

Biomarker analysis report

2017-08-29

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1 The dataset

The dataset have 368 entries. In which 223 are in biomarker evaluable population (BEP).

- Endpoint of interest: Lab_ontrt
- Biomarker: KRAS.mutant
- Biomarker type: categorical

2 Representativeness: Selection Bias of Biomarker Population

In this section, we are trying to answer the question: *Are biomarker evaluable population representative of the full population population?*

Key baseline demographics and prognostic characteristics (including stratification variables and any variables with known prognostic effect) and efficacy outcomes should be summarized by treatment groups and compared between biomarker evaluable population (BEP) and the full population. These analyses are conducted to investigate any potential selection bias associated with the availability of the biomarker (e.g. we may not get enough tissue for patients whose tumor size is small. Therefore they may be excluded from BEP).

The key clinical biomarkers considered are:

```
clinical.vars.class
```

```
##           Sex           Age
## "categorical"    "numeric"
```

2.1 Check selection bias in terms of key clinical variables, between full population and BEP

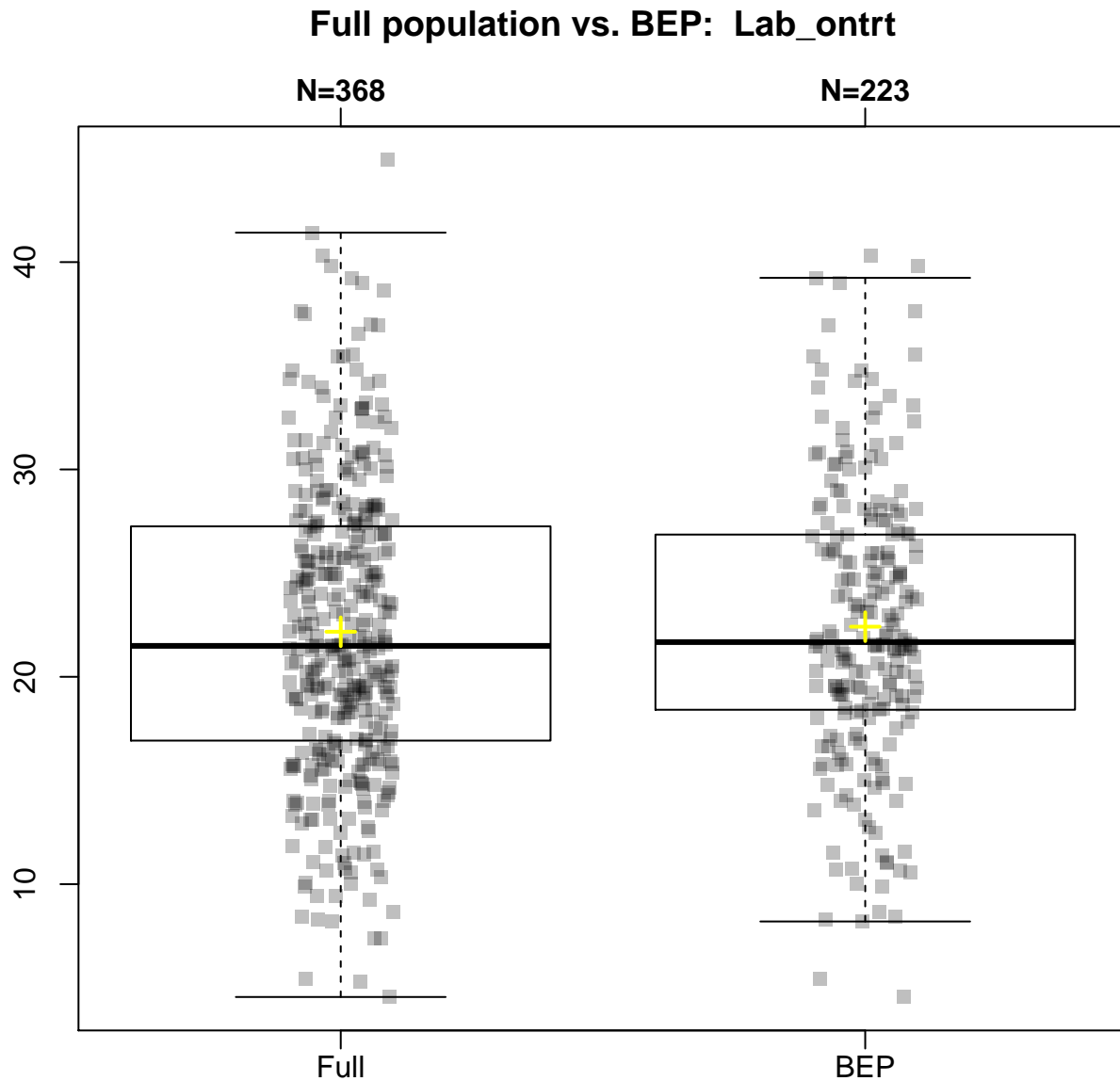
```
kable(SummaryVars(data=input, trt=trt, subgroup=BEP, subgroup.indicator = BEP.indicator,
var=clinical.vars, var.class=clinical.vars.class))
```

