

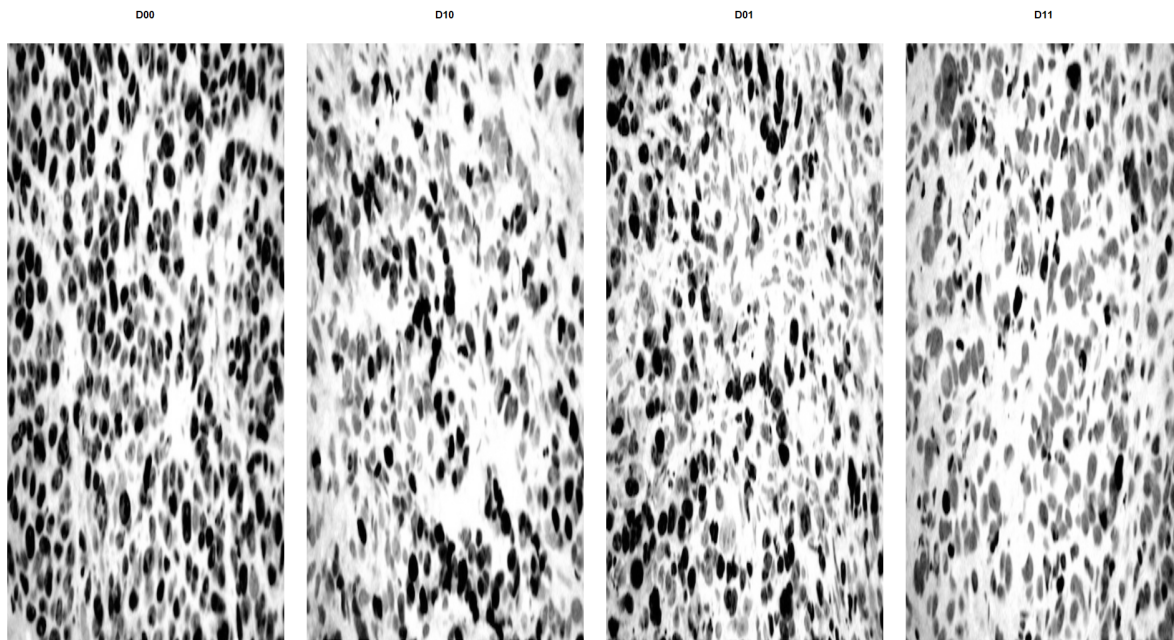
### Exercise 4.1.1

The grayscale images can be produced using the following R code:

```
1 img = rep(list(NA),4)
2 img[[1]] = jpeg::readJPEG("D00.jpg")
3 img[[2]] = jpeg::readJPEG("D10.jpg")
4 img[[3]] = jpeg::readJPEG("D01.jpg")
5 img[[4]] = jpeg::readJPEG("D11.jpg")
6
7 # Plots of color images
8 par(mfrow=c(1,4),mar=c(0,0,1,0))
9 for(i in 1:4){
10     plot(0:1,0:1,cex=0,axes=FALSE,main=c("D00","D10","D01","D11")[i])
11     rasterImage(img[[i]],0,0,1,1)
12 }
13
14 # Produce grayscale images
15 get.gray = function(img, w=c(0.22,0.71,0.07)){
16     img[, ,1]*w[1] + img[, ,2]*w[2] + img[, ,3]*w[3]
17 }
18 img0 = rep(list(NA),4)
19 for(i in 1:4){
20     img0[[i]] = get.gray(img[[i]])
21 }
22
23 # Plot of grayscale images
24 par(mfrow=c(1,4),mar=c(0,0,1,0))
25 for(i in 1:4){
26     plot(0:1,0:1,cex=0,axes=FALSE,main=c("D00","D10","D01","D11")[i])
27     rasterImage(img0[[i]], 0, 0, 1, 1)
28 }
```

### Exercise 4.1.1 (Cont'd)

The following plot will be produced:

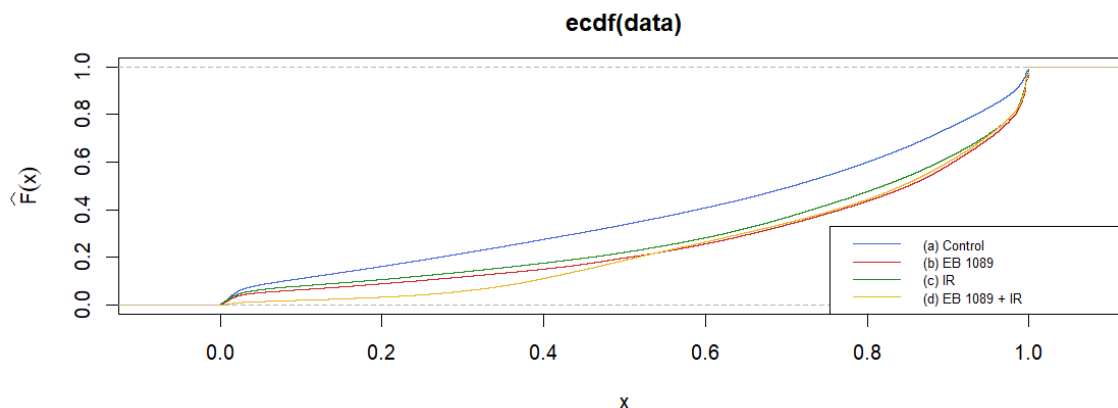


### Exercise 4.1.2

The empirical cdfs of the grayscale pixel values can be produced using the following R code:

```
1 col = c("royalblue", "firebrick3", "forestgreen", "goldenrod2")
2 par(mfrow=c(1,1), mar=c(4.5, 5, 3, 2))
3 for(j in 1:4){
4   data = c(img0[[j]])
5   plot(ecdf(data),
6        col=col[j], add=j>1, ylab=expression(widehat(F)(x)), xlab=expression(x))
7 }
8 treatment = c("(a) Control", "(b) EB 1089", "(c) IR", "(d) EB 1089 + IR")
9 legend("bottomright", treatment, col=col, lwd=1, bg="white", cex=0.65)
```

The following plot will be produced:



### Exercise 4.1.3

The null hypothesis is

$$H_0 : F_B(\cdot) = F_C(\cdot)$$

and the alternative hypothesis is

$$H_1 : F_B(\cdot) \neq F_C(\cdot)$$

### Exercise 4.1.4

The hypotheses can be tested using the following R code:

```
1 treatment_b = c(img0[[2]])  
2 treatment_c = c(img0[[3]])  
3 ks.test(treatment_b, treatment_c)$p.value # p-value < 2.2e-16
```

As the computed p-value  $\approx 0 < \alpha = 0.05$ , we reject  $H_0$ . So, it is concluded that the cancer treatments (b) EB 1089 and (c) IR have different effects on killing the cancer cells.

**Exercise 4.2.1**

$$\begin{aligned}
P(F_A(A) < u) &= P(A \leq F_A^{-1}(u)) \\
&= F_A(F_A^{-1}(u)) \\
&= u \\
\therefore F_A(A) &= U \sim \text{Unif}(0, 1) \\
A &= F_A^{-1}(U)
\end{aligned}$$

Simiarly,

$$\begin{aligned}
P(F_D(D) < u) &= P(D \leq F_D^{-1}(v)) \\
&= F_D(F_D^{-1}(v)) \\
&= v \\
\therefore F_D(D) &= V \sim \text{Unif}(0, 1) \\
D &= F_D^{-1}(V)
\end{aligned}$$

**Exercise 4.2.2**

We try to assume the contrary:

$$\begin{aligned}
F_A^{-1}(U) &\geq F_D^{-1}(U) \\
\therefore F_A(F_A^{-1}(U)) &\geq F_A(F_D^{-1}(U)) \text{ as } F_A(\cdot) \text{ is a strictly increasing function} \\
U &\geq F_A(F_D^{-1}(U))
\end{aligned}$$

Besides, the question suppose that

$$\begin{aligned}
F_A(t) &> F_D(t) \\
\therefore F_A(F_D^{-1}(U)) &> F_D(F_D^{-1}(U)) \\
F_A(F_D^{-1}(U)) &> U
\end{aligned}$$

The above inequalities will lead to contradiction:

$$U \geq F_A(F_D^{-1}(U)) > U$$

Thus, the contrary assumed is not true, therefore:

$$F_A^{-1}(U) < F_D^{-1}(U)$$

**Exercise 4.2.3**

$$\begin{aligned}
F_A^{-1}(U) &< F_D^{-1}(U) \\
\int_0^1 F_A^{-1}(U) du &< \int_0^1 F_D^{-1}(U) du \\
E(A) &= E(D)
\end{aligned}$$

### Exercise 4.2.4

The hypotheses can be tested using the following R code:

```
1 A = c(img0[[1]])      # the pixel value of image corresponding to treatment A
2 D = c(img0[[4]])      # the pixel value of image corresponding to treatment D
3 ks.test(A, D, alternative="greater")$p.value # p-value < 2.2e-16
```

As the computed p-value  $\approx 0 < \alpha = 0.05$ , we reject  $H_0$ . So, it is concluded that treatment D is more effective in killing cancer cells than treatment A.