 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.MRDPOSITIVE	19.65	0.05089	3.171	121.8

Concordance= 0.686 (se = 0.077)

Likelihood ratio test= 9.22 on 1 df, p=0.002

Wald test = 10.24 on 1 df, p=0.001

Score (logrank) test = 19.56 on 1 df, p=1e-05

[Hide](#)

```
cox_fit_summary <- summary(cox_fit)
```

```
#Extract values for HR, 95% CI, and p-value
```

```
HR <- cox_fit_summary$coefficients[2]
```

```
lower_CI <- cox_fit_summary$conf.int[3]
```

```
upper_CI <- cox_fit_summary$conf.int[4]
```

```
p_value <- cox_fit_summary$coefficients[5]
```

```
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
```

```
print(label_text)
```

```
[1] "HR = 19.65 (3.17-121.75); p = 0.001"
```

[Hide](#)

```
#Fisher plot for ctDNA MRD time point and Rec Status
```

```
circ_data$ctDNA.MRD <- factor(circ_data$ctDNA.MRD, levels = c("NEGATIVE", "POSITIVE"), labels = c("ctDNA(-)", "ctDNA(+)))
```

```
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence"))
```

```
contingency_table <- table(circ_data$ctDNA.MRD, circ_data$DFS.Event)
```

```
chi_square_test <- chisq.test(contingency_table)
```

```
Warning in stats::chisq.test(x, y, ...) :
```

```
Chi-squared approximation may be incorrect
```

[Hide](#)

```
print(chi_square_test)
```

Pearson's Chi-squared test with Yates' continuity correction

```
data: contingency_table
X-squared = 3.3151, df = 1, p-value = 0.06865
```

Hide

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.03652
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 0.8257527      Inf
sample estimates:
odds ratio
      Inf
```

Hide

```
print(contingency_table)
```

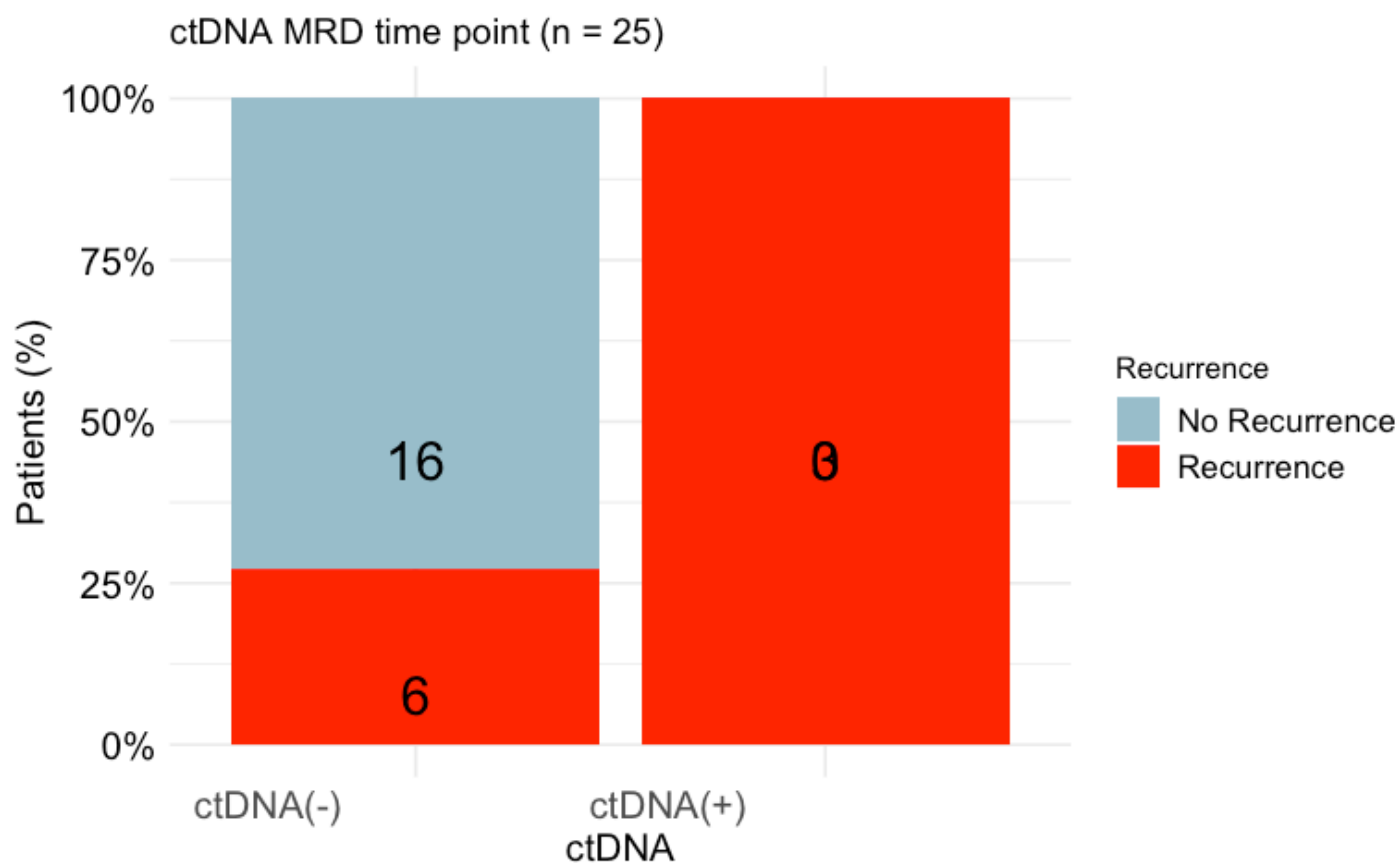
	No Recurrence	Recurrence
ctDNA(-)	16	6
ctDNA(+)	0	3

Hide


```

table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA MRD time point (n = 25)", x = "ctDNA", y = "Patients (%)", fill =
    "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis
    text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size

```



#DFS by ctDNA at post-surgery

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.postop, data = circ_data)
```

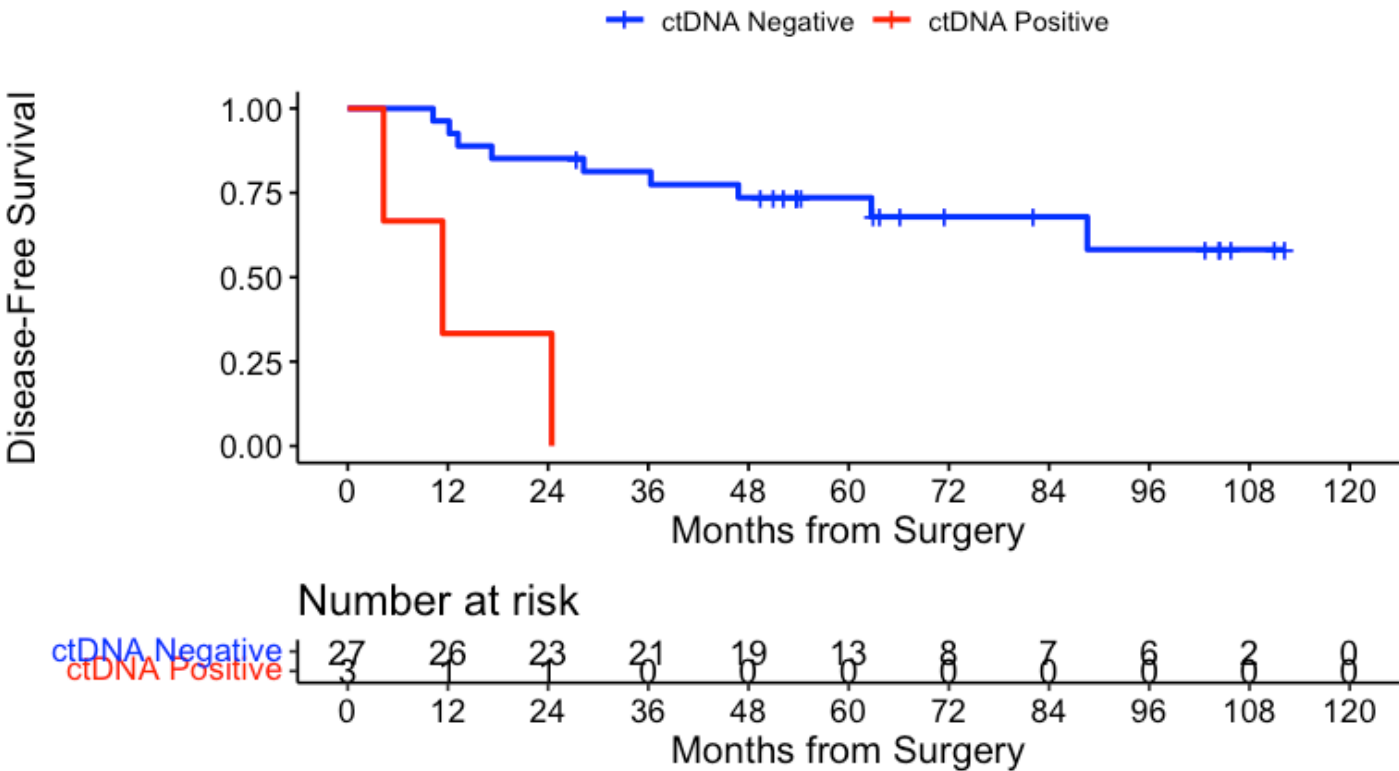
Call: survfit(formula = Surv(time = circ_data\$DFS.months, event = circ_data\$DFS.Event) ~ ctDNA.postop, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.postop=NEGATIVE	27	9	NA	88.6	NA
ctDNA.postop=POSITIVE	3	3	11.4	4.3	NA

[Hide](#)

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.postop, data = circ_data,conf.int=0.95,conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue","red"), title="DFS - ctDNA post-surgery", ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("ctDNA Negative", "ctDNA Positive"), legend.title="")
```

DFS - ctDNA post-surgery



Hide

```
summary(KM_curve, times= c(12))
```

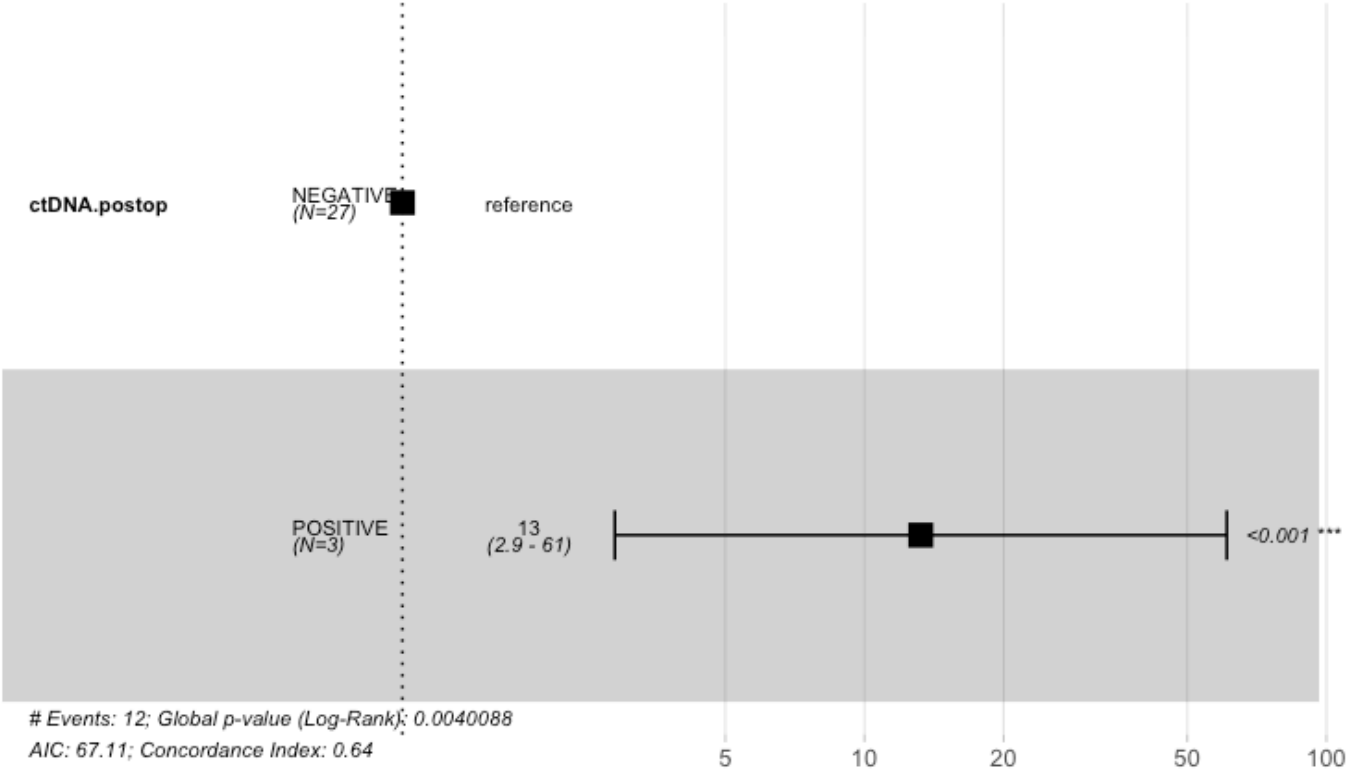
Call: `survfit(formula = surv_object ~ ctDNA.postop, data = circ_data, conf.int = 0.95, conf.type = "log-log")`

ctDNA.postop=NEGATIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
47	12.0000	26.0000	1.0000	0.9630	0.0363	0.7649	0.99
ctDNA.postop=POSITIVE							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
15	12.00000	1.00000	2.00000	0.33333	0.27217	0.00896	0.774

Hide