	All	BEP
Sex		
Total	368	223
NA's	0	0
F	184 (50%)	111 (49.78%)
M	184 (50%)	112 (50.22%)
Age		
N	368	223
Mean	54.03	54.27
Median	54	54
Min-Max	30...89	33...89
NA's	0	0

2.2 Check whether the clinical outcome in BEP is comparable to the full population

The following plot compares continuous outcome in BEP vs. the full population. The response category distribution is plotted for each arm. The BEP bars are expected to be comparable to those in full population.

```
BoxPlot(list(Full=input[[outcome.var]], BEP=input.bep[[outcome.var]]),
Title=list(main=paste("Full population vs. BEP: ",outcome.var)), mar=c(5,5,5,1))
```



2.3 Examine whether the prognostic/null trend of key clinical variables holds in BEP

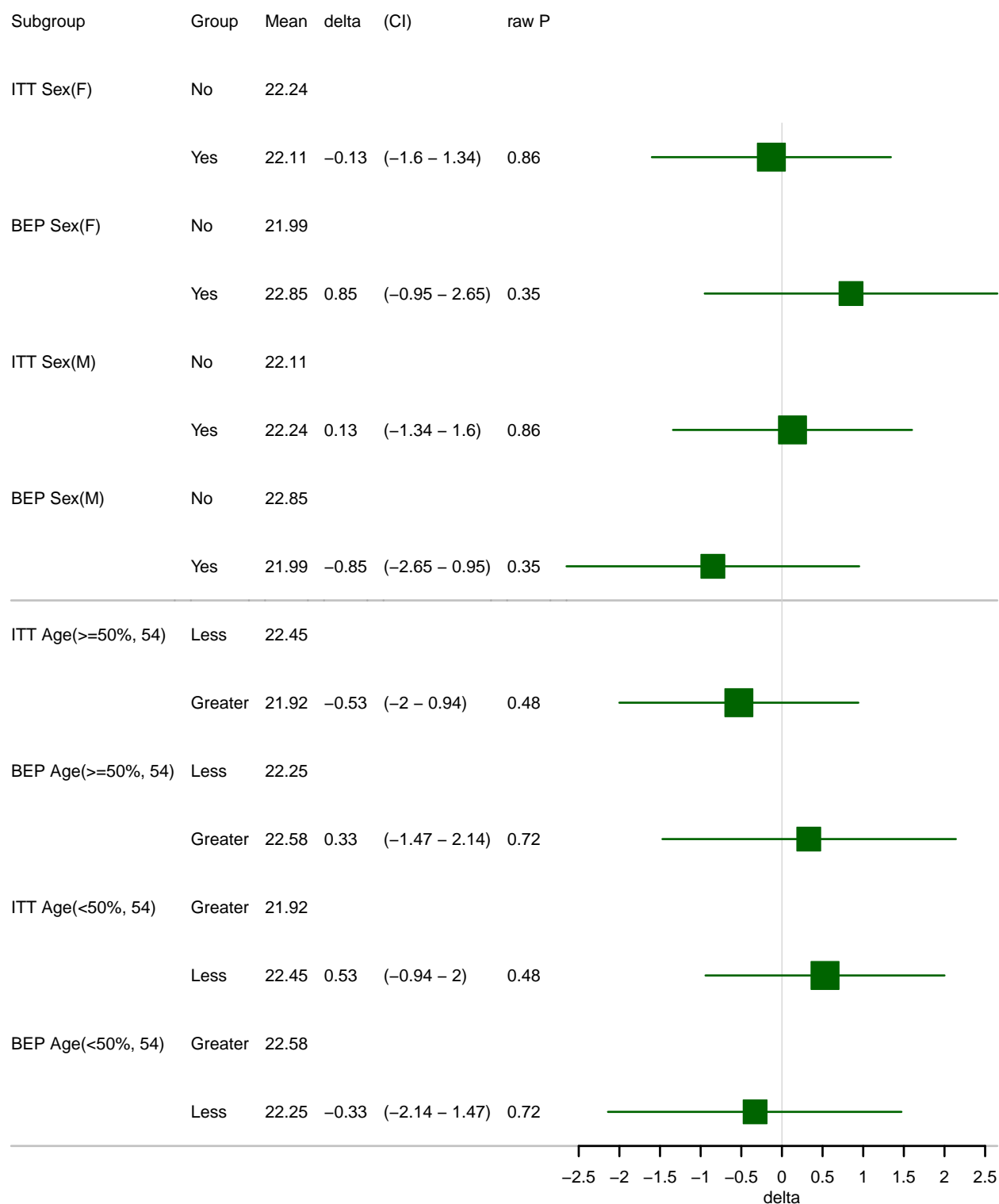
The following forest plot can be used to examine whether any of the key prognostic clinical variables still show prognostic trend in BEP:

```
forest.bep <- PlotTabForestMulti(data=input,
                                outcome.class=outcome.class,
                                outcome.var=outcome.var,
                                trt=trt,
                                var=clinical.vars,
                                var.class=clinical.vars.class,
                                bep=BEP,bep.indicator=BEP.indicator,
                                compare.bep.itt=TRUE
                                )
```

