Supporting Information for Hilvo et al. (2015)

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1 Libraries and functions

The R scripts provided in this vignette fully reproduces the TCGA survival analyses of Hilvo et al (2015). The following R packages may be required:

• survival

We use a local library for some low level functions. This will be loaded from http://www.markowetzlab.org/supplements/if not present on the local machine.

2 External Data Sources

The package Hilvo2015-supplement-data.Rdata contains the publicly available data described in the following subsections. This will be loaded from http://www.markowetzlab.org/supplements/ if not present on the local machine.

```
## load data file from local copy or from URL
if (file.exists("Hilvo2015-supplementary-data.Rdata")){
    load("Hilvo2015-supplementary-data.Rdata")
    cat("Hilvo2015-supplementary-data.Rdata loaded from local copy")
} else {
    #load(url("https://github.com/InesdeSantiago/survivalTCGAOC/blob/master/Hilvo2015-supplementary-data.Rdata loaded from URL")
    cat("Hilvo2015-supplementary-data.Rdata loaded from URL")
    cat("please download file from github repository: https://github.com/InesdeSantiago/specific from the supplementary-data.Rdata loaded from local copy
## Hilvo2015-supplementary-data.Rdata loaded from local copy
```

2.1 TCGA data

Data from 489 high-grade serous ovarian cancers reported by The Cancer Genome Atlas Research Network (Nature, 2011, 609-615) were obtained from https://tcga-data.nci.nih.gov/docs/publications/ov_2011/. The exp.TCGAOC object contains the gene expression matrix obtained from https://tcga-data.nci.nih.gov/docs/publications/ov_2011/:

```
dim(exp.TCGAOC)
## [1] 11864 489
```

The clinical.TCGAOC object contains the clinical annotations for all samples obtained from: $\label{eq:clinical} https://tcga-data.nci.nih.gov/docs/dictionary/TCGA_BCR_DataDictionary.xml$

```
dim(clinical.TCGAOC)
## [1] 488 12
colnames(clinical.TCGAOC)
   [1] "AgeAtDiagnosis..yrs."
                                         "VITALSTATUS"
    [3] "TUMORSTAGE"
##
                                         "TUMORGRADE"
##
   [5] "TUMORRESIDUALDISEASE"
                                         "PRIMARYTHERAPYOUTCOMESUCCESS"
   [7] "PERSONNEOPLASMCANCERSTATUS"
                                         "OverallSurvival.mos."
## [9] "ProgressionFreeStatus"
                                         "ProgressionFreeSurvival..mos.."
## [11] "PlatinumFreeInterval..mos.."
                                         "PlatinumStatus"
```

2.2 Copy Number Alterations (CNA)

The CNA data was downloaded from cBioPortal (http://cbioportal.org) using the TCGA set of 489 high-grade serous ovarian cancers. Only 316 tumours have CNA data available:

```
dim(CNAtable)
## [1] 316 69
```

3 Survival analysis for the TCGA cohort

To validate the metabolomics findings with gene expression data, survival analyses were performed for The Cancer Genome Atlas (TCGA) data. Survival analysis based on gene expression and copy number was performed for genes of those selected KEGG pathways that showed most interesting metabolomics results: fatty acid import and beta oxidation, omega oxidation, ketone body production, pentose phosphate pathway reactions related to 3-erythritol accumulation, leucine degradation reactions related to 3-hydroxyisovaleric acid accumulation, peroxisomal biogenesis factors related to Zellweger syndrome as well as ALDH5A1 gene encoding for SSADH enzyme.

```
colnames(CNAtable)
##
    [1] "ACAA1"
                   "ACAA2"
                              "ACADL"
                                         "ACADM"
                                                   "ACADS"
                                                              "ACADSB"
                                                                         "ACADVL"
##
    [8] "ACAT1"
                   "ACOX1"
                              "ACSBG1"
                                         "ACSBG2"
                                                   "ACSL1"
                                                              "ACSL3"
                                                                         "ACSL5"
                                                   "ADH5"
   [15] "ACSL6"
                              "ADH1B"
                                         "ADH1C"
                                                              "ADH6"
                                                                         "ADH7"
##
                   "ADH1A"
## [22] "ALDH1B1" "ALDH2"
                              "ALDH3A2" "ALDH5A1" "ALDH7A1" "ALDH9A1" "AUH"
## [29] "BDH1"
                   "BDH2"
                              "BTD"
                                         "CPT1A"
                                                   "CPT1B"
                                                              "CPT2"
                                                                         "CYP4A11"
## [36] "CYP4B1"
                   "CYP4F11"
                              "DLD"
                                         "ECHS1"
                                                   "ECI1"
                                                              "ECI2"
                                                                         "EHHADH"
## [43] "FABP4"
                   "HADH"
                              "HADHA"
                                         "HADHB"
                                                   "HLCS"
                                                              "HMGCL"
                                                                         "HMGCS1"
## [50] "HMGCS2"
                   "MCCC1"
                              "MCCC2"
                                         "OXCT1"
                                                   "OXCT2"
                                                              "PEX1"
                                                                         "PEX10"
## [57] "PEX12"
                   "PEX13"
                              "PEX14"
                                         "PEX16"
                                                   "PEX19"
                                                              "PEX3"
                                                                         "PEX5"
## [64] "PEX6"
                   "PEX7"
                              "RPIA"
                                         "SHPK"
                                                   "TALDO1"
                                                              "TKT"
```

For each gene of interest, we subdivided the TCGA Ovarian Cancer cohort according to:

- Four quartiles of gene expression
- Different copy number alterations: Loss, wt and gain

First, we simplified the Gistic scores obtained from the downloaded table as "loss", "wt", "gain":

```
CNAallgenes <- CNAtable

CNAallgenes [CNAallgenes==-2] <- "loss" #hom del

CNAallgenes [CNAallgenes==-1] <- "loss" #het loss

CNAallgenes [CNAallgenes==0] <- "wt"

CNAallgenes [CNAallgenes==1] <- "gain" #gain

CNAallgenes [CNAallgenes==2] <- "gain" #amp

CNAallgenes <- t(CNAallgenes)
```

```
# Survival wrapper function
survwrapp <- function(gene, plot = TRUE) {</pre>
```

```
\# par(mfrow=c(1,2), mar=c(20,4,5,4))
    # by quartiles
    group <- cut(as.numeric(exp.TCGAOC[gene, ]), quantile(as.numeric(exp.TCGAOC[gene,</pre>
        ]), probs = c(0, 0.25, 0.5, 0.75, 1)))
    names(group) <- names(exp.TCGAOC[gene, ])</pre>
    survmat <- cbind(groups = group[rownames(clinical.TCGAOC)], clinical.TCGAOC)</pre>
    pvalue1 <- plotsurv(survmat, colors = c("blue", "lightblue", "gray", "red"),</pre>
        maint = "", plot = plot)
    if (plot) {
        title(main = paste(gene, "\nBy expression quartile"))
    # by mutation type
    group <- factor(CNAallgenes[gene, ], levels = c("loss", "wt", "gain"))</pre>
    survmat <- cbind(groups = group[rownames(clinical.TCGAOC)], clinical.TCGAOC)</pre>
    pvalue2 <- plotsurv(survmat, colors = c("turquoise4", "darkgray", "orangered"),</pre>
        maint = "", plot = plot)
    if (plot) {
        title(main = paste(gene, "\nBy CNA type"))
    # this function returns the chisq p-values obtained in both cases
    # 'by_ExpQuartiles' and 'by_CNA'
    return(c(by_ExpQuartiles = pvalue1, by_CNA = pvalue2))
plotsurv <- function(survmat, colors, maint, timecol = "OverallSurvival.mos.",</pre>
    eventcol = "VITALSTATUS", plot = TRUE) {
    library(survival)
    # fix survmat
    if (eventcol == "VITALSTATUS") {
        survmat$VITALSTATUS <- as.character(survmat$VITALSTATUS)</pre>
        survmat$VITALSTATUS[survmat$VITALSTATUS == "LIVING"] <- 0</pre>
        survmat$VITALSTATUS[survmat$VITALSTATUS == "DECEASED"] <- 1</pre>
        survmat$VITALSTATUS <- as.numeric(survmat$VITALSTATUS)</pre>
    if (eventcol == "ProgressionFreeStatus") {
        survmat$ProgressionFreeStatus <- as.character(survmat$ProgressionFreeStatus)</pre>
        survmat$ProgressionFreeStatus[survmat$ProgressionFreeStatus == "DiseaseFree"] <-</pre>
        survmat$ProgressionFreeStatus[survmat$ProgressionFreeStatus == "Recurred/ProgressionFreeStatus"]
```

```
survmat$ProgressionFreeStatus <- as.numeric(survmat$ProgressionFreeStatus)</pre>
survmat <- subset(survmat, select = c("groups", timecol, eventcol))</pre>
survmat <- na.omit(survmat)</pre>
# plot
ss <- Surv(as.numeric(survmat[, timecol]), survmat[, eventcol])</pre>
if (plot) {
    plot(survfit(ss ~ survmat$groups), col = colors, xlab = paste(timecol,
         "[months]"), ylab = "Survival Probability", lwd = 3, las = 1, main = maint)
    legendtext = paste(levels(survmat$group), ": n=", table(survmat$group),
        sep = "")
    legend("topright", col = colors, legend = legendtext, lty = 1)
}
# Pvalue and HR
res <- survdiff(ss ~ survmat$group)</pre>
p <- 1 - pchisq(res$chisq, 1)</pre>
if (p < 0.001) {
    p <- format(p, scientific = TRUE, nsmall = 3)</pre>
} else {
    p \leftarrow round(p, 3)
if (plot) {
    text(100, 0.6, paste("p-value=", p))
    options(scipen = 10)
return(p)
```