```
circ_data$ctDNA.postop <- factor(circ_data$ctDNA.postop, levels=c("NEGATIVE", "POSITIVE"))
cox_fit <- coxph(surv_object ~ ctDNA.postop, data=circ_data)
ggforest(cox_fit, data = circ_data)
```

Hazard ratio



Hide

```
summary(cox_fit)
```

Call:
coxph(formula = surv_object ~ ctDNA.postop, data = circ_data)

n= 30, number of events= 12

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.postopPOSITIVE	2.5835	13.2436	0.7781	3.32	0.000899 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.postopPOSITIVE	13.24	0.07551	2.882	60.85

Concordance= 0.637 (se = 0.063)
Likelihood ratio test= 8.28 on 1 df, p=0.004
Wald test = 11.03 on 1 df, p=9e-04
Score (logrank) test = 18.23 on 1 df, p=2e-05

Hide

```
cox_fit_summary <- summary(cox_fit)

#Extract values for HR, 95% CI, and p-value
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 13.24 (2.88-60.85); p = 0.001"
```

Hide

```
#Fisher plot for ctDNA post-surgery and Rec Status
circ_data$ctDNA.postop <- factor(circ_data$ctDNA.postop, levels = c("NEGATIVE", "POSITIVE"), labels = c("ctDNA(-)", "ctDNA(+)))
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels = c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$ctDNA.postop, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

```
Warning in stats::chisq.test(x, y, ...) :
  Chi-squared approximation may be incorrect
```

Hide

```
print(chi_square_test)
```

Pearson's Chi-squared test with Yates' continuity correction

```
data: contingency_table
X-squared = 2.608, df = 1, p-value = 0.1063
```

Hide

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.05419
alternative hypothesis: true odds ratio is not equal to 1
95 percent confidence interval:
 0.6698606      Inf
sample estimates:
odds ratio
      Inf
```

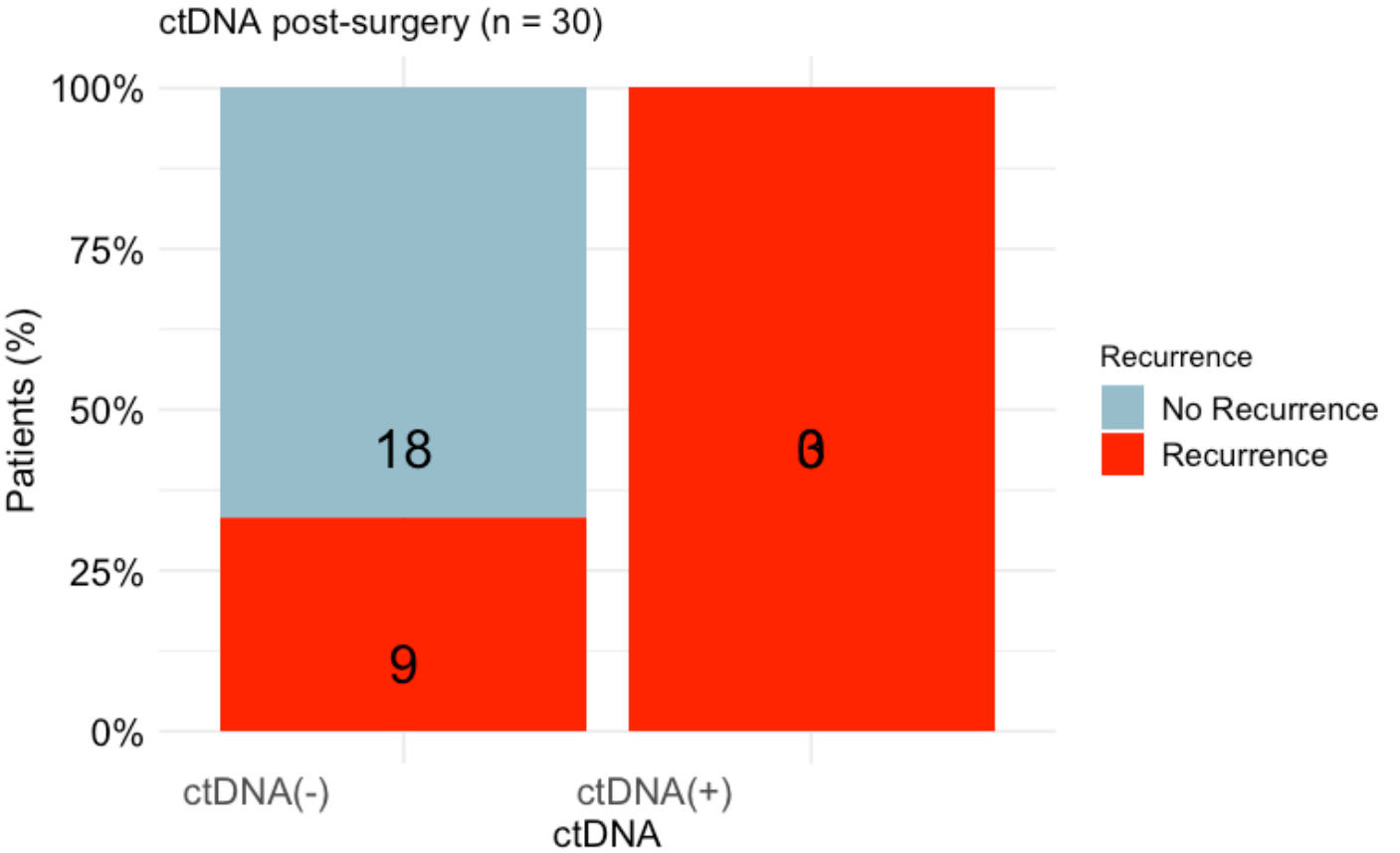
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
ctDNA(-)	18	9
ctDNA(+)	0	3

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black", vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA post-surgery (n = 30)", x = "ctDNA", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
        axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
        axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
        axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
        legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#DFS by NAR score

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~NAR.Status, data
= circ_data)
```

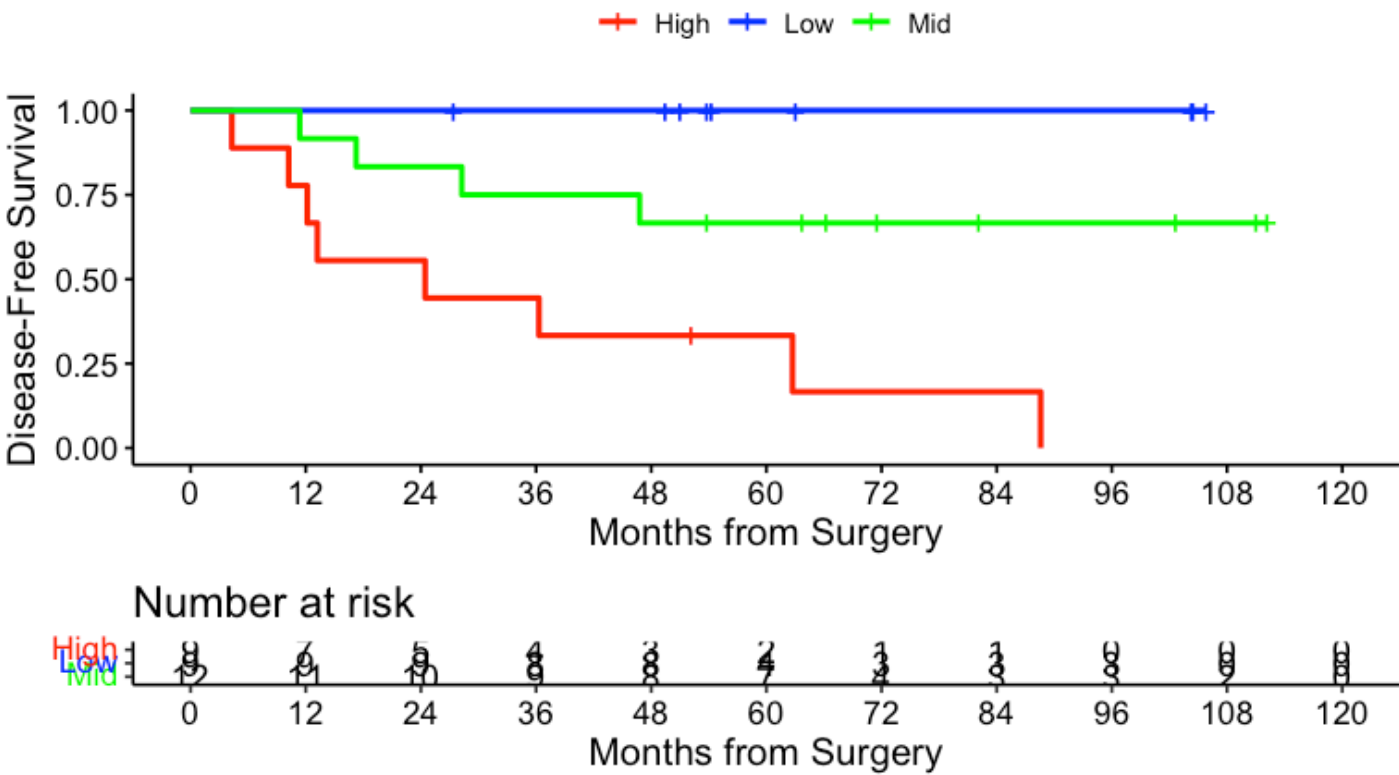
Call: survfit(formula = Surv(time = circ_data\$DFS.months, event = circ_data\$DFS.Event) ~
NAR.Status, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
NAR.Status=High	9	8	24.4	12.2	NA
NAR.Status=Low	9	0	NA	NA	NA
NAR.Status=Mid	12	4	NA	46.8	NA

Hide

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ NAR.Status, data = circ_data, conf.int=0.95, conf.type
="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE,
break.time.by=12, palette=c("red","blue","green"), title="DFS - NAR Score", ylab= "Di
sease-Free Survival", xlab="Months from Surgery", legend.labs=c("High", "Low", "Mid"), l
egend.title="")
```


DFS - NAR Score



Hide

```
summary(KM_curve, times= c(12))
```

Call: `survfit(formula = surv_object ~ NAR.Status, data = circ_data, conf.int = 0.95, conf.type = "log-log")`

NAR.Status=High							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12.000	7.000	2.000	0.778	0.139	0.365	0.9
39							
NAR.Status=Low							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12	9	0	1	0	1	
1							
NAR.Status=Mid							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12.0000	11.0000	1.0000	0.9167	0.0798	0.5390	0.98
78							

Hide

```
circ_data$NAR.Status <- factor(circ_data$NAR.Status, levels=c("Low","Mid","High"))
cox_fit <- coxphf(surv_object ~ NAR.Status, data=circ_data, firth=TRUE, pl=TRUE, maxit=1000)
summary(cox_fit)
```

```
coxphf(formula = surv_object ~ NAR.Status, data = circ_data,
       pl = TRUE, maxit = 1000, firth = TRUE)
```

Model fitted by Penalized ML

Confidence intervals and p-values by Profile Likelihood

	coef	se(coef)	exp(coef)	lower 0.95	upper 0.95	Chisq	p
NAR.StatusMid	1.992185	1.584534	7.331539	0.7811386	972.0415	2.906293	0.0882344495
NAR.StatusHigh	3.388167	1.548959	29.611635	3.6546958	3837.6720	13.490758	0.0002397413

Likelihood ratio test=14.08219 on 2 df, p=0.000875166, n=30

Wald test = 8.552634 on 2 df, p = 0.01389374

Covariance-Matrix:

	NAR.StatusMid	NAR.StatusHigh
NAR.StatusMid	2.510748	2.260077
NAR.StatusHigh	2.260077	2.399275

[Hide](#)

#Fisher plot for NAR Score and Rec Status