Stratification is not supported for continuous outcome

```
## Stratification is not supported for continuous outcome
## Stratification is not supported for continuous outcome
## Stratification is not supported for continuous outcome
```

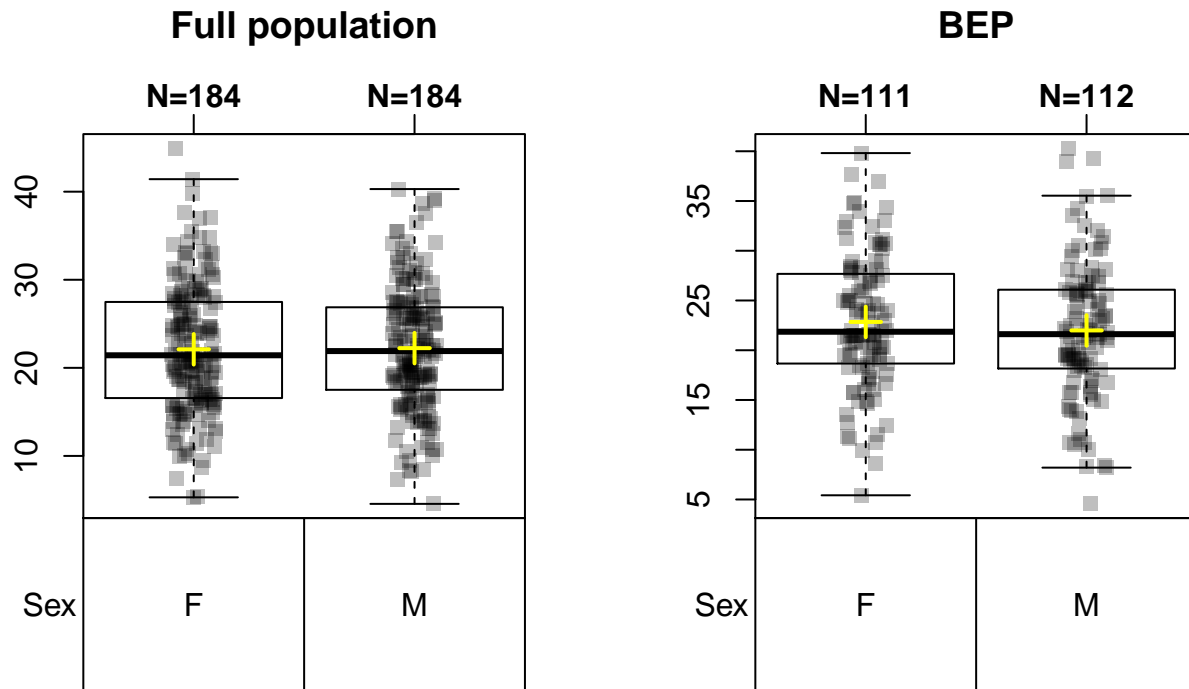
**Within arm, Compare BEP vs. All
Lab_ontrt
Unadjusted, unstratified analysis**