3.1 Supplementary Table S7: survival analyses using TCGA data

Survival analysis was performed for the following genes:

```
genes <- c('ALDH5A1','DLD','PEX6','HMGCL','ACSL3','ADH1B','HLCS','AUH','ALDH7A1',
'MCCC2','PEX12','OXCT1','ALDH1B1','EHHADH','HMGCS1','FABP4','ECI1','HADHA','CPT1B')</pre>
```

For each one of the selected genes, we investigated if low expression and loss in copy numbers of the gene were associated with worse overall survival. The following code reproduces the p-values in supplementary table 7.

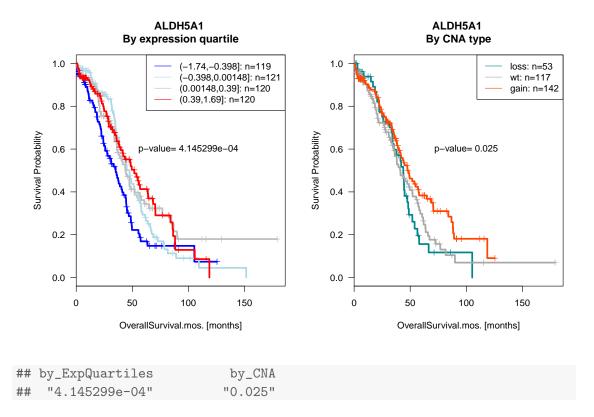
```
library(knitr)
pvalues <- t(sapply(genes, survwrapp, plot=FALSE))
kable(pvalues)</pre>
```

	by_ExpQuartiles	by_CNA
ALDH5A1	4.145299e-04	0.025
DLD	4.627128e-04	0.12
PEX6	8.578187e-04	0.235
HMGCL	0.004	0.56
ACSL3	0.009	0.137
ADH1B	0.027	0.641
HLCS	0.029	0.388
AUH	0.033	0.093
ALDH7A1	0.029	0.022
MCCC2	0.887	1.702884e-04
PEX12	0.427	0.003
OXCT1	0.178	0.004
OXCT1 ALDH1B1	0.178	0.004
ALDH1B1	0.217	0.005
ALDH1B1 EHHADH	0.217 0.063	0.005 0.01
ALDH1B1 EHHADH HMGCS1	0.217 0.063 0.082	0.005 0.01 0.012
ALDH1B1 EHHADH HMGCS1 FABP4	0.217 0.063 0.082 0.059	0.005 0.01 0.012 0.018
ALDH1B1 EHHADH HMGCS1 FABP4 ECI1	0.217 0.063 0.082 0.059 0.355	0.005 0.01 0.012 0.018 0.025

3.2 Figure 2E and Figure S5: Low gene expression and loss of ALDH5A1 copy number were both associated to poor overall survival

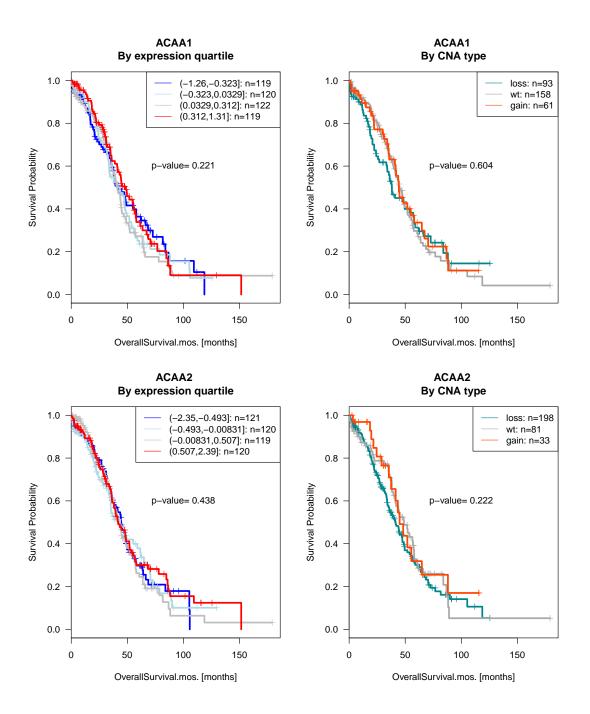
Intriguingly, the most significant findings were obtained for the ALDH5A1 gene, as in the TCGA data set both low expression and loss in copy numbers of the gene were associated with worse overall survival of the patients.

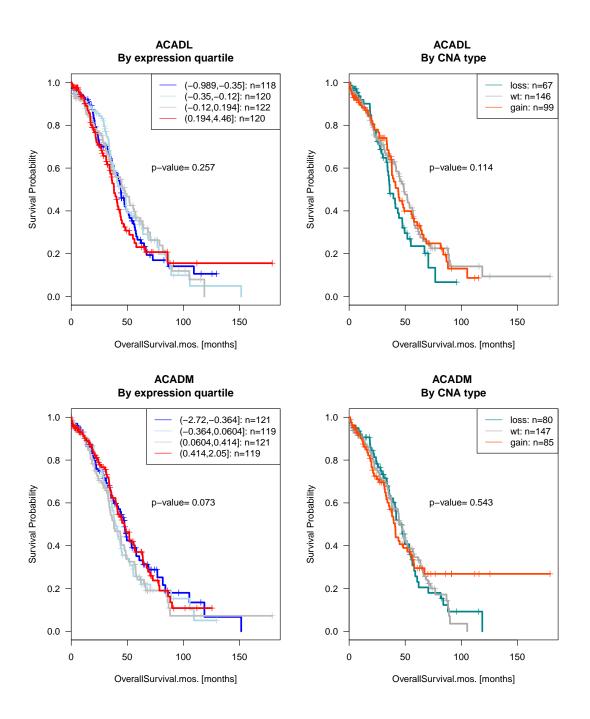
```
par(mfrow=c(1,2))
survwrapp("ALDH5A1")
```