```
circ_data$NAR.Status <- factor(circ_data$NAR.Status, levels=c("Low","Mid","High"))
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels =
c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$NAR.Status, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

Warning in stats::chisq.test(x, y, ...) :

Chi-squared approximation may be incorrect

[Hide](#)

```
print(chi_square_test)
```

Pearson's Chi-squared test

data: contingency_table

X-squared = 15.185, df = 2, p-value = 0.0005042

[Hide](#)

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.0002174
alternative hypothesis: two.sided
```

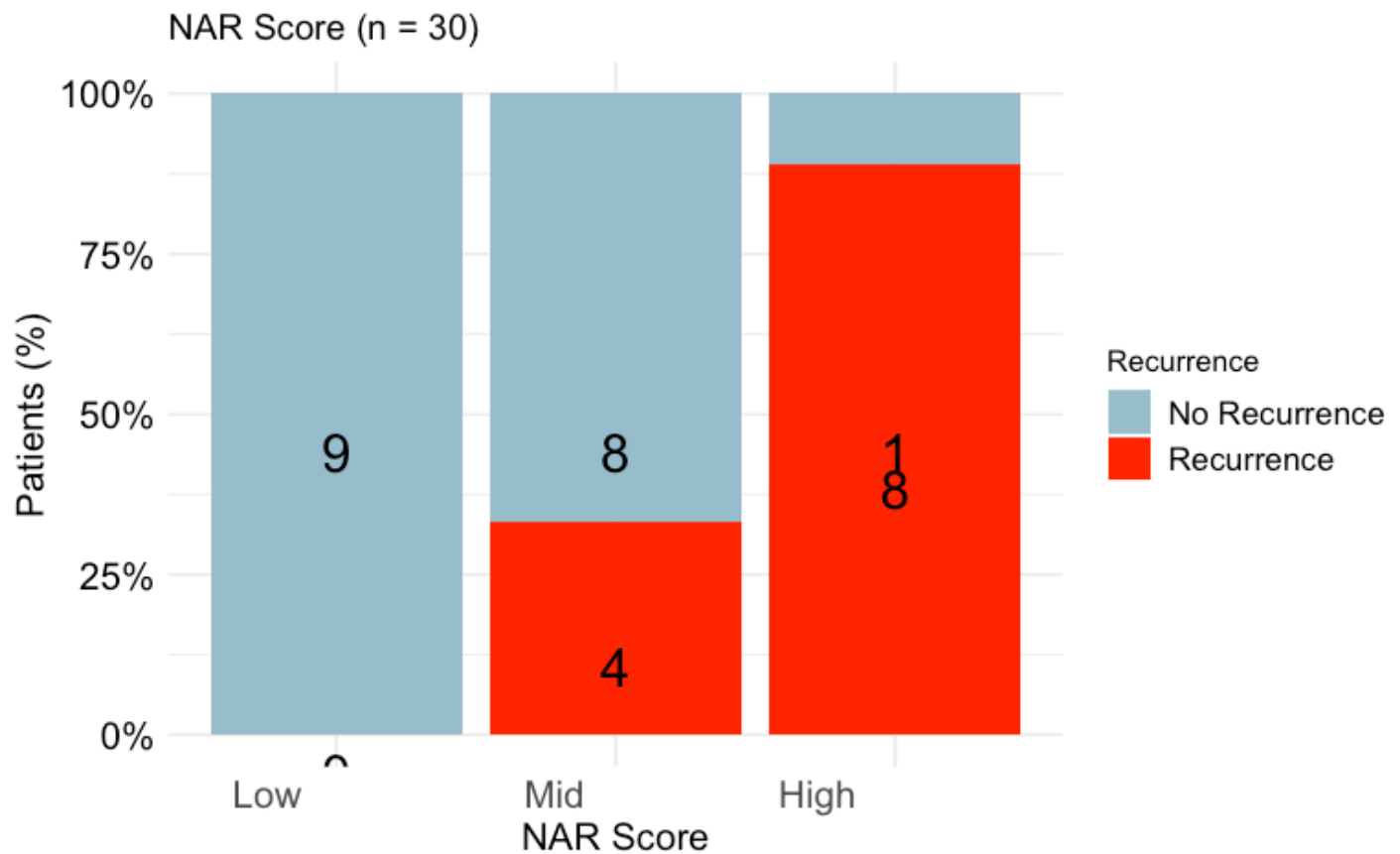
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
Low	9	0
Mid	8	4
High	1	8

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "NAR Score (n = 30)", x = "NAR Score", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis
    text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#Association of ctDNA post-NAT with NAR Score

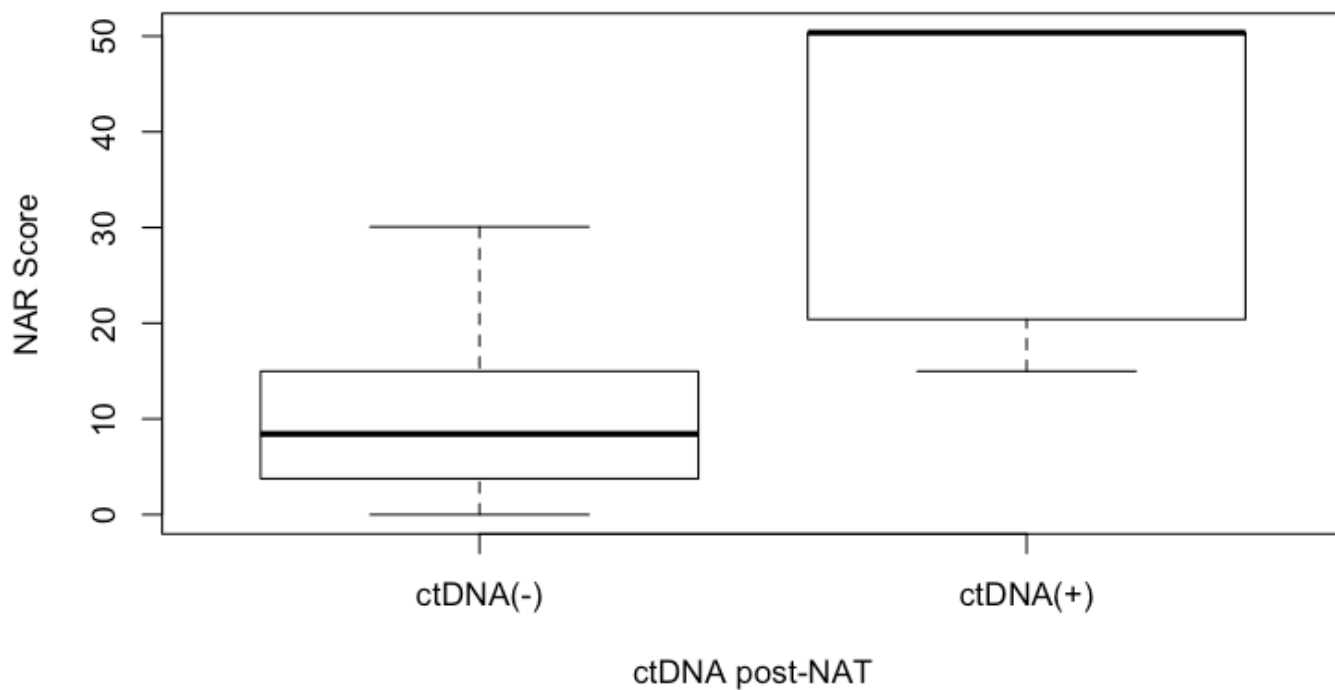
Hide

```
#Boxplot & Fisher exact test for ctDNA post-NAT status with NAR score
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

circ_data$ctDNA.postNAC <- factor(circ_data$ctDNA.postNAC, levels = c("NEGATIVE", "POSITIVE"), labels = c("ctDNA(-)", "ctDNA(+)))
circ_data$NAR.Status <- factor(circ_data$NAR.Status, levels = c("Low","Mid","High"), labels = c("Low","Mid","High"))

boxplot(NAR ~ ctDNA.postNAC, data=circ_data, main="ctDNA post-NAT | NAR Score", xlab="ctDNA post-NAT", ylab="NAR Score", col="white",border="black")
```

ctDNA post-NAT | NAR Score


[Hide](#)

```
m1 <- wilcox.test(NAR ~ ctDNA.postNAC, data=circ_data, na.rm=TRUE, paired=FALSE, exact=FALSE, conf.int=TRUE)
print(m1)
```

Wilcoxon rank sum test with continuity correction

```
data: NAR by ctDNA.postNAC
W = 11.5, p-value = 0.004455
alternative hypothesis: true location shift is not equal to 0
95 percent confidence interval:
 -41.93006 -11.22999
sample estimates:
difference in location
 -29.96004
```

[Hide](#)