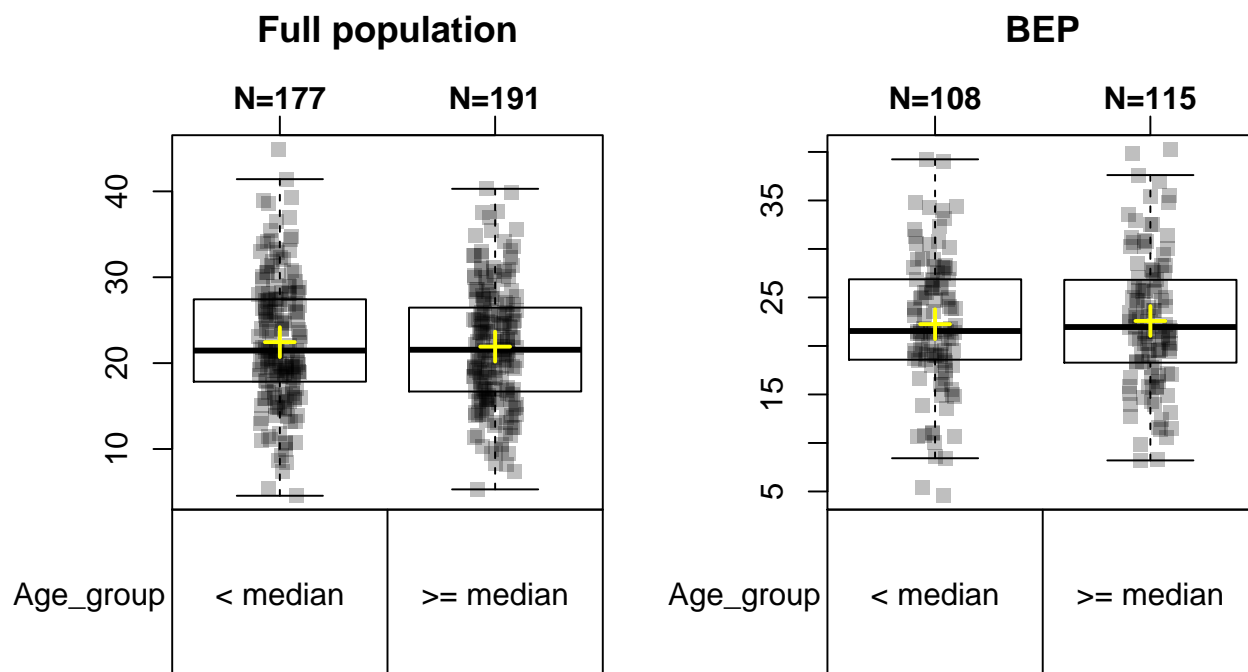
```

for(i in clinical.vars.2){
  par(mfrow=c(1,2))
  BoxPlot(obj=input, form=as.formula(paste(outcome.var,"~",i)),
          XaxisTab=list(font=2, col="darkblue", cex=1.25),
          Title=list(main="Full population"), mar=c(5,5,5,1))
  BoxPlot(obj=input.bep, form=as.formula(paste(outcome.var,"~",i)),
          XaxisTab=list(font=2, col="darkblue", cex=1.25),
          Title=list(main="BEP"), mar=c(5,5,5,1))
  print("")
}

