Denoising Images

Applications of the Walsh-Hadamard Transform

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Abstract

The Hadamard transform is an example of a generalized Fourier transform. It has numerous applications in signal processing, data compression, and image/video coding to name a few. The Fast Walsh-Hadamard transform (FWHT) is an efficient algorithm for denoising images by using various methods of pixel thresholding. In this note, we explore how this might be implented on a noisy image.

Hadamard Matrices

For Hadamard matrices H_m and H_n we have the following useful recursive properties (note that since m and n are powers of 2, the matrices are symmetric),

$$H_m H_m^T = 2^k I (1)$$

$$H_m^{-1} = \frac{1}{m} H_m \tag{2}$$

Proposition 1.

If we let $\hat{Z} = H_m Z H_n$, then $Z = H_m \hat{Z} H_n / (mn)$

Proof

$$\hat{Z} = H_m Z H_n$$

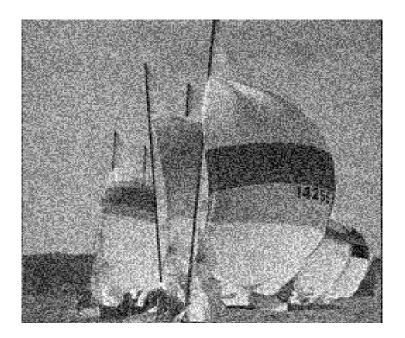
$$\iff (1/m) H_m \hat{Z}(1/n) H_n = H_m^{-1} H_m Z H_n H_n^{-1}$$

$$\iff H_m \hat{Z} H_n / (mn) = Z$$

Loading in the image as a matrix:

```
boats <- matrix(scan("boats.txt"), ncol=256, byrow=T)
image(boats, axes=F, col=grey(seq(0,1,length=256)), main = "Original")</pre>
```

Original



Below is an implementation of the FWHT.

```
fwht2d <- function(x) {</pre>
  h <- 1
  length <- ncol(x)</pre>
  while (h < length) {</pre>
    for (i in seq(1, length, by=h*2)) {
       for (j in seq(i,i+h-1)) {
         a \leftarrow x[,j]
         b \leftarrow x[,j+h]
         x[,j] <- a + b
         x[,j+h] \leftarrow a - b
    }
    h <- 2*h
  h <- 1
  length <- nrow(x)</pre>
  while (h < length) {</pre>
    for (i in seq(1, length, by=h*2)) {
       for (j in seq(i,i+h-1)) {
         a <- x[j, ]
         b <- x[j+h,]
         x[j, ] \leftarrow a + b
```

```
x[j+h, ] <- a - b
}
h <- 2*h
}
x
</pre>
```

Below are the functions to perform soft and hard thresholding.

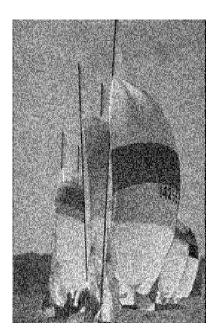
```
hard.thresh <- function(x, lambda) {ifelse(abs(x) > lambda, x, 0)}
soft.thresh <- function(x, lambda) {
   sign(x) * (abs(x) >= lambda) * (abs(x) - lambda)
}
```

The function below will divide the original image up into subimages depending on the function parameter "size"; referring to n where $n=2^k$, $2 \le k \le 8$ so that the subimages are n by n pixel sub-images.

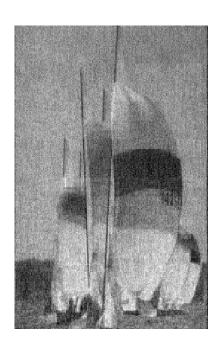
```
sub.im.transform <- function(image, size, Th.type="soft", thresh) {</pre>
  # image size must be square
  N <- ncol(image)
  if ((N %% size) != 0) {
    stop("image size is not divisible by specified sub-image size")
  # create a list of submatrices to apply the transform to
  flat <- as.vector(image)</pre>
  subimages <- list()</pre>
  A <- N^2/size^2 # index span for each subimage
  for (j in 0:(A-1)) {
    subimages[[j+1]] \leftarrow matrix(flat[(j*(size^2)+1):((j+1)*(size^2))], ncol=size)
  # apply transform to each sub-image and reconstruct the original matrix/image
  subimages.hat <- list()</pre>
  i <- 1
  for (im in subimages) {
    Zhat <- fwht2d(im)
    if (Th.type == "soft") {
      Zhat.star <- soft.thresh(Zhat, thresh)</pre>
    } else {
      Zhat.star <- hard.thresh(Zhat, thresh)</pre>
    }
    Zstar <- fwht2d(Zhat.star) / size</pre>
    subimages.hat[[i]] <- Zstar</pre>
    i <- i + 1
  }
  # reconstruct original image
  original <- c()
  for (i in 1:A) {
    original <- c(original, as.vector(subimages.hat[[i]]))</pre>
  original <- matrix(original, ncol = 256)
  original
```

Results

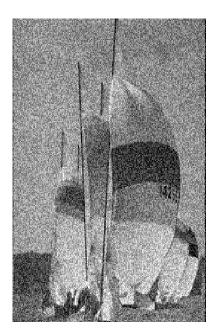
Original



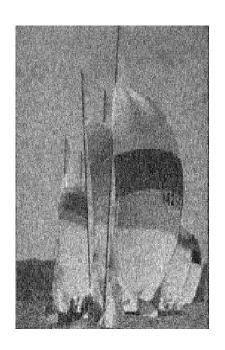
Soft Threshold: lam=7



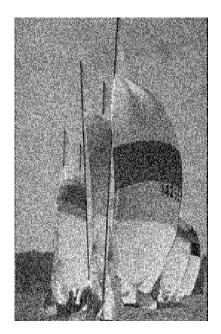
Original



Hard Threshold: lam=6



Original



Soft Threshold: lam=0.8