```
contingency_table <- table(circ_data$NAR.Status, circ_data$ctDNA.postNAC)
print(contingency_table)
```

	ctDNA(-)	ctDNA(+)
Low	9	0
Mid	11	1
High	5	4

Hide

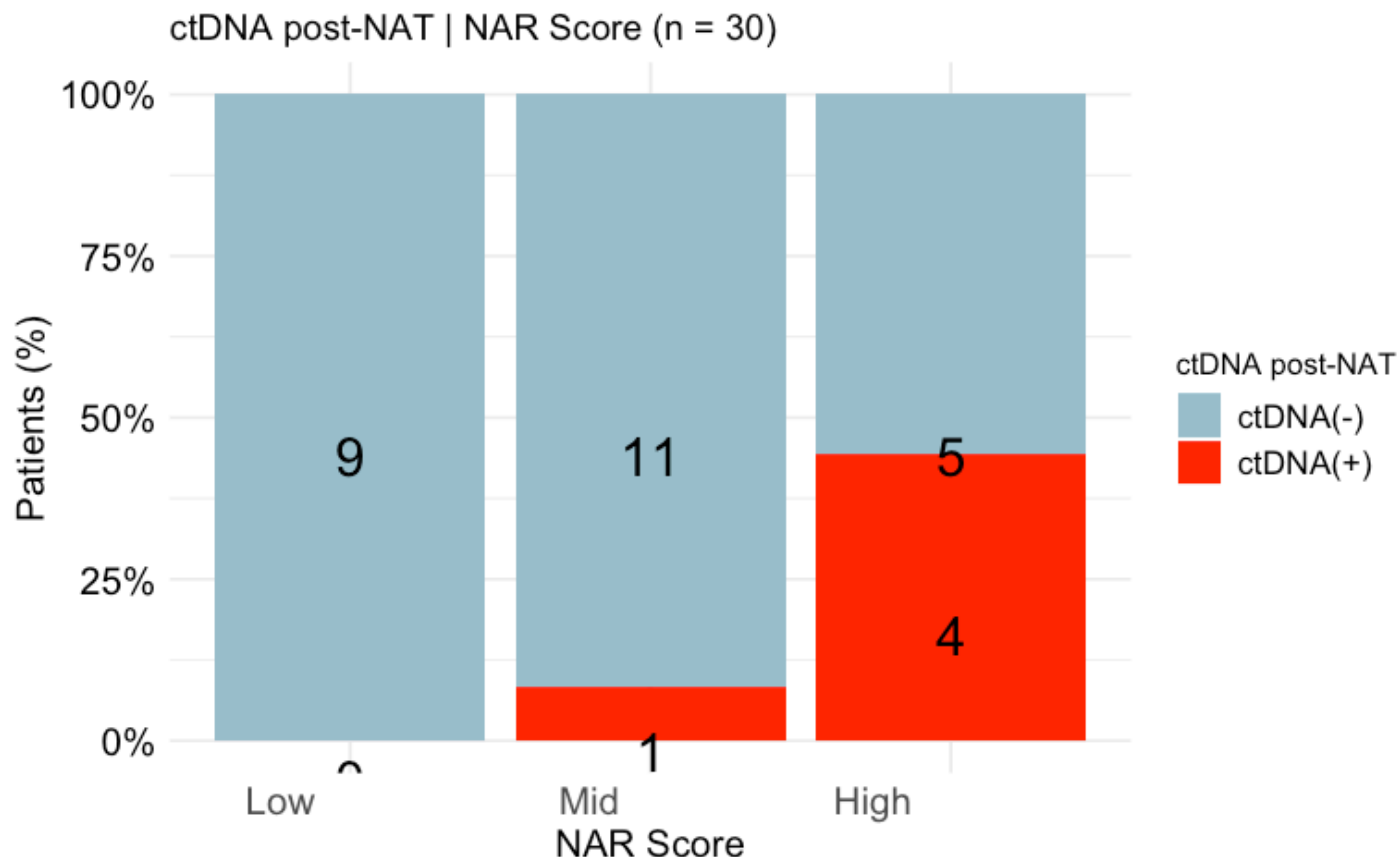
```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.04446
alternative hypothesis: two.sided
```

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA post-NAT | NAR Score (n = 30)", x = "NAR Score", y = "Patients (%)", fill = "ctDNA post-NAT") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("ctDNA(-)" = "lightblue3", "ctDNA(+)" = "red")) + # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#DFS by ctDNA at post-NAT and NAR Score - 3 groups

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

circ_data$ctDNA.postNAT.NAR <- NA #first we create the variable for the ctDNA & NAC combination, and we assign values
circ_data <- circ_data %>%
  mutate(ctDNA.postNAT.NAR = case_when(
    NAR.Score == "Low" & ctDNA.postNAC == "NEGATIVE" ~ 1,
    NAR.Score == "Mid/High" & ctDNA.postNAC == "NEGATIVE" ~ 2,
    NAR.Score == "Mid/High" & ctDNA.postNAC == "POSITIVE" ~ 3
  ))

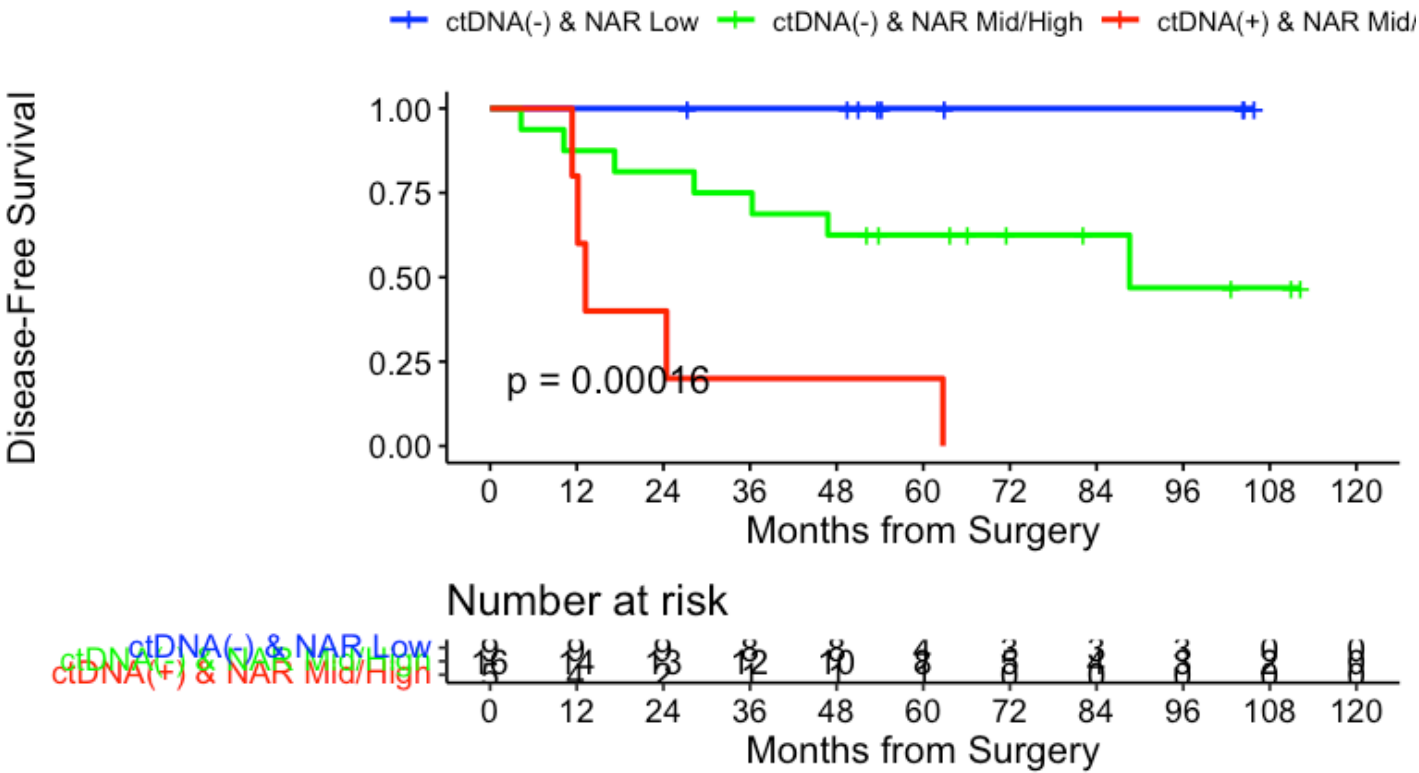
survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.postNAT.NAR, data = circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$DFS.months, event = circ_data\$DFS.Event) ~ ctDNA.postNAT.NAR, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.postNAT.NAR=1	9	0	NA	NA	NA
ctDNA.postNAT.NAR=2	16	7	88.6	36.3	NA
ctDNA.postNAT.NAR=3	5	5	13.2	12.2	NA

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.postNAT.NAR, data = circ_data, conf.int=0.95, conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = TRUE, conf.int = FALSE, risk.table = TRUE,
break.time.by=12, palette=c("blue","green", "red"), title="DFS - ctDNA post-NAT | NAR Score",
ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("ctDNA(-) & NAR Low",
"ctDNA(-) & NAR Mid/High", "ctDNA(+) & NAR Mid/High"), legend.title="")
```


DFS - ctDNA post-NAT | NAR Score



Hide

```
summary(KM_curve, times= c(36))
```

```
Call: survfit(formula = surv_object ~ ctDNA.postNAT.NAR, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

ctDNA.postNAT.NAR=1							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
NA	36	8	0	1	0	NA	

ctDNA.postNAT.NAR=2							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
98	36.000	12.000	4.000	0.750	0.108	0.463	0.8

ctDNA.postNAT.NAR=3							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
85	36.00000	1.00000	4.00000	0.20000	0.17889	0.00837	0.581

Hide

```
circ_data$ctDNA.postNAT.NAR <- factor(circ_data$ctDNA.postNAT.NAR, levels=c("1", "2",
"3"), labels = c("ctDNA(-) & NAR Low", "ctDNA(-) & NAR Mid/High", "ctDNA(+) & NAR Mid/Hi
gh"))
cox_fit <- coxphf(surv_object ~ ctDNA.postNAT.NAR, data=circ_data)
summary(cox_fit)
```

```
coxphf(formula = surv_object ~ ctDNA.postNAT.NAR, data = circ_data)
```

Model fitted by Penalized ML

Confidence intervals and p-values by Profile Likelihood

			coef	se(coef)	exp(coef)	lower 0.95	upper 0.
95	Chisq	p					
ctDNA.postNAT.NAR	ctDNA(-) & NAR Mid/High	2.278706	1.549400	9.764036	1.188733	1267.6	
19	4.700723	0.0301499343					
ctDNA.postNAT.NAR	ctDNA(+) & NAR Mid/High	3.861533	1.585868	47.538187	5.087168	6337.2	
80	14.225859	0.0001621272					

Likelihood ratio test=14.22784 on 2 df, p=0.0008136994, n=30

Wald test = 9.637369 on 2 df, p = 0.008077406

Covariance-Matrix:

	ctDNA.postNAT.NAR	ctDNA(-) & NAR Mid/High	ctDNA.
postNAT.NAR	ctDNA(+) & NAR Mid/High		
ctDNA.postNAT.NAR	ctDNA(-) & NAR Mid/High	2.400641	
2.255337			
ctDNA.postNAT.NAR	ctDNA(+) & NAR Mid/High	2.255337	
2.514977			

[Hide](#)

```
#Fisher plot for ctDNA post-NAT and NAR Score combination with Rec Status
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels =
c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$ctDNA.postNAT.NAR, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

Warning in stats::chisq.test(x, y, ...) :
Chi-squared approximation may be incorrect

[Hide](#)

```
print(chi_square_test)
```

Pearson's Chi-squared test

```
data: contingency_table
X-squared = 13.594, df = 2, p-value = 0.001117
```

Hide

```
fisher_exact_test <- fisher.test(contingency_table)
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.00036
alternative hypothesis: two.sided
```

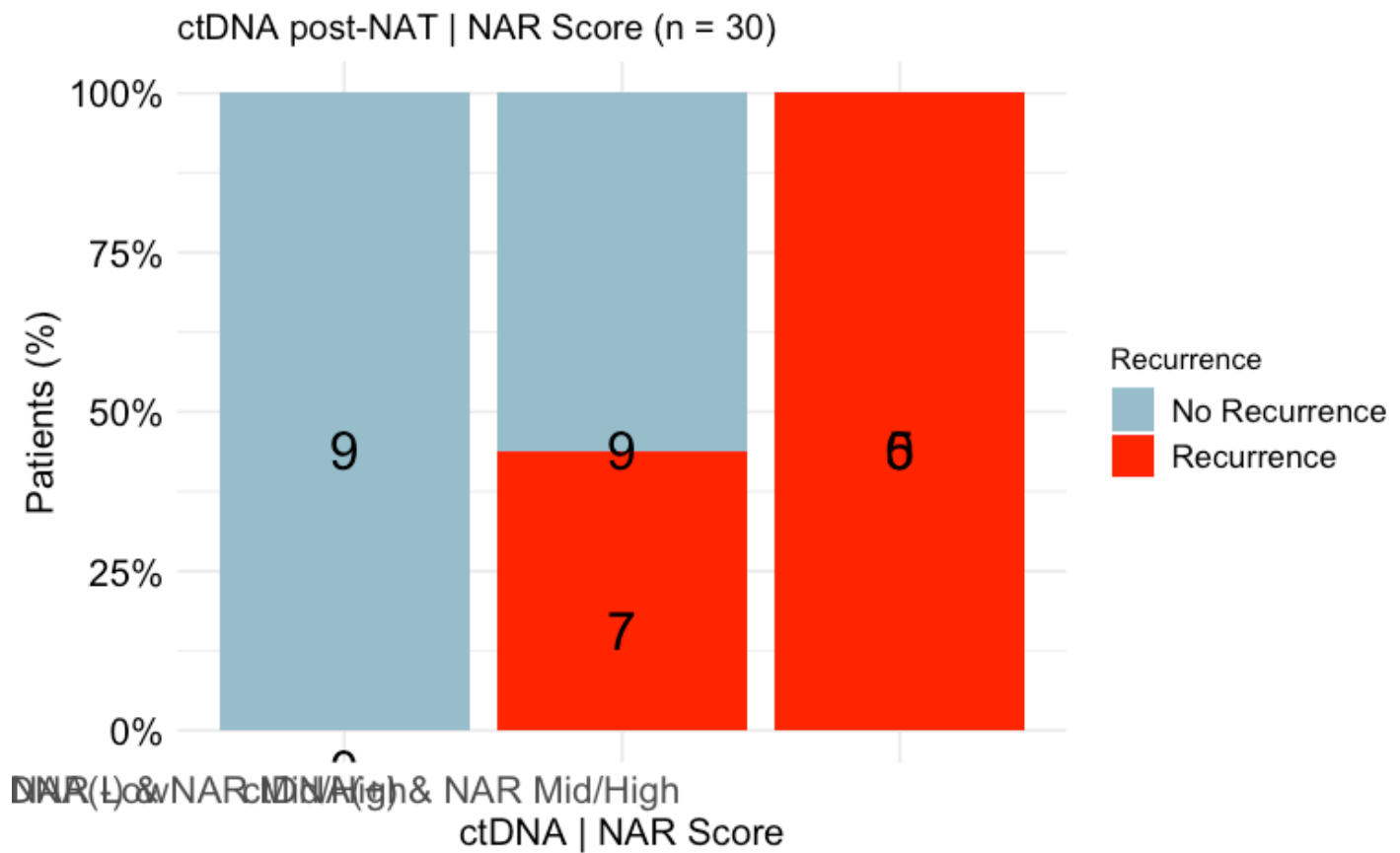
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
ctDNA(-) & NAR Low	9	0
ctDNA(-) & NAR Mid/High	9	7
ctDNA(+) & NAR Mid/High	0	5

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA post-NAT | NAR Score (n = 30)", x = "ctDNA | NAR Score", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```