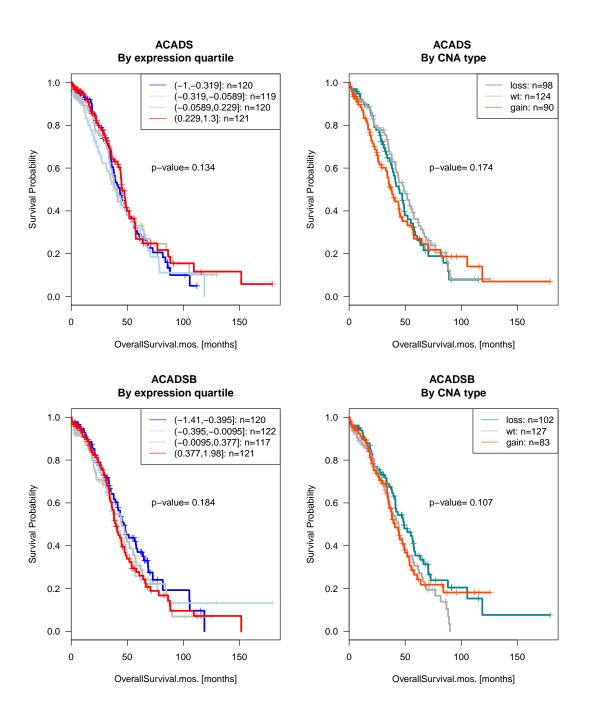


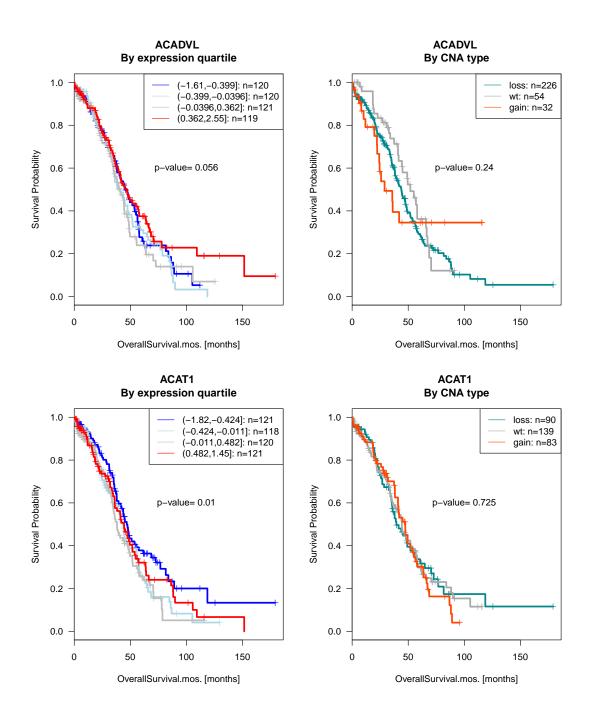
The survival plots for all analysed genes can be inspected with the following code:

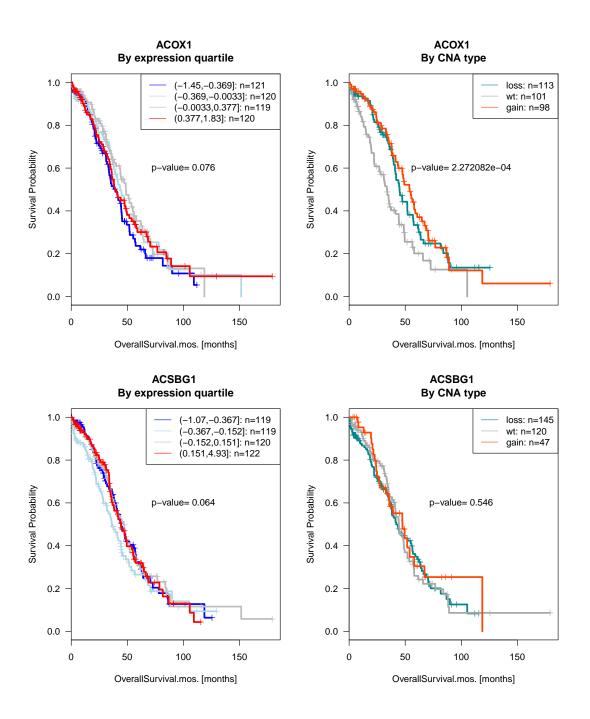
```
par(mfrow=c(1,2))
pvalues <- sapply(colnames(CNAtable), survwrapp, plot=TRUE)</pre>
```

