

# Sliding DCT

**Mathematical Models and Methods for Image Processing**

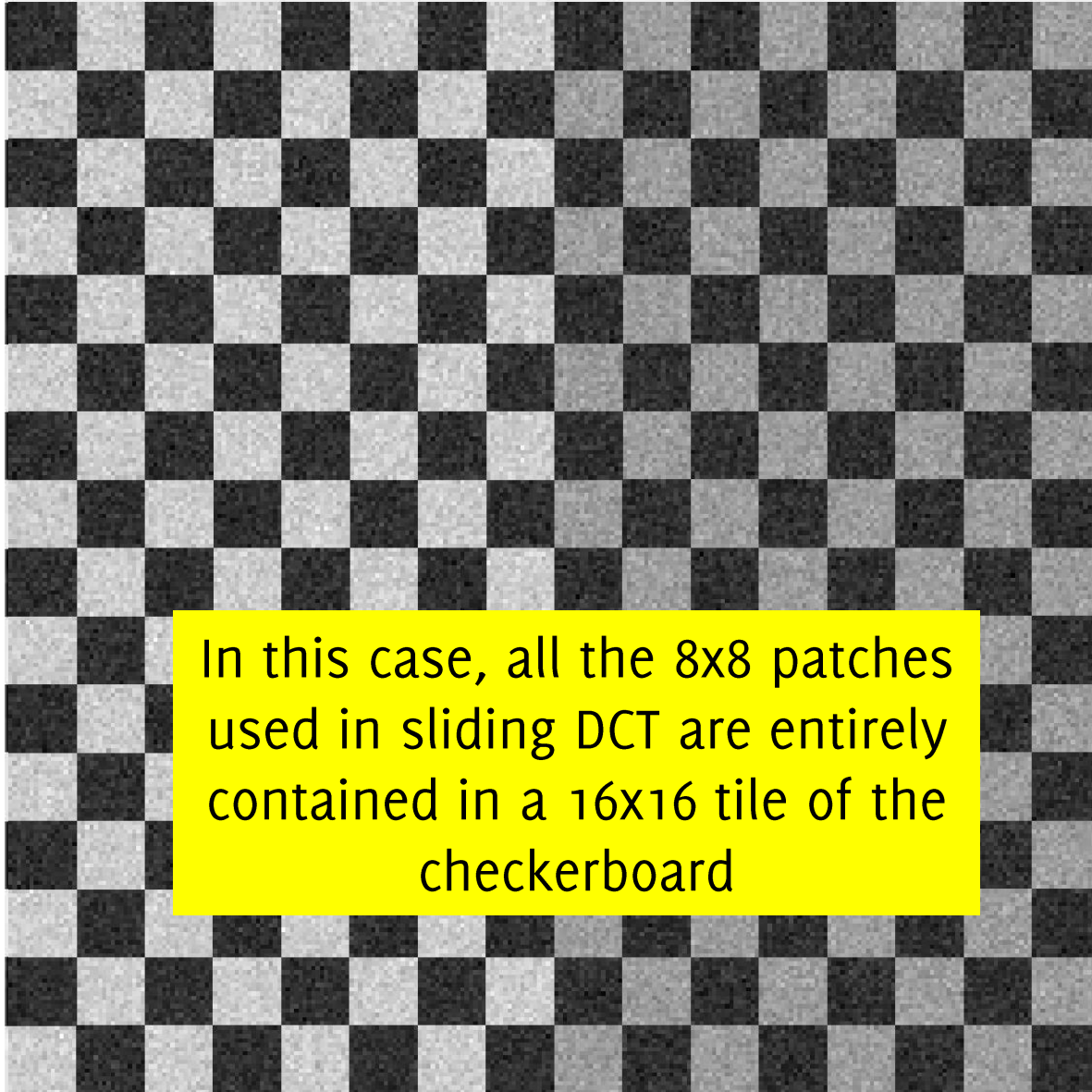
Giacomo Boracchi