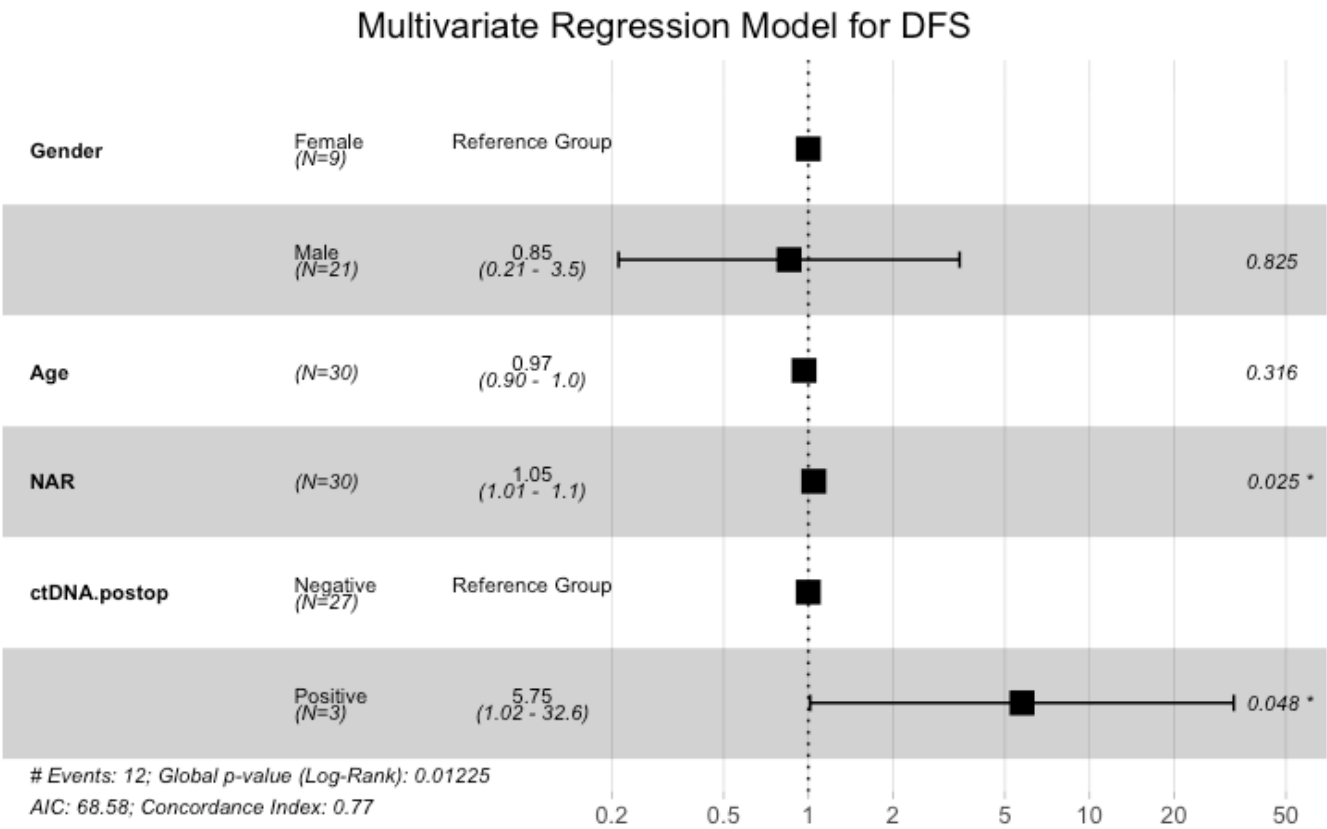


#Multivariate regression model for DFS

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)

circ_datadf$Gender <- factor(circ_datadf$Gender, levels = c("Female", "Male"), labels =
c("Female", "Male"))
circ_datadf$ctDNA.postop <- factor(circ_datadf$ctDNA.postop, levels=c("NEGATIVE","POSITI
VE"), labels = c("Negative", "Positive"))
surv_object<-Surv(time = circ_datadf$DFS.months, event = circ_datadf$DFS.Event)
cox_fit <- coxph(surv_object ~ Gender + Age + NAR + ctDNA.postop, data=circ_datadf)
ggforest(cox_fit, data = circ_datadf, main = "Multivariate Regression Model for DFS", re
fLabel = "Reference Group")
```



Hide

```
test.ph <- cox.zph(cox_fit)
```

#DFS by ctDNA at post-NAT and pCR - 3 groups

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

circ_data$ctDNA.pCR <- NA
circ_data <- circ_data %>%
  mutate(ctDNA.pCR = case_when(
    ctDNA.postNAC == "NEGATIVE" & pCR == "TRUE" ~ "1",
    ctDNA.postNAC == "NEGATIVE" & pCR == "FALSE" ~ "2",
    ctDNA.postNAC == "POSITIVE" & pCR == "FALSE" ~ "3",
  ))

survfit(Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)~ctDNA.pCR, data =
circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$RFS.months, event = circ_data\$RFS.Event) ~
ctDNA.pCR, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.pCR=1	6	0	NA	NA	NA
ctDNA.pCR=2	19	3	NA	NA	NA
ctDNA.pCR=3	5	5	13.2	12.2	NA

Hide

```
circ_data <- circ_data[!is.na(circ_data$ctDNA.pCR),]
circ_data$ctDNA.pCR <- factor(circ_data$ctDNA.pCR, levels=c("1","2", "3"), labels=c("pCR
ctDNA (-)","No pCR ctDNA (-)", "No pCR ctDNA (+)"))
survfit(Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)~ctDNA.pCR, data =
circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$RFS.months, event = circ_data\$RFS.Event) ~
ctDNA.pCR, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.pCR=pCR ctDNA (-)	6	0	NA	NA	NA
ctDNA.pCR=No pCR ctDNA (-)	19	3	NA	NA	NA
ctDNA.pCR=No pCR ctDNA (+)	5	5	13.2	12.2	NA

Hide

```
surv_object <-Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.pCR, data = circ_data,conf.int=0.95,conf.type="l
og-log")
summary(KM_curve, times= c(36))
```

Call: `survfit(formula = surv_object ~ ctDNA.pCR, data = circ_data, conf.int = 0.95, conf.type = "log-log")`

CI	time	ctDNA.pCR=pCR	ctDNA (-)	survival	std.err	lower 95% CI	upper 95%
		n.risk	n.event				
NA	36	5	0	1	0	NA	

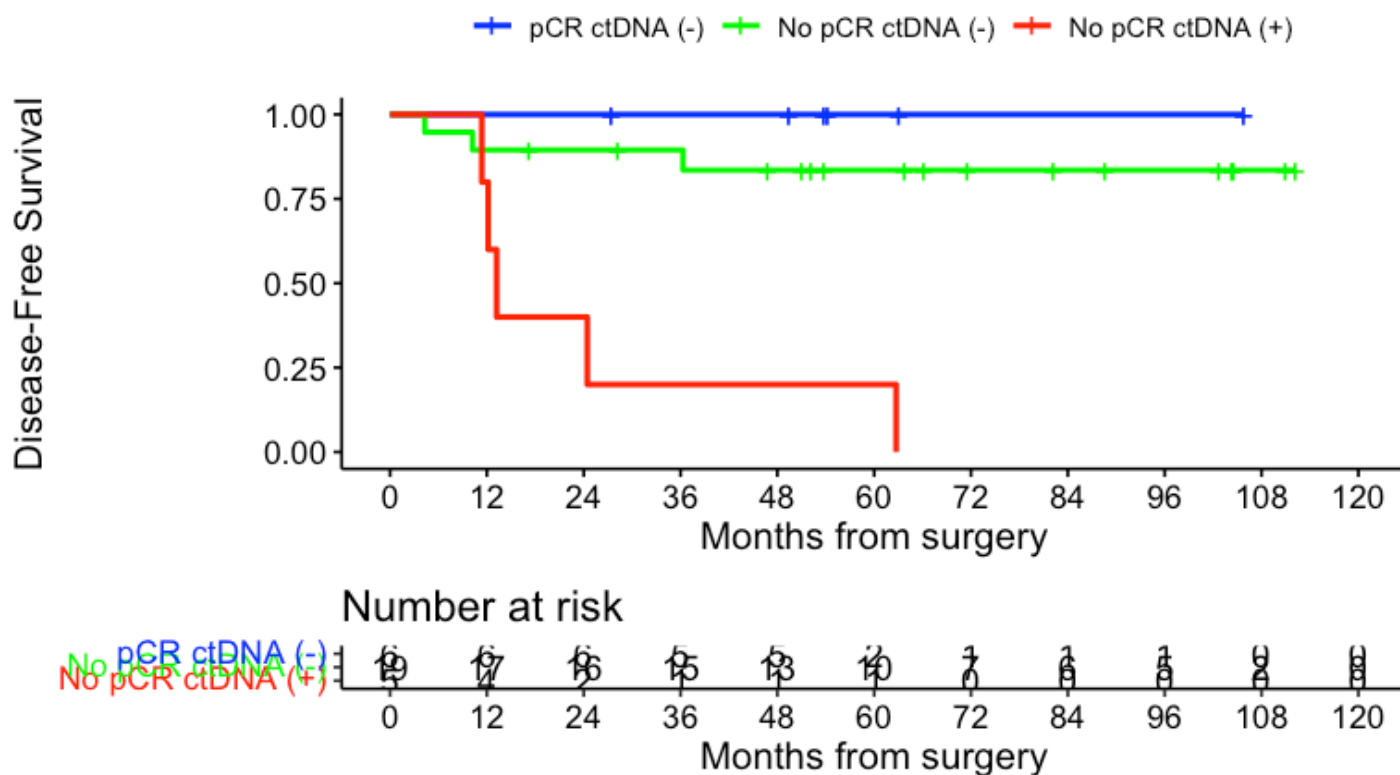
CI	time	ctDNA.pCR=No pCR	ctDNA (-)	survival	std.err	lower 95% CI	upper 95%
		n.risk	n.event				
26	36.0000	15.0000	2.0000	0.8947	0.0704	0.6408	0.97

CI	time	ctDNA.pCR=No pCR	ctDNA (+)	survival	std.err	lower 95% CI	upper 95%
		n.risk	n.event				
85	36.00000	1.00000	4.00000	0.20000	0.17889	0.00837	0.581

Hide

`ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue", "green", "red"), title="DFS - ctDNA post-NAC & pCR", ylab= "Disease-Free Survival", xlab="Months from surgery", legend.labs=c("pCR ctDNA (-)","No pCR ctDNA (-)", "No pCR ctDNA (+)"), legend.title="")`

DFS - ctDNA post-NAC & pCR



Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

circ_data$ctDNA.pCR <- NA
circ_data <- circ_data %>%
  mutate(ctDNA.pCR = case_when(
    ctDNA.postNAC == "NEGATIVE" & pCR == "TRUE" ~ "1",
    ctDNA.postNAC == "NEGATIVE" & pCR == "FALSE" ~ "2",
    ctDNA.postNAC == "POSITIVE" & pCR == "FALSE" ~ "3"
  ))
circ_data$ctDNA.pCR <- factor(circ_data$ctDNA.pCR, levels=c("1","2","3"))
surv_object <- Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)
cox_fit <- coxphf(surv_object ~ ctDNA.pCR, data=circ_data)
summary(cox_fit)
```



```
coxphf(formula = surv_object ~ ctDNA.pCR, data = circ_data)
```

Model fitted by Penalized ML

Confidence intervals and p-values by Profile Likelihood

	coef	se(coef)	exp(coef)	lower 0.95	upper 0.95	Chisq	p
ctDNA.pCR2	0.8115447	1.653795	2.251383	0.2171987	303.1025	0.3498855	0.554177937
ctDNA.pCR3	3.0528823	1.623237	21.176292	2.3228317	2803.8027	8.4378266	0.003674952

Likelihood ratio test=11.84993 on 2 df, p=0.002671899, n=30

Wald test = 10.29557 on 2 df, p = 0.005812257

Covariance-Matrix:

	ctDNA.pCR2	ctDNA.pCR3
ctDNA.pCR2	2.735037	2.390062
ctDNA.pCR3	2.390062	2.634899

#DFS by ctDNA at MRD Timepoint and pCR - 3 groups

[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.MRD!="",]
circ_data$DFS.months=circ_data$DFS.months-3
circ_data <- circ_data[circ_data$DFS.months>=0,]

circ_data$ctDNA.pCR <- NA
circ_data <- circ_data %>%
  mutate(ctDNA.pCR = case_when(
    ctDNA.MRD == "NEGATIVE" & pCR == "TRUE" ~ "1",
    ctDNA.MRD == "NEGATIVE" & pCR == "FALSE" ~ "2",
    ctDNA.MRD == "POSITIVE" & pCR == "FALSE" ~ "3",
  ))

survfit(Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)~ctDNA.pCR, data =
circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$RFS.months, event = circ_data\$RFS.Event) ~
ctDNA.pCR, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.pCR=1	5	0	NA	NA	NA
ctDNA.pCR=2	17	4	NA	NA	NA
ctDNA.pCR=3	3	3	11.4	4.3	NA

[Hide](#)