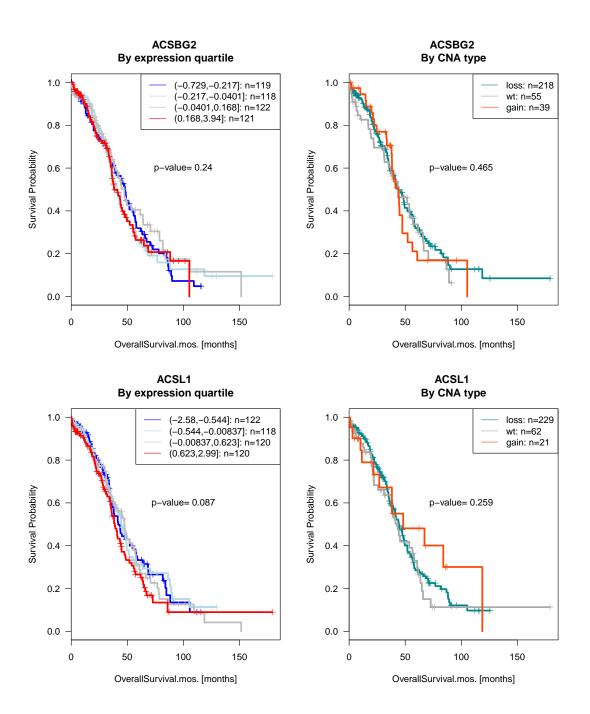


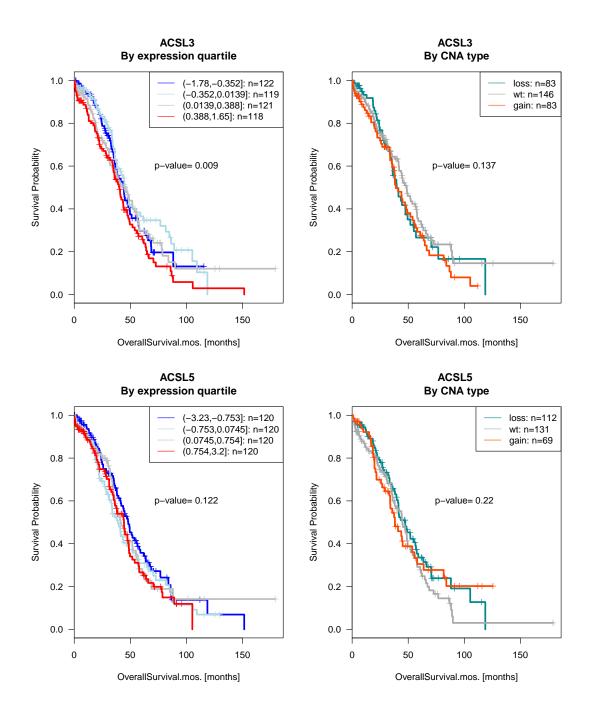


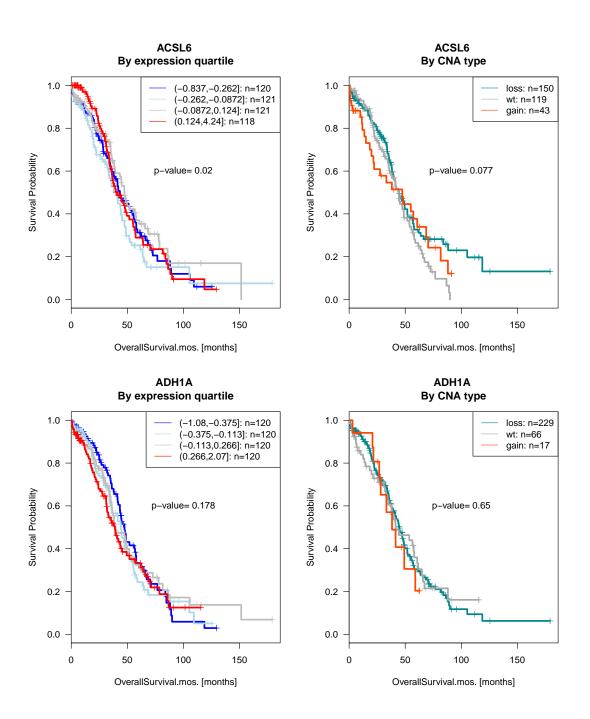


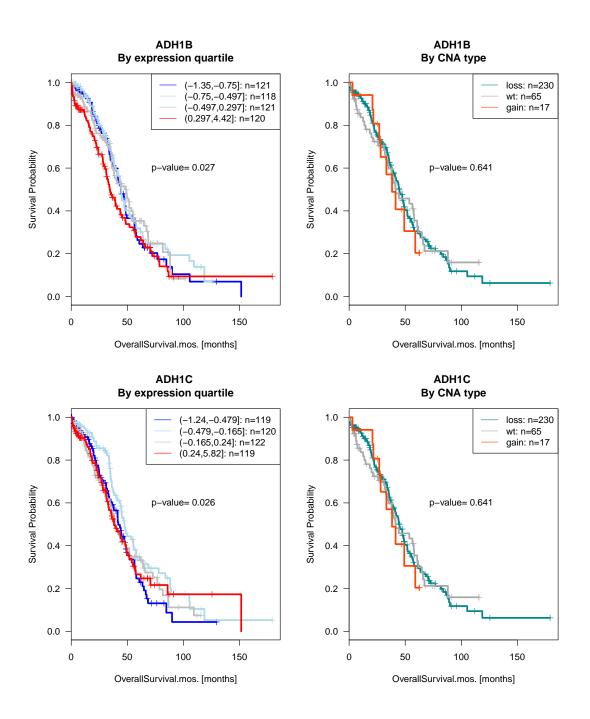


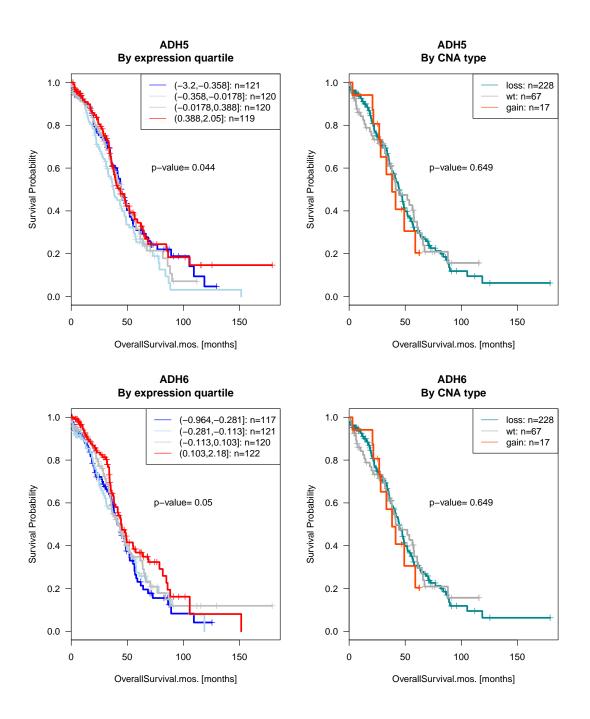


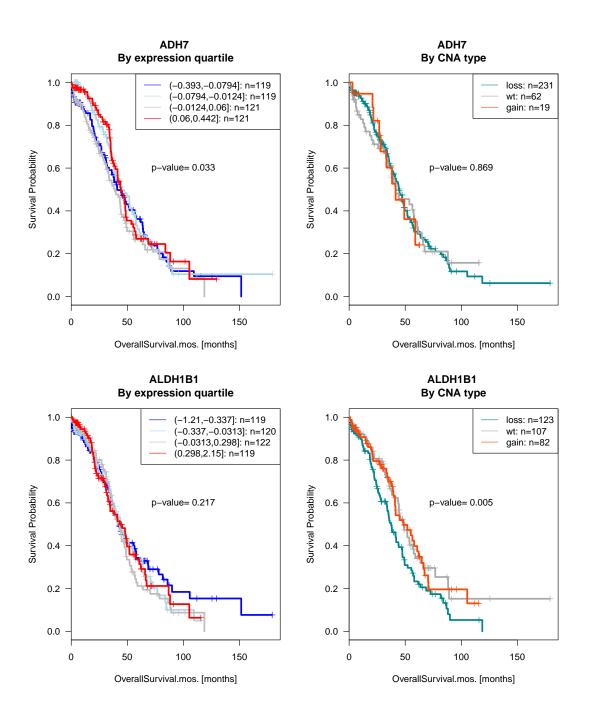


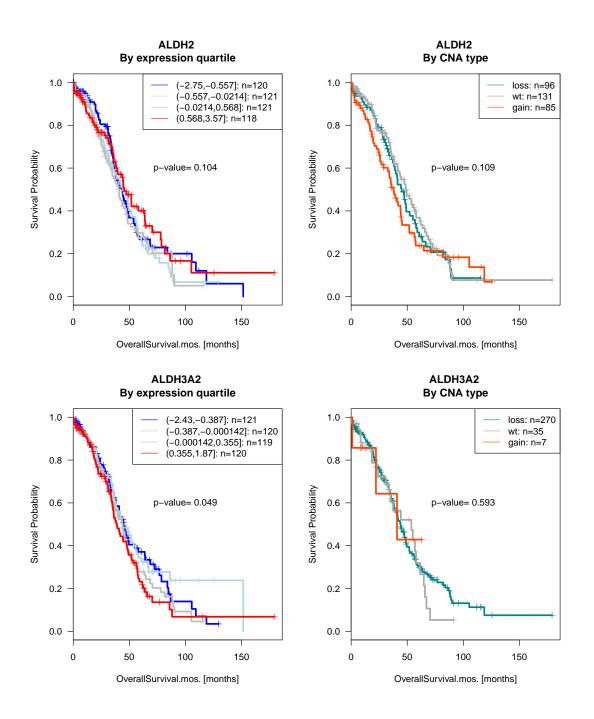


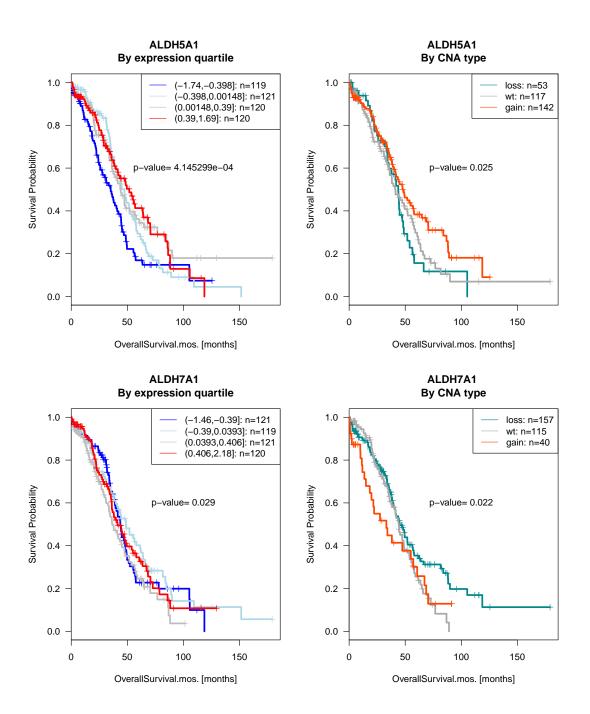


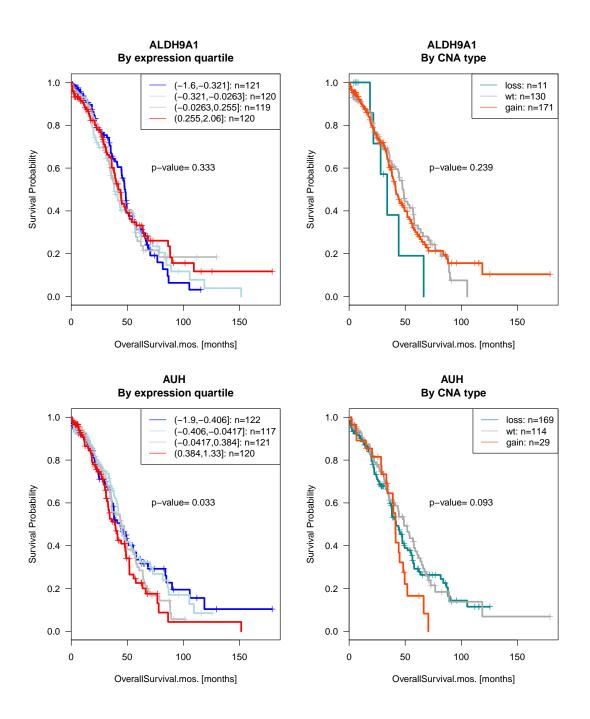