```



```
## [1] ""
```



```
## [1] ""
```

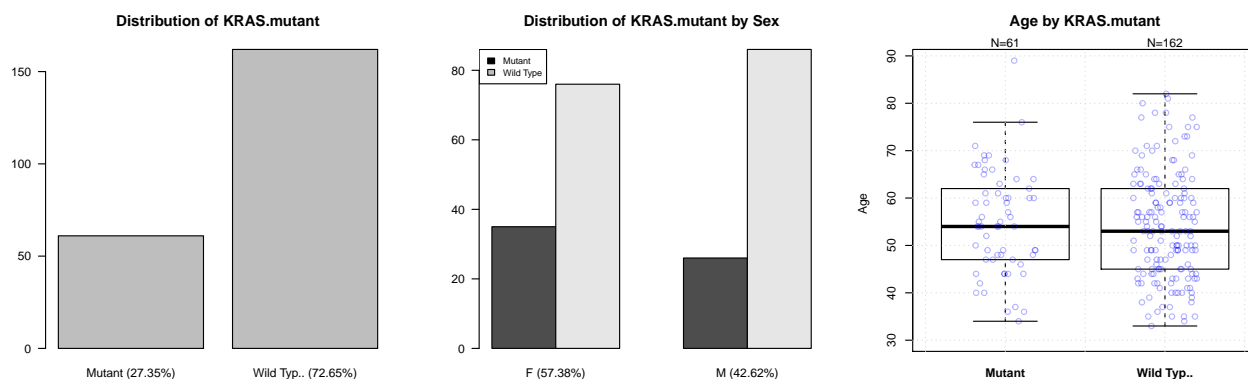
3 Biomarker property and its association to clinical variables

Relationship between the biomarker and key demographic and prognostic variables should also be investigated using bivariate plots. Prognostic property of the biomarker should also be assessed, by estimates of the clinical efficacy in the control arm.

3.1 Biomarker property and relationship to clinical variable

The following results show single variate plot for biomarker and bi-variate plots to investigate biomarker-clinical variable relationship.

```
PlotProperty(data=input, biomarker.var=bm, biomarker.class=bm.class,
             var=clinical.vars,
             var.class=clinical.vars.class,
             log2=FALSE, par.param = list(mfrow=c(2,3)))
```



4 Biomarker subgroup analysis

4.1 Estimations within each subgroup

The following figures investigate whether the biomarker shows a within-arm effect (e.g. patients in one biomarker subgroup tend to have better clinical outcome):

```
if(bm.class=="numeric"){
  if(!is.null(numerical.finalcut)) levs <- paste0(c(">=", "<"), numerical.finalcut)
  if(is.null(numerical.finalcut)) {
    nm <- quantile(input.bep[[bm]], percentile.finalcut, 2) # default quantile type in forest plot function
    numerical.finalcut <- round(nm, 2) # default rounding decimal in forest plots
    levs <- paste0(c(">=", "<"), percentile.finalcut*100, "%")
  }

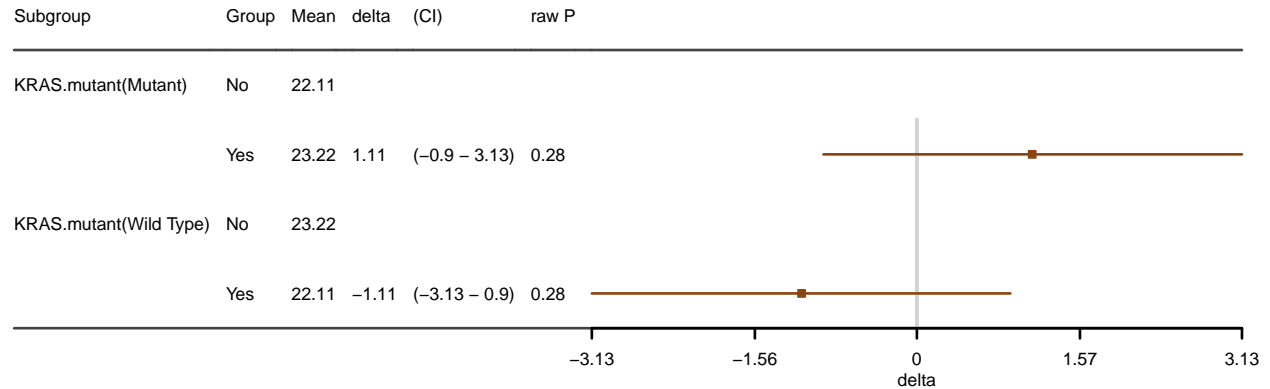
  bm2 <- paste0(bm, "_Dx")
  input[[bm2]] <- ifelse(input[[bm]] >= numerical.finalcut, levs[1], levs[2])
  input[[bm2]] <- factor(input[[bm2]], levels=levs) # ">=" as Dx+
}

if(bm.class=="categorical") {
  bm2 <- bm
  levs <- unique(input[[bm2]])
}

res.2group <- PlotTabForestBiomarker(data=input,
                                     outcome.class=outcome.class,
                                     outcome.var=outcome.var,
                                     trt=trt,
                                     var=bm2,
                                     var.class="categorical",
                                     greater=TRUE, less=TRUE,
                                     show.itt=TRUE, show.bep=TRUE,
                                     covariate=covariate, strata=strata)

## Stratification is not supported for continuous outcome
## Some NAs in var column, will define the non NA entries as BEP
## only 1 arm; show.itt is set to FALSE
## only 1 arm; show.bep is set to FALSE
```


Within-arm Effect of Biomarker
Lab_ontrt, KRAS.mutant
Unadjusted, unstratified analysis



The forest plots above show within-arm mean difference (delta) across biomarker subgroups. For a given arm, if the delta is not all around 0, it indicates that within this arm the biomarker has an association to the clinical outcome.