```
circ_data <- circ_data[!is.na(circ_data$ctDNA.pCR),]
circ_data$ctDNA.pCR <- factor(circ_data$ctDNA.pCR, levels=c("1","2", "3"), labels=c("pCR
ctDNA (-)","No pCR ctDNA (-)", "No pCR ctDNA (+)"))
survfit(Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)~ctDNA.pCR, data =
circ_data)
```

Call: survfit(formula = Surv(time = circ_data\$RFS.months, event = circ_data\$RFS.Event) ~
ctDNA.pCR, data = circ_data)

	n	events	median	0.95LCL	0.95UCL
ctDNA.pCR=pCR ctDNA (-)	5	0	NA	NA	NA
ctDNA.pCR=No pCR ctDNA (-)	17	4	NA	NA	NA
ctDNA.pCR=No pCR ctDNA (+)	3	3	11.4	4.3	NA

Hide

```
surv_object <-Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.pCR, data = circ_data,conf.int=0.95,conf.type="l
og-log")
summary(KM_curve, times= c(24))
```

Call: survfit(formula = surv_object ~ ctDNA.pCR, data = circ_data,
conf.int = 0.95, conf.type = "log-log")

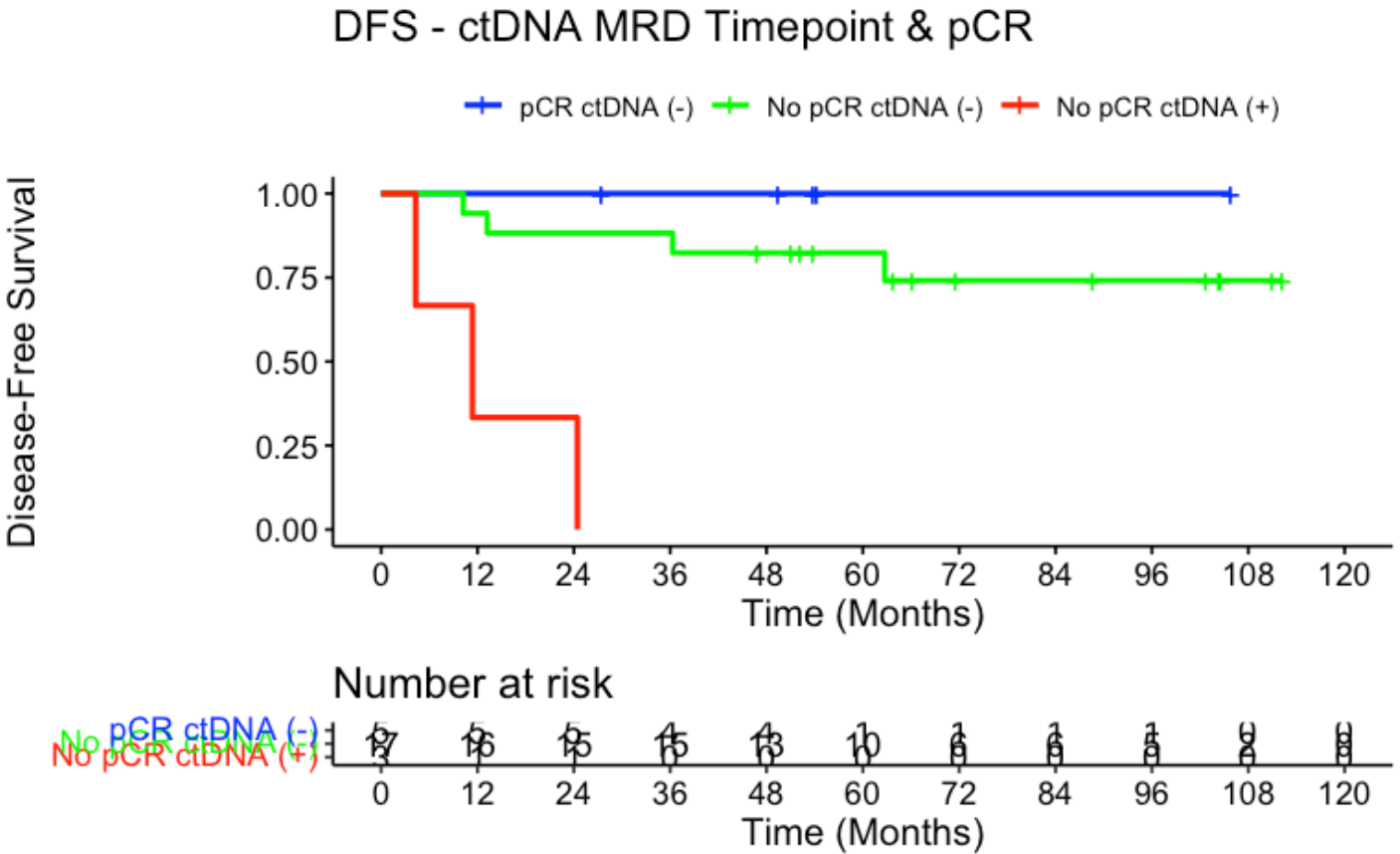
	time	ctDNA.pCR=pCR ctDNA (-) n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
1	24	5	0	1	0	1	

	time	ctDNA.pCR=No pCR ctDNA (-) n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
92	24.0000	15.0000	2.0000	0.8824	0.0781	0.6060	0.96

	time	ctDNA.pCR=No pCR ctDNA (+) n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI							
15	24.00000	1.00000	2.00000	0.33333	0.27217	0.00896	0.774

Hide

```
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRU
E, break.time.by=12, palette=c("blue", "green", "red"), title="DFS - ctDNA MRD Timepoint
& pCR", ylab= "Disease-Free Survival", xlab="Time (Months)", legend.labs=c("pCR ctDNA
(-)","No pCR ctDNA (-)", "No pCR ctDNA (+)"), legend.title="")
```



Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$ctDNA.MRD!="",]
circ_data$DFS.months=circ_data$DFS.months-3
circ_data <- circ_data[circ_data$DFS.months>=0,]

circ_data$ctDNA.pCR <- NA
circ_data <- circ_data %>%
  mutate(ctDNA.pCR = case_when(
    ctDNA.MRD == "NEGATIVE" & pCR == "TRUE" ~ "1",
    ctDNA.MRD == "NEGATIVE" & pCR == "FALSE" ~ "2",
    ctDNA.MRD == "POSITIVE" & pCR == "FALSE" ~ "3"
  ))
circ_data$ctDNA.pCR <- factor(circ_data$ctDNA.pCR, levels=c("1","2","3"))
surv_object <- Surv(time = circ_data$RFS.months, event = circ_data$RFS.Event)
cox_fit <- coxphf(surv_object ~ ctDNA.pCR, data=circ_data)
summary(cox_fit)
```

```
coxphf(formula = surv_object ~ ctDNA.pCR, data = circ_data)
```

Model fitted by Penalized ML

Confidence intervals and p-values by Profile Likelihood

	coef	se(coef)	exp(coef)	lower 0.95	upper 0.95	Chisq	p
ctDNA.pCR2	0.814095	1.639885	2.257132	0.2317876	301.287	0.3620868	0.547349552
ctDNA.pCR3	3.462331	1.709254	31.891234	2.7928253	4445.788	8.4316797	0.003687396

Likelihood ratio test=10.24975 on 2 df, p=0.005946952, n=25

Wald test = 9.203581 on 2 df, p = 0.01003385

Covariance-Matrix:

	ctDNA.pCR2	ctDNA.pCR3
ctDNA.pCR2	2.689222	2.364747
ctDNA.pCR3	2.364747	2.921551

#Univariate regression models for DFS for the factors used in the MVA

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)
surv_object <- Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
circ_datadf$Gender <- factor(circ_datadf$Gender, levels = c("Female", "Male"), labels =
c("Female", "Male")) #univariate for gender
cox_fit <- coxph(surv_object ~ Gender, data=circ_data)
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ Gender, data = circ_data)
```

```
n= 30, number of events= 12
```

	coef	exp(coef)	se(coef)	z	Pr(> z)
GenderMale	0.2140	1.2386	0.6675	0.321	0.749

	exp(coef)	exp(-coef)	lower .95	upper .95
GenderMale	1.239	0.8074	0.3348	4.583

Concordance= 0.506 (se = 0.073)

Likelihood ratio test= 0.11 on 1 df, p=0.7

Wald test = 0.1 on 1 df, p=0.7

Score (logrank) test = 0.1 on 1 df, p=0.7

[Hide](#)

```
cox_fit_summary <- summary(cox_fit)
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 1.24 (0.33-4.58); p = 0.749"
```

[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)
surv_object <- Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
cox_fit <- coxph(surv_object ~ Age, data=circ_data) #univariate for age
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ Age, data = circ_data)
```

```
n= 30, number of events= 12
```

	coef	exp(coef)	se(coef)	z	Pr(> z)
Age	0.01041	1.01047	0.03617	0.288	0.773

	exp(coef)	exp(-coef)	lower .95	upper .95
Age	1.01	0.9896	0.9413	1.085

Concordance= 0.575 (se = 0.115)

Likelihood ratio test= 0.08 on 1 df, p=0.8

Wald test = 0.08 on 1 df, p=0.8

Score (logrank) test = 0.08 on 1 df, p=0.8

[Hide](#)

```
cox_fit_summary <- summary(cox_fit)
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 1.01 (0.94-1.08); p = 0.773"
```

[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)
surv_object <- Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
cox_fit <- coxph(surv_object ~ NAR, data=circ_data) #univariate for NAR
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ NAR, data = circ_data)
```

```
n= 30, number of events= 12
```

```
      coef exp(coef) se(coef)      z Pr(>|z|)
NAR 0.05036   1.05165   0.01579 3.19  0.00142 **
```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
      exp(coef) exp(-coef) lower .95 upper .95
NAR      1.052      0.9509      1.02      1.085
```

```
Concordance= 0.788 (se = 0.051 )
```

```
Likelihood ratio test= 8.99 on 1 df,  p=0.003
```

```
Wald test              = 10.18 on 1 df,  p=0.001
```

```
Score (logrank) test = 12.41 on 1 df,  p=4e-04
```

[Hide](#)

```
cox_fit_summary <- summary(cox_fit)
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 1.05 (1.02-1.08); p = 0.001"
```

[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_datadf <- as.data.frame(circ_data)
surv_object <- Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
circ_data$ctDNA.postop <- factor(circ_data$ctDNA.postop, levels=c("NEGATIVE", "POSITIVE"))
cox_fit <- coxph(surv_object ~ ctDNA.postop, data=circ_data) #univariate for ctDNA post-surgery
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ ctDNA.postop, data = circ_data)
```

```
n= 30, number of events= 12
```

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.postopPOSITIVE	2.5835	13.2436	0.7781	3.32	0.000899 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.postopPOSITIVE	13.24	0.07551	2.882	60.85

Concordance= 0.637 (se = 0.063)

Likelihood ratio test= 8.28 on 1 df, p=0.004

Wald test = 11.03 on 1 df, p=9e-04

Score (logrank) test = 18.23 on 1 df, p=2e-05

[Hide](#)

```
cox_fit_summary <- summary(cox_fit)
HR <- cox_fit_summary$coefficients[2]
lower_CI <- cox_fit_summary$conf.int[3]
upper_CI <- cox_fit_summary$conf.int[4]
p_value <- cox_fit_summary$coefficients[5]
label_text <- paste0("HR = ", round(HR, 2), " (", round(lower_CI, 2), "-", round(upper_CI, 2), "); p = ", round(p_value, 3))
print(label_text)
```

```
[1] "HR = 13.24 (2.88-60.85); p = 0.001"
```

#DFS by ctDNA Dynamics post-NAT to post-surgery - 3 groups

[Hide](#)

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")

circ_data$ctDNA.Dynamics <- NA #first we create the variable
circ_data <- circ_data %>%
  mutate(ctDNA.Dynamics = case_when(
    ctDNA.postNAC == "NEGATIVE" & ctDNA.postop == "NEGATIVE" ~ 1,
    ctDNA.postNAC == "POSITIVE" & ctDNA.postop == "NEGATIVE" ~ 2,
    ctDNA.postNAC == "POSITIVE" & ctDNA.postop == "POSITIVE" ~ 3))

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.Dynamics, data = circ_data)
```



```
Call: survfit(formula = Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event) ~
      ctDNA.Dynamics, data = circ_data)
```