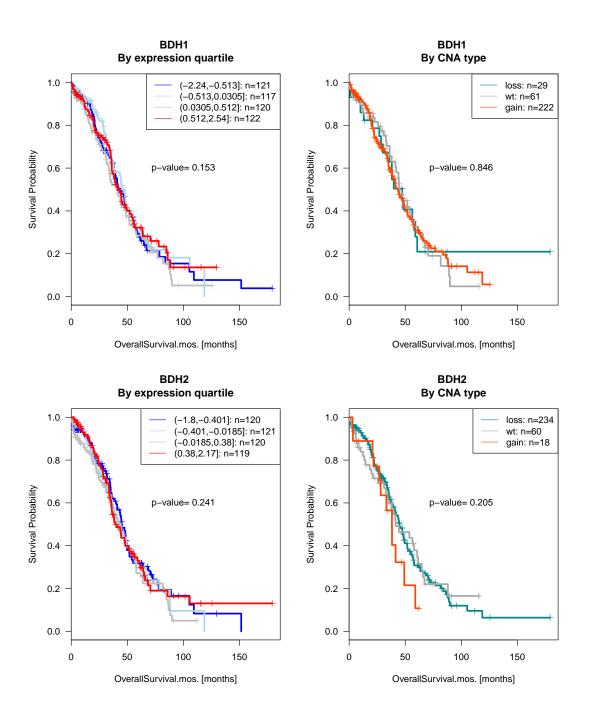


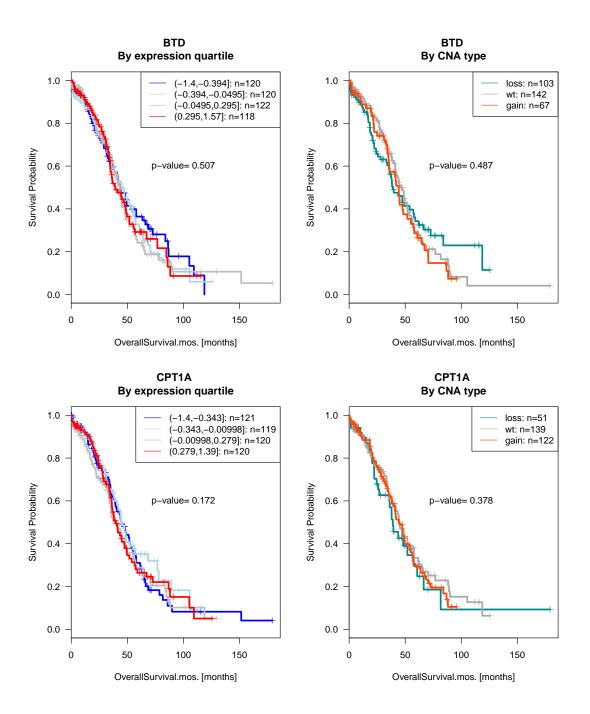


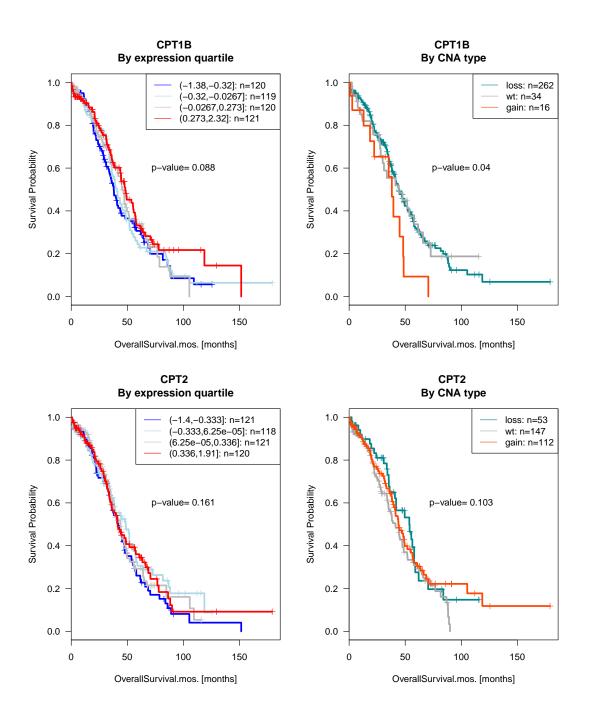


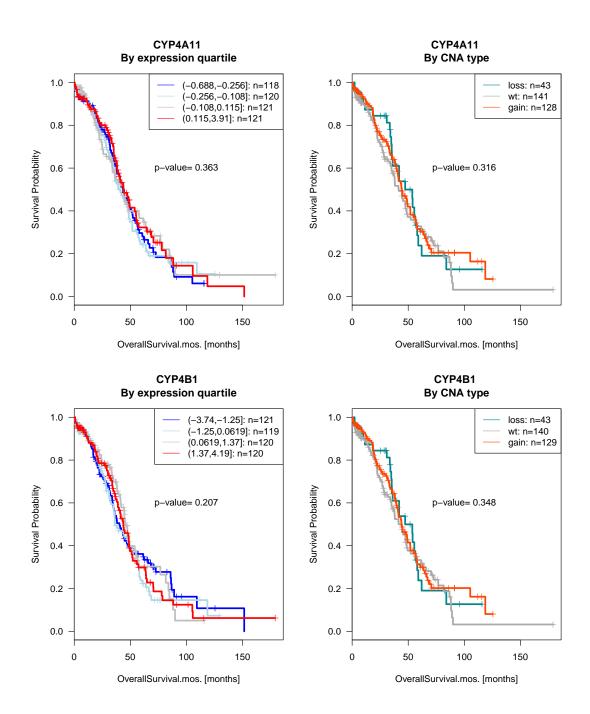


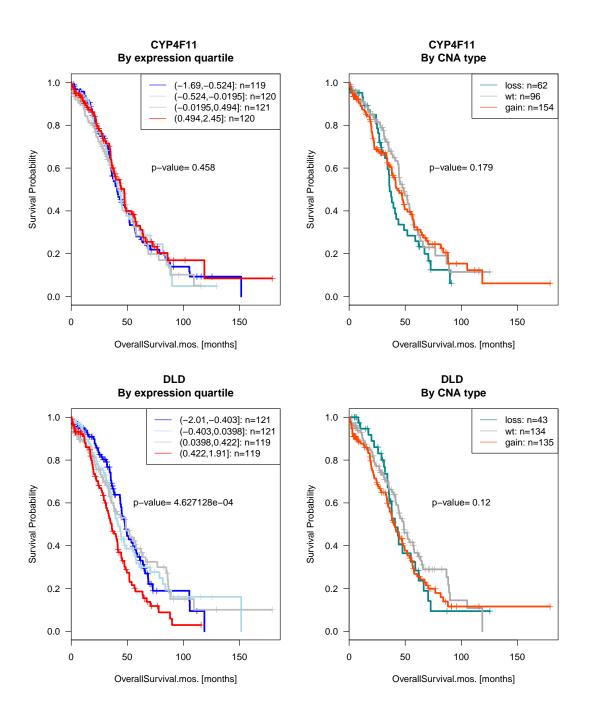


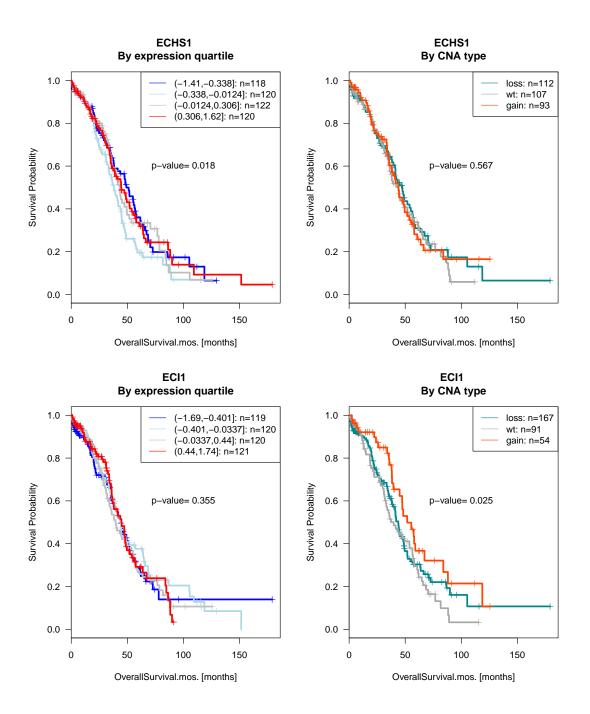


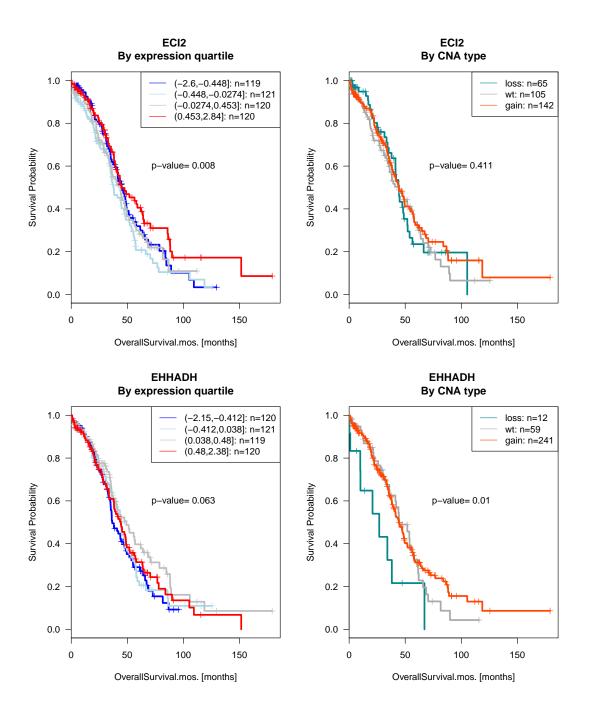


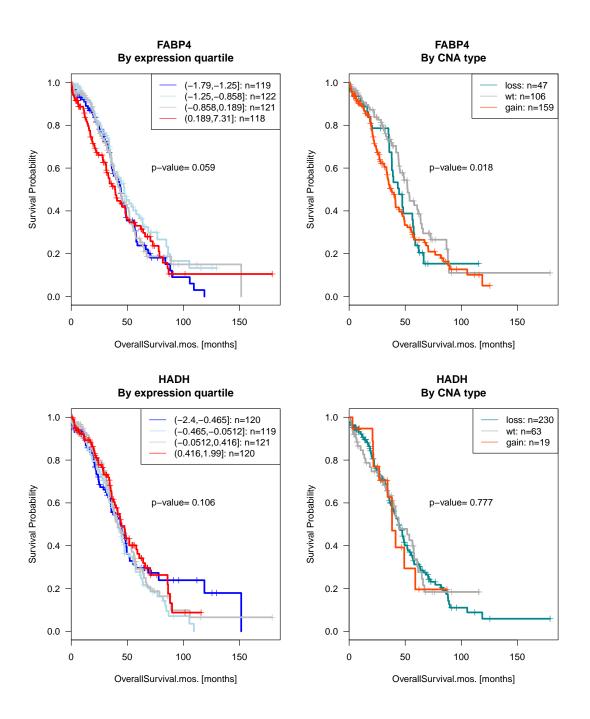