<https://boracchi.faculty.polimi.it/teaching/MMMIP.htm>

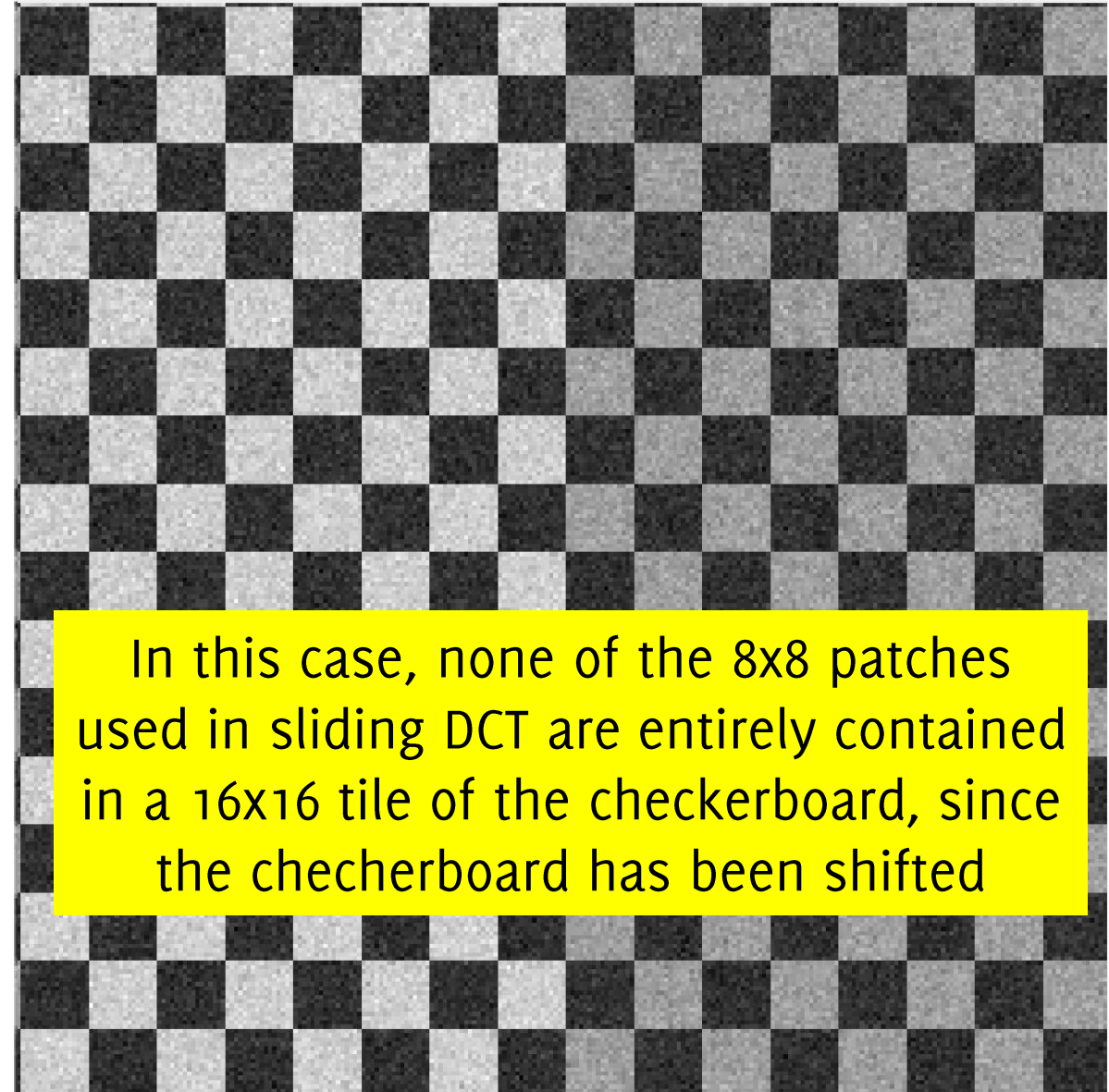
February 25th 2025

# Sliding DCT

## Original Checkerboard