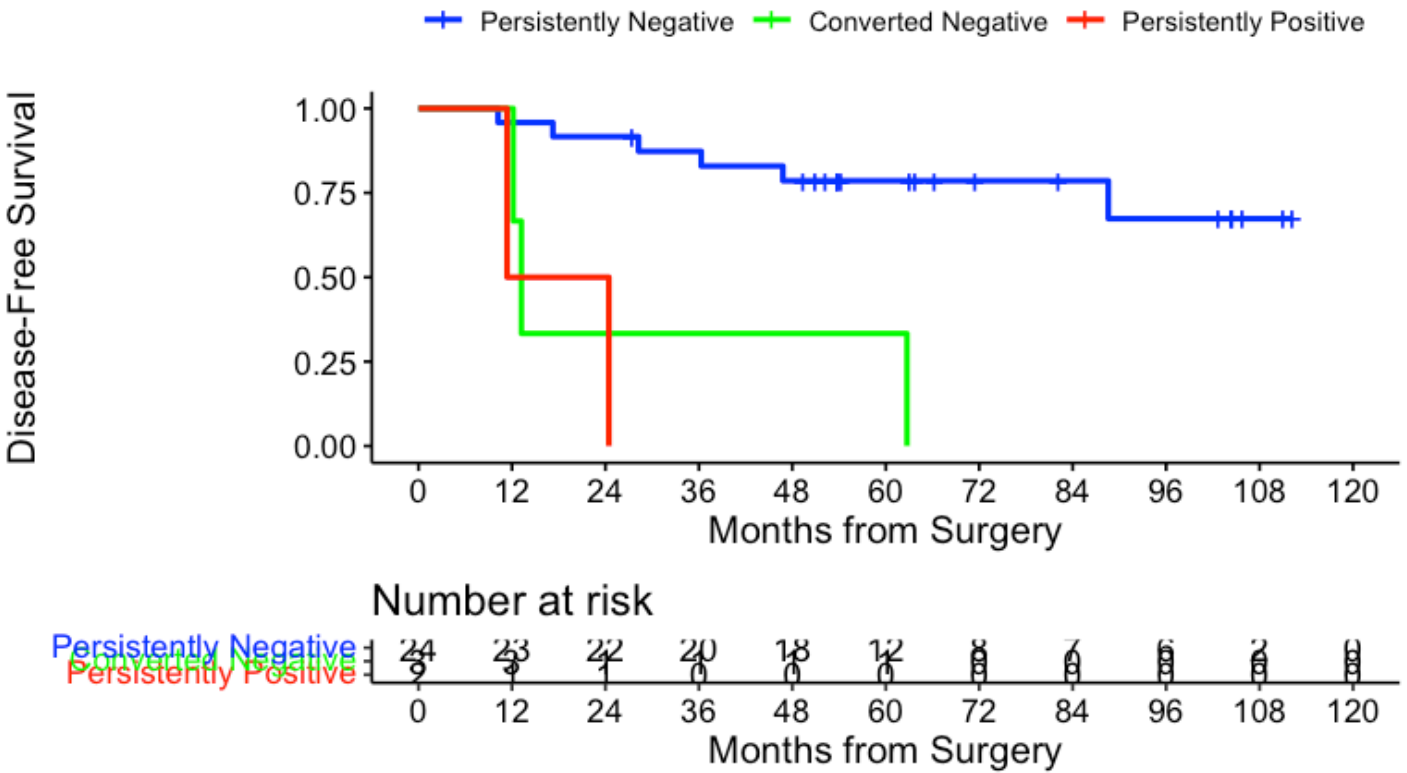
1 observation deleted due to missingness

	n	events	median	0.95LCL	0.95UCL
ctDNA.Dynamics=1	24	6	NA	88.6	NA
ctDNA.Dynamics=2	3	3	13.2	12.2	NA
ctDNA.Dynamics=3	2	2	17.9	11.4	NA

[Hide](#)

```
surv_object <-Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.Dynamics, data = circ_data,conf.int=0.95,conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue","green", "red"), title="DFS - ctDNA Dynamics post-NAT to post-surgery", ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("Persistently Negative", "Converted Negative","Persistently Positive"), legend.title="")
```

DFS - ctDNA Dynamics post-NAT to post-surgery



Hide

```
summary(KM_curve, times= c(12))
```

```
Call: survfit(formula = surv_object ~ ctDNA.Dynamics, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

1 observation deleted due to missingness

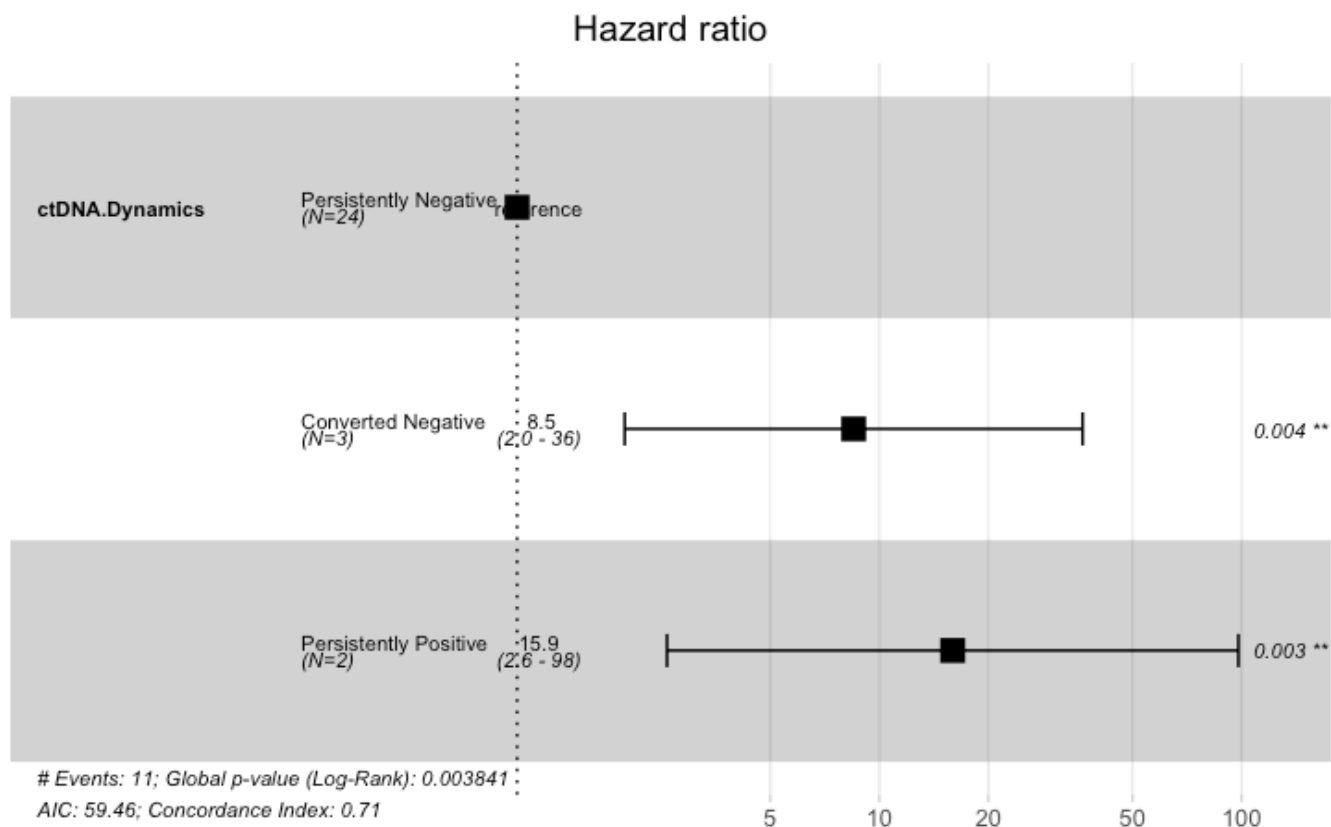
ctDNA.Dynamics=1							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12.0000	23.0000	1.0000	0.9583	0.0408	0.7392	0.99
40							
ctDNA.Dynamics=2							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12	3	0	1	0	1	
1							
ctDNA.Dynamics=3							
	time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95%
CI	12.00000	1.00000	1.00000	0.50000	0.35355	0.00598	0.910
41							

Hide

```

circ_data$ctDNA.Dynamics <- factor(circ_data$ctDNA.Dynamics, levels=c("1", "2", "3"), labels = c("Persistently Negative", "Converted Negative", "Persistently Positive"))
cox_fit <- coxph(surv_object ~ ctDNA.Dynamics, data=circ_data)
ggforest(cox_fit, data = circ_data)

```


[Hide](#)

```
summary(cox_fit)
```

Call:

```
coxph(formula = surv_object ~ ctDNA.Dynamics, data = circ_data)
```

n= 29, number of events= 11

(1 observation deleted due to missingness)

	coef	exp(coef)	se(coef)	z	Pr(> z)
ctDNA.DynamicsConverted Negative	2.1387	8.4884	0.7424	2.881	0.00397 **
ctDNA.DynamicsPersistently Positive	2.7683	15.9319	0.9263	2.988	0.00280 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

	exp(coef)	exp(-coef)	lower .95	upper .95
ctDNA.DynamicsConverted Negative	8.488	0.11781	1.981	36.37
ctDNA.DynamicsPersistently Positive	15.932	0.06277	2.593	97.90

Concordance= 0.706 (se = 0.07)

Likelihood ratio test= 11.12 on 2 df, p=0.004

Wald test = 12.88 on 2 df, p=0.002

Score (logrank) test = 19.67 on 2 df, p=5e-05

Hide

```
#Fisher plot for ctDNA Dynamics with Rec Status
```

```
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels =  
c("No Recurrence", "Recurrence"))
```

```
contingency_table <- table(circ_data$ctDNA.Dynamics, circ_data$DFS.Event)
```

```
chi_square_test <- chisq.test(contingency_table)
```

Warning in stats::chisq.test(x, y, ...) :

Chi-squared approximation may be incorrect

Hide

```
print(chi_square_test)
```

Pearson's Chi-squared test

data: contingency_table

X-squared = 9.8864, df = 2, p-value = 0.007132

Hide

```
fisher_exact_test <- fisher.test(contingency_table)
```

```
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table
p-value = 0.00389
alternative hypothesis: two.sided
```

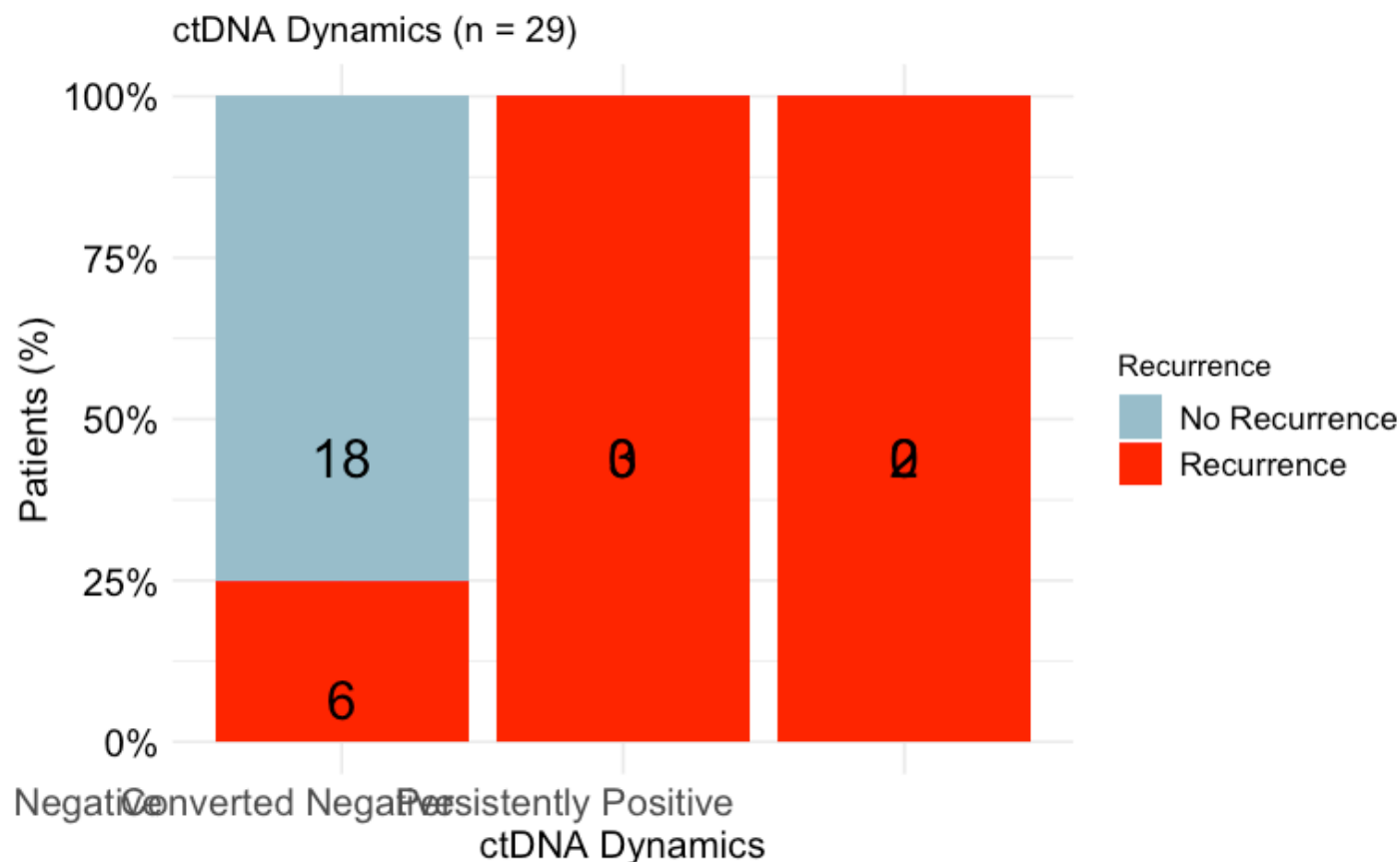
Hide