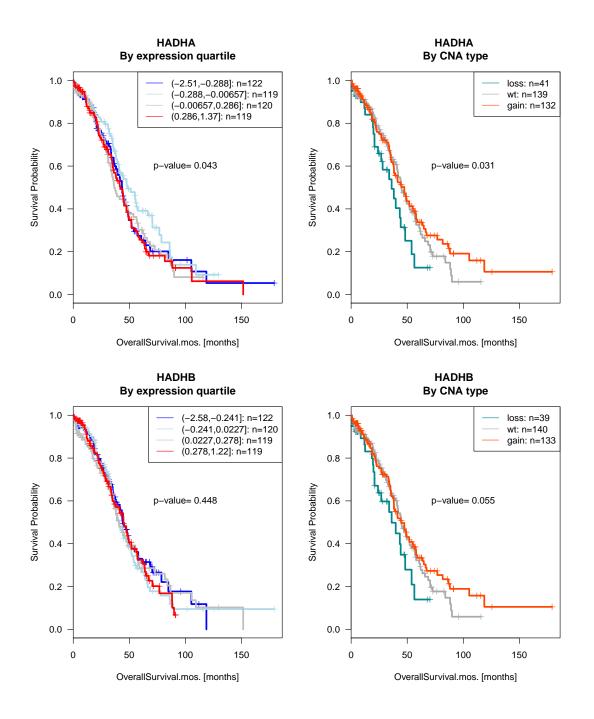


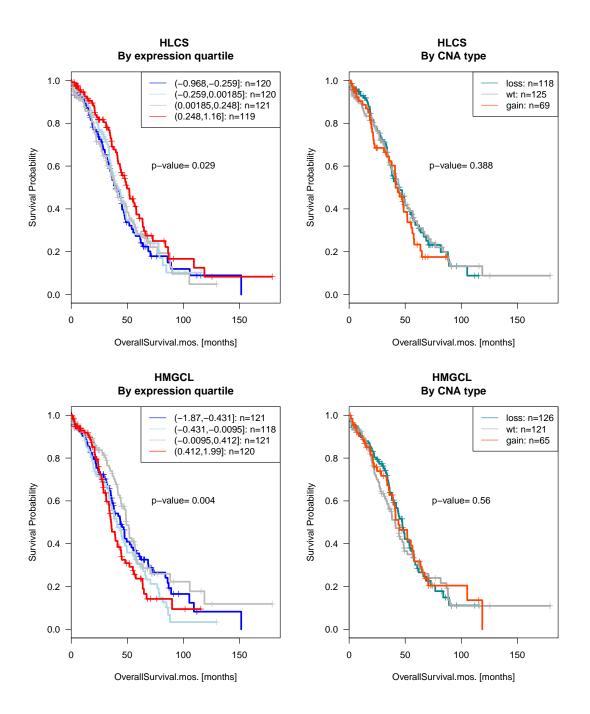


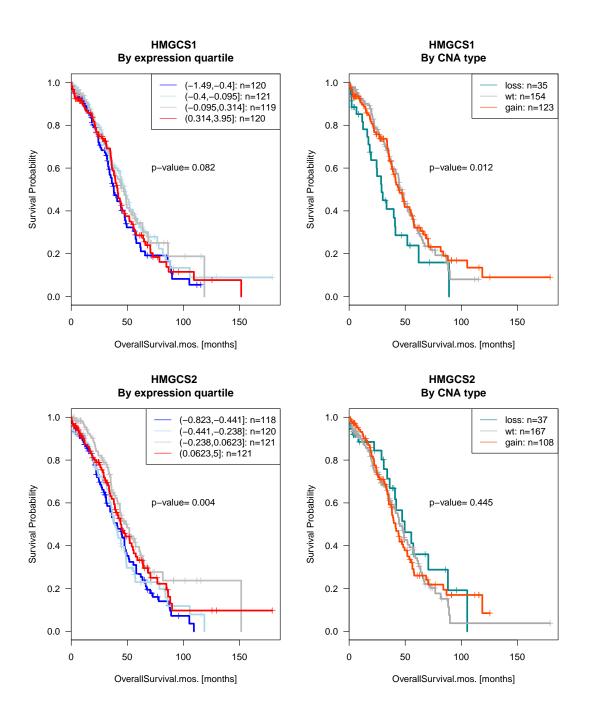


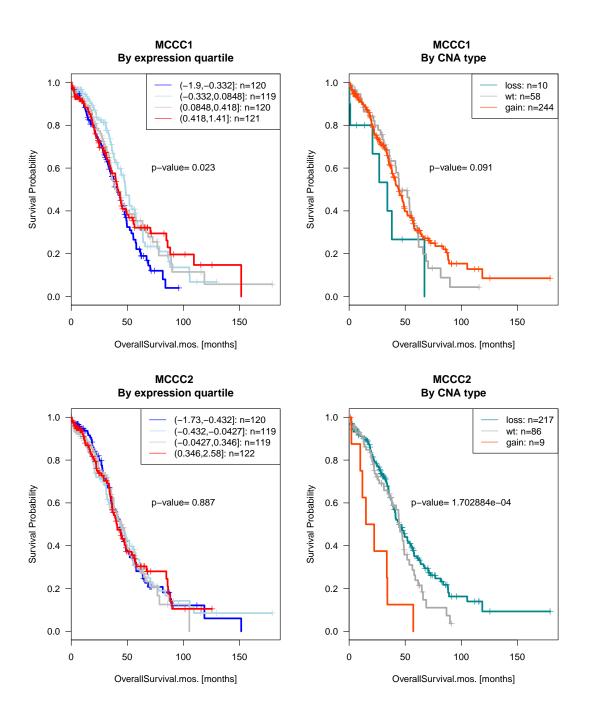


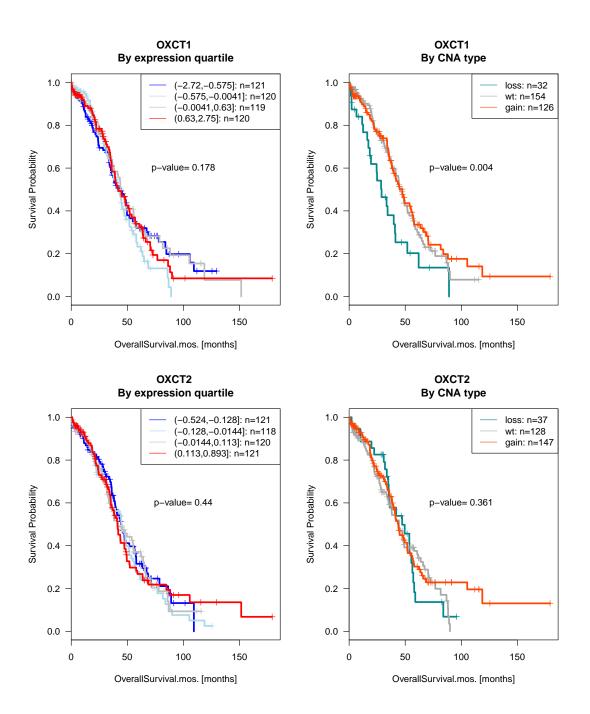


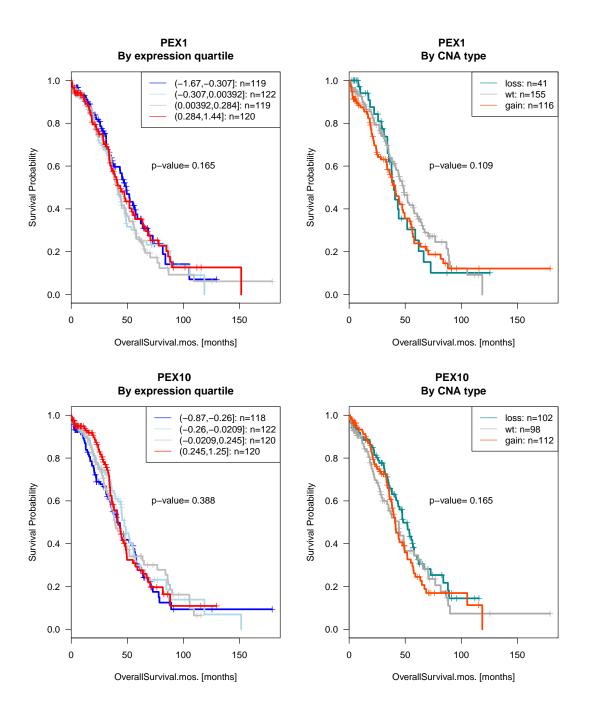


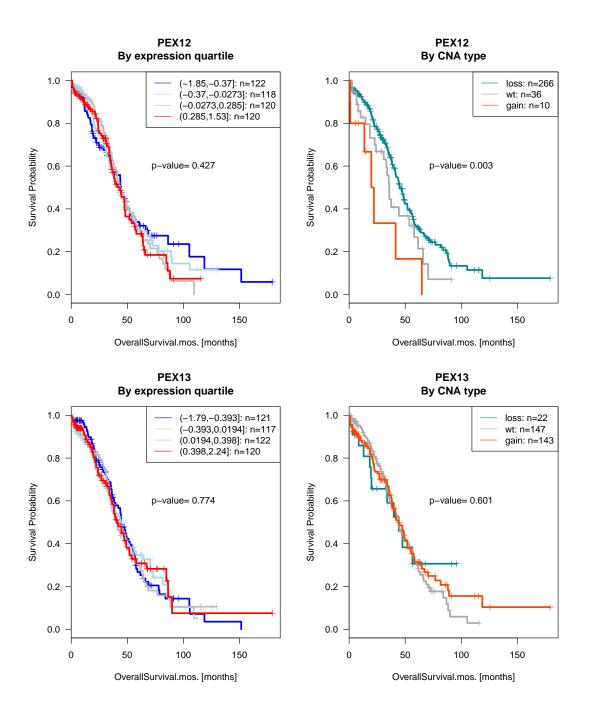


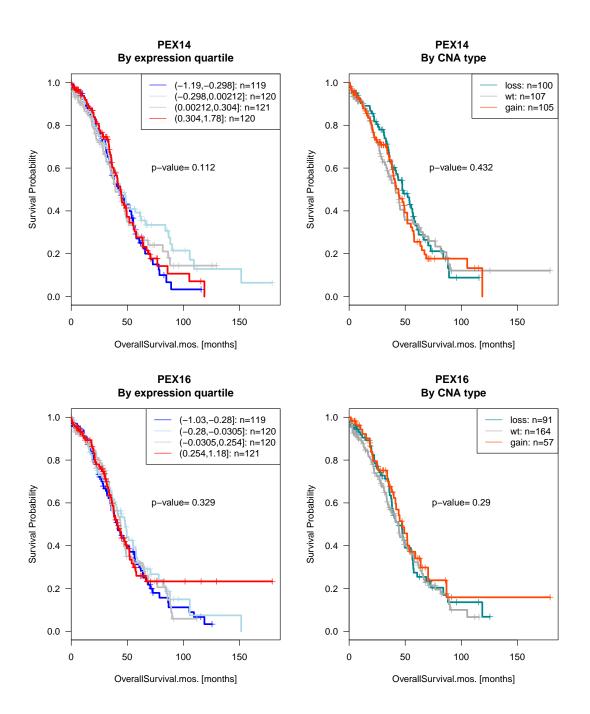


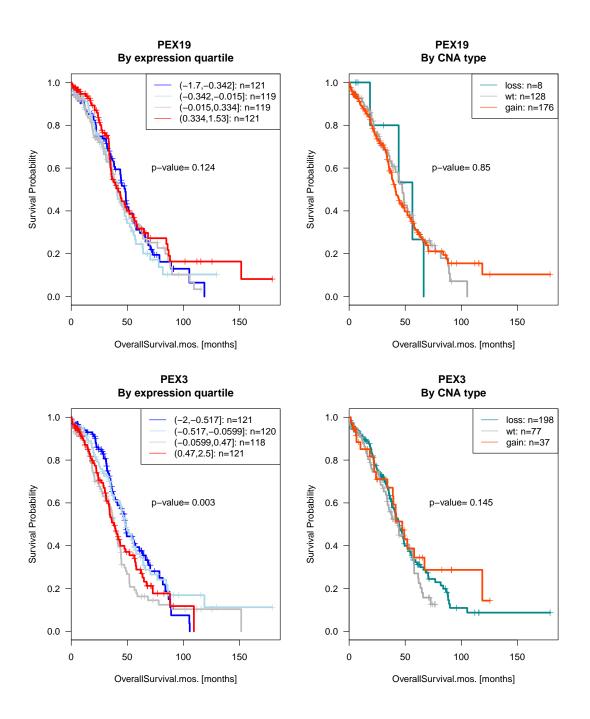