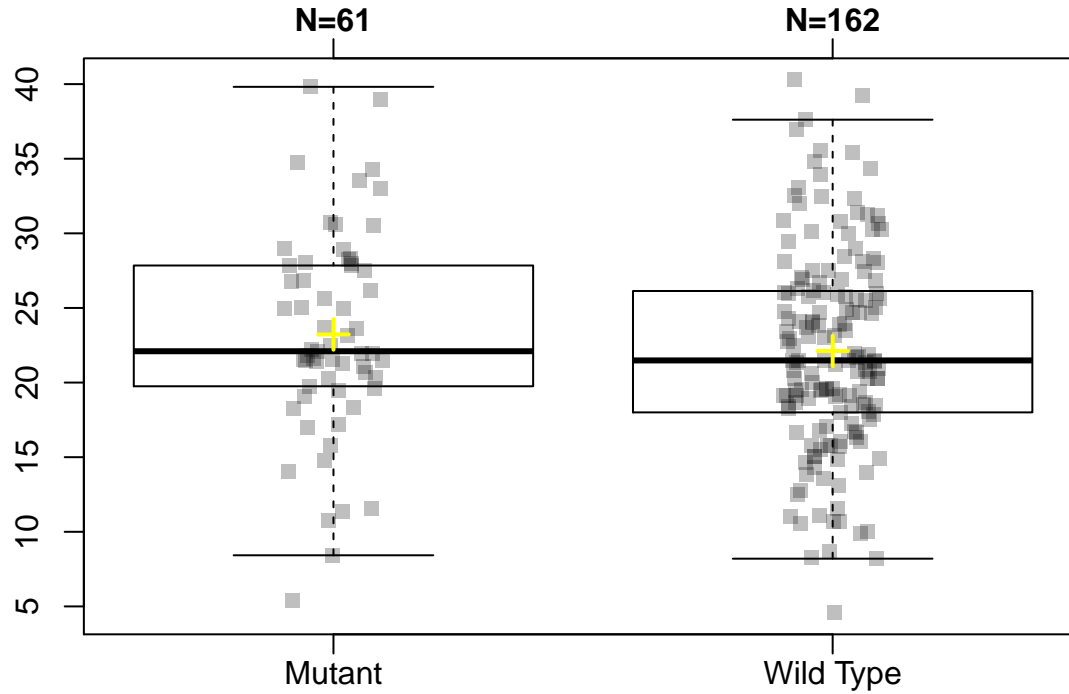
4.2 Subgroup analysis

The following figure show distribution of the continuous endpoint within the biomarker subgroups:

```
input.bep <- input[which(input[[BEP]]==BEP.indicator),]

BoxPlot(obj=input.bep, form=as.formula(paste(outcome.var,"~",bm)),
  # XaxisTab=list(font=2, col="darkblue", cex=1.25),
  Title=list(main="BEP: biomarker subgroups"),
  mar=c(5,8,5,1))
```

BEP: biomarker subgroups



4.3 Check whether biomarker subgroup is confounded with key clinical variables

The following table checks whether the biomarker subgroup is confounded with clinical variables. Here we check whether the clinical variable distribution is comparable in biomarker subgroups.

```
kable(
  SummaryVars(data=input.bep,trt=trt, subgroup=bm2, var=clinical.vars,
    var.class=clinical.vars.class, subgroup.indicator=levs[1],compare.subgroup=TRUE)
)
```

	KRAS.mutant__Mutant	KRAS.mutant__Wild Type
Sex		
Total	61	162
NA's	0	0
F	35 (57.38%)	76 (46.91%)
M	26 (42.62%)	86 (53.09%)
Age		
N	61	162
Mean	54.92	54.03
Median	54	53
Min-Max	34...89	33...82
NA's	0	0