```
print(contingency_table)
```

	No Recurrence	Recurrence
Persistently Negative	18	6
Converted Negative	0	3
Persistently Positive	0	2

Hide

```
table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA Dynamics (n = 29)", x = "ctDNA Dynamics", y = "Patients (%)", fill = "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence label size
```



#DFS by ctDNA post-NAT and rNAR - 3 groups

Hide

```
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$rNAR.Score!="",]

circ_data$ctDNA.postNAT.NAR <- NA #first we create the variable for the ctDNA & NAC combination, and we assign values
circ_data <- circ_data %>%
  mutate(ctDNA.postNAT.NAR = case_when(
    rNAR.Score == "Low" & ctDNA.postNAC == "NEGATIVE" ~ 1,
    rNAR.Score == "Mid/High" & ctDNA.postNAC == "NEGATIVE" ~ 2,
    rNAR.Score == "Low" & ctDNA.postNAC == "POSITIVE" ~ 3,
    rNAR.Score == "Mid/High" & ctDNA.postNAC == "POSITIVE" ~ 4
  ))

survfit(Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)~ctDNA.postNAT.NAR, data = circ_data)
```

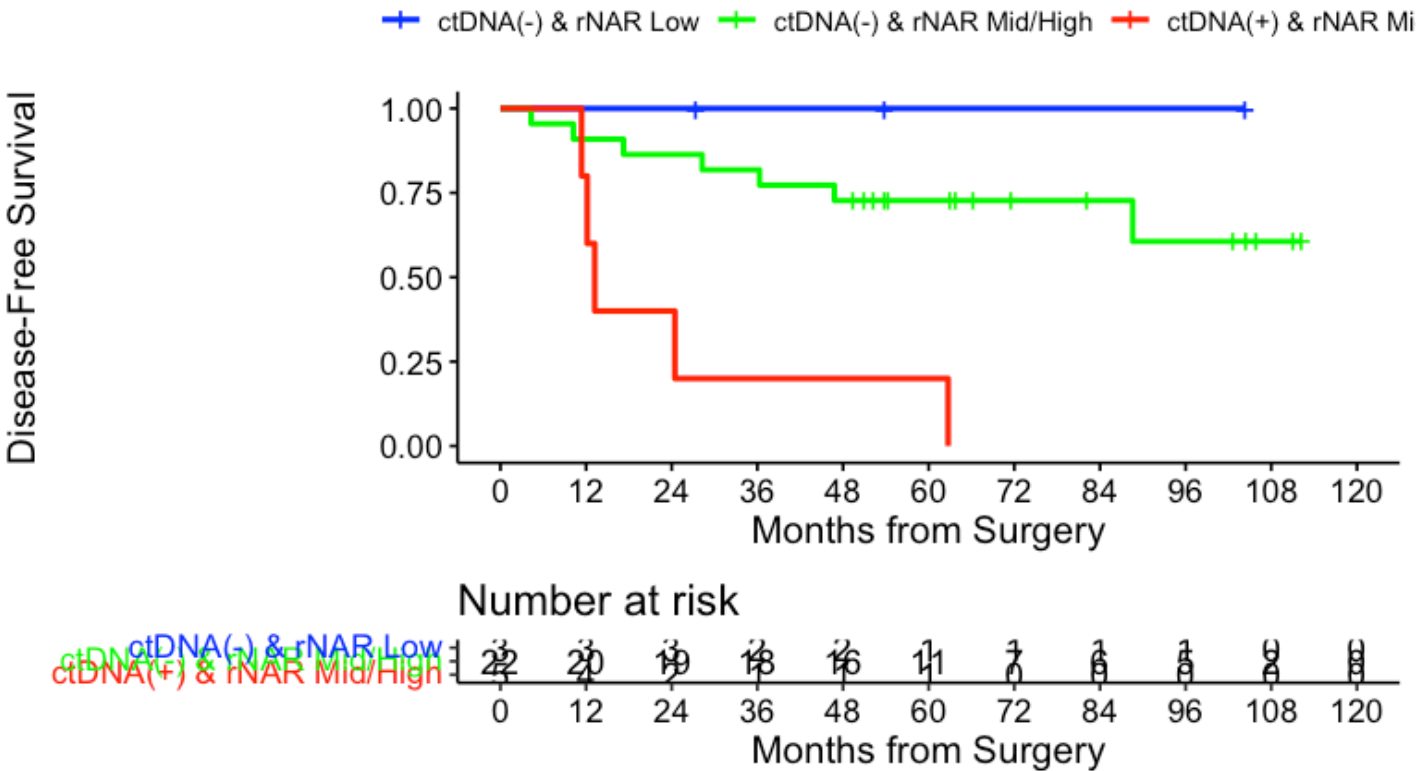
```
Call: survfit(formula = Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event) ~
      ctDNA.postNAT.NAR, data = circ_data)
```

	n	events	median	0.95LCL	0.95UCL
ctDNA.postNAT.NAR=1	3	0	NA	NA	NA
ctDNA.postNAT.NAR=2	22	7	NA	88.6	NA
ctDNA.postNAT.NAR=4	5	5	13.2	12.2	NA

[Hide](#)

```
surv_object <- Surv(time = circ_data$DFS.months, event = circ_data$DFS.Event)
KM_curve <- survfit(surv_object ~ ctDNA.postNAT.NAR, data = circ_data, conf.int=0.95, conf.type="log-log")
ggsurvplot(KM_curve, data = circ_data, pval = FALSE, conf.int = FALSE, risk.table = TRUE, break.time.by=12, palette=c("blue","green","red"), title="DFS - ctDNA post-NAT | rNAR", ylab= "Disease-Free Survival", xlab="Months from Surgery", legend.labs=c("ctDNA(-) & rNAR Low", "ctDNA(-) & rNAR Mid/High", "ctDNA(+) & rNAR Mid/High"), legend.title="")
```

DFS - ctDNA post-NAT | rNAR



Hide

```
summary(KM_curve, times= c(12, 24, 36))
```

```
Call: survfit(formula = surv_object ~ ctDNA.postNAT.NAR, data = circ_data,
  conf.int = 0.95, conf.type = "log-log")
```

ctDNA.postNAT.NAR=1

time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI
12	3	0	1	0	1	1
24	3	0	1	0	1	1
36	2	0	1	0	NA	NA

ctDNA.postNAT.NAR=2

time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI
12	20	2	0.909	0.0613	0.683	0.976
24	19	1	0.864	0.0732	0.634	0.954
36	18	1	0.818	0.0822	0.585	0.928

ctDNA.postNAT.NAR=4

time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI
12	4	1	0.8	0.179	0.20381	0.969
24	2	2	0.4	0.219	0.05198	0.753
36	1	1	0.2	0.179	0.00837	0.582

Hide


```
circ_data$ctDNA.postNAT.NAR <- factor(circ_data$ctDNA.postNAT.NAR, levels=c
("1","2","4"), labels = c("ctDNA(-) & rNAR Low", "ctDNA(-) & rNAR Mid/High", "ctDNA(+) &
rNAR Mid/High"))
cox_fit <- coxphf(surv_object ~ ctDNA.postNAT.NAR, data=circ_data)
summary(cox_fit)
```

```
coxphf(formula = surv_object ~ ctDNA.postNAT.NAR, data = circ_data)
```

Model fitted by Penalized ML

Confidence intervals and p-values by Profile Likelihood

	coef	se(coef)	exp(coef)	lower 0.95	upper
0.95	Chisq	p			
ctDNA.postNAT.NARctDNA(-) & rNAR Mid/High	0.7029159	1.549204	2.019633	0.2452021	262.3526
ctDNA.postNAT.NARctDNA(+) & rNAR Mid/High	2.6477515	1.583745	14.122249	1.5198676	1881.0076
	5.8480431	0.01559438			

Likelihood ratio test=9.805837 on 2 df, p=0.007424883, n=30

Wald test = 10.15246 on 2 df, p = 0.006243397

Covariance-Matrix:

	ctDNA.postNAT.NARctDNA(-) & rNAR Mid/High	ctDNA.postNAT.NARctDNA(+) & rNAR Mid/High
A.postNAT.NARctDNA(+) & rNAR Mid/High		
ctDNA.postNAT.NARctDNA(-) & rNAR Mid/High	2.400034	
	2.250714	
ctDNA.postNAT.NARctDNA(+) & rNAR Mid/High		2.250714
	2.508248	

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```
#Fisher plot for ctDNA post-NAT and rNAR.Score combination with Rec Status
rm(list=ls())
setwd("~/Downloads")
circ_data <- read.csv("Rectal IRST_Clinical Data.csv")
circ_data <- circ_data[circ_data$rNAR.Score!="",]

circ_data$ctDNA.postNAT.NAR <- NA #first we create the variable for the ctDNA & NAC comb
ination, and we assign values
circ_data <- circ_data %>%
  mutate(ctDNA.postNAT.NAR = case_when(
    rNAR.Score == "Low" & ctDNA.postNAC == "NEGATIVE" ~ 1,
    rNAR.Score == "Mid/High" & ctDNA.postNAC == "NEGATIVE" ~ 2,
    rNAR.Score == "Low" & ctDNA.postNAC == "POSITIVE" ~ 3,
    rNAR.Score == "Mid/High" & ctDNA.postNAC == "POSITIVE" ~ 4
  ))
circ_data$DFS.Event <- factor(circ_data$DFS.Event, levels = c("FALSE", "TRUE"), labels =
c("No Recurrence", "Recurrence"))
contingency_table <- table(circ_data$ctDNA.postNAT.NAR, circ_data$DFS.Event)
chi_square_test <- chisq.test(contingency_table)
```

```
Warning in stats::chisq.test(x, y, ...) :  
  Chi-squared approximation may be incorrect
```

[Hide](#)

```
print(chi_square_test)
```

Pearson's Chi-squared test

```
data: contingency_table  
X-squared = 10.114, df = 2, p-value = 0.006366
```

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```
fisher_exact_test <- fisher.test(contingency_table)  
print(fisher_exact_test)
```

Fisher's Exact Test for Count Data

```
data: contingency_table  
p-value = 0.004492  
alternative hypothesis: two.sided
```

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```
print(contingency_table)
```

	No Recurrence	Recurrence
1	3	0
2	15	7
4	0	5

[Hide](#)

```

table_df <- as.data.frame(contingency_table)
table_df$Total <- ave(table_df$Freq, table_df$Var1, FUN = sum)
table_df$Percentage <- table_df$Freq / table_df$Total
table_df$MiddlePercentage <- table_df$Percentage / 2
ggplot(table_df, aes(x = Var1, y = Percentage, fill = Var2)) +
  geom_bar(stat = "identity") +
  geom_text(aes(y = MiddlePercentage, label = Freq), position = "stack", color = "black",
    vjust = 1.5, size = 7) +
  theme_minimal() +
  labs(title = "ctDNA post-NAT | rNAR", x = "ctDNA | rNAR", y = "Patients (%)", fill =
    "Recurrence") +
  scale_y_continuous(labels = scales::percent_format()) +
  scale_fill_manual(values = c("No Recurrence" = "lightblue3", "Recurrence" = "red")) +
  # define custom colors
  theme(axis.text.x = element_text(angle = 0, hjust = 1.5, size = 14), # increase x-axis
    text size
    axis.text.y = element_text(size = 14, color = "black"), # increase y-axis text s
    size
    axis.title.x = element_text(size = 14, color = "black"), # increase x-axis label
    size
    axis.title.y = element_text(size = 14, color = "black"), # increase y-axis label
    size
    legend.text = element_text(size = 12, color = "black")) # increase Recurrence l
    abel size

```