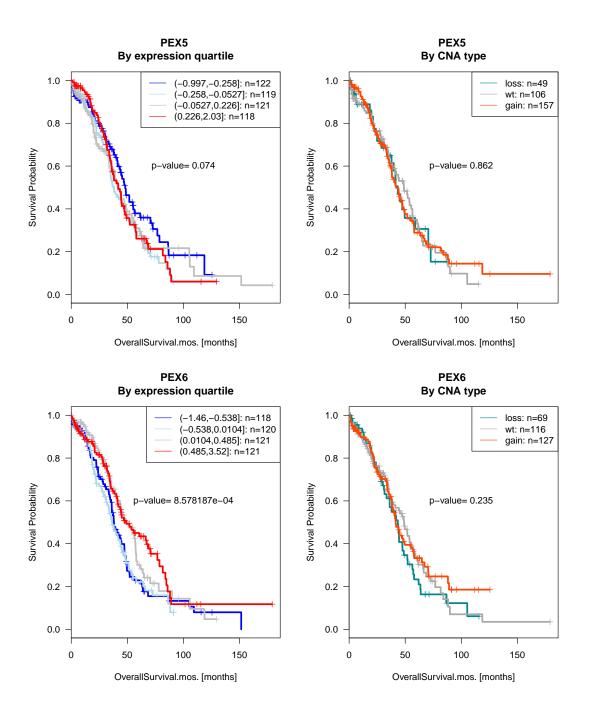


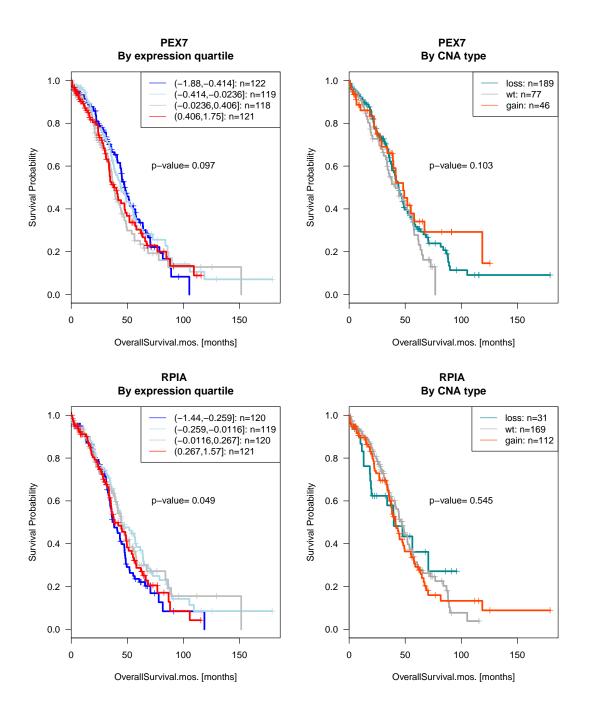


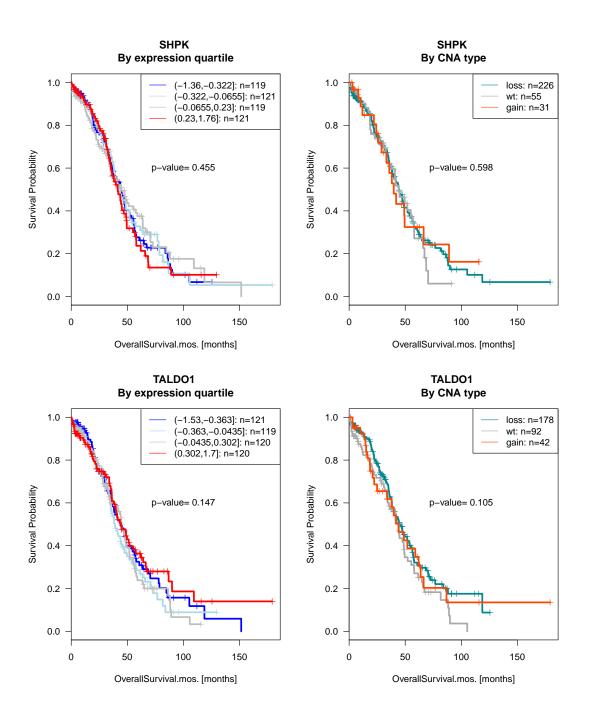


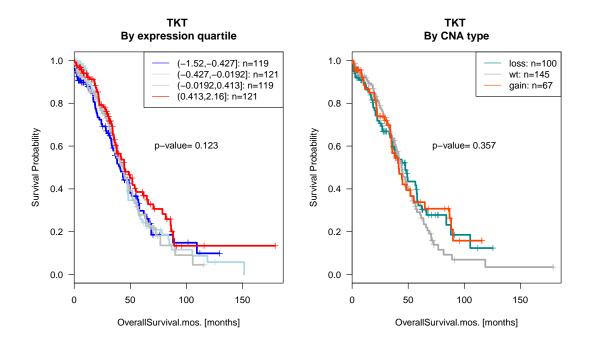












4 Session info: R-packages and their versions used for this analysis

```
## R version 3.1.2 (2014-10-31)
## Platform: x86_64-apple-darwin10.8.0 (64-bit)
##
## locale:
## [1] en_GB.UTF-8/en_GB.UTF-8/en_GB.UTF-8/C/en_GB.UTF-8/en_GB.UTF-8
##
## attached base packages:
## [1] stats graphics grDevices utils datasets methods base
##
## other attached packages:
## [1] survival_2.38-1 knitr_1.9
##
## loaded via a namespace (and not attached):
## [1] evaluate_0.6 formatR_1.1 highr_0.4.1 splines_3.1.2 stringr_0.6.2
## [6] tools_3.1.2
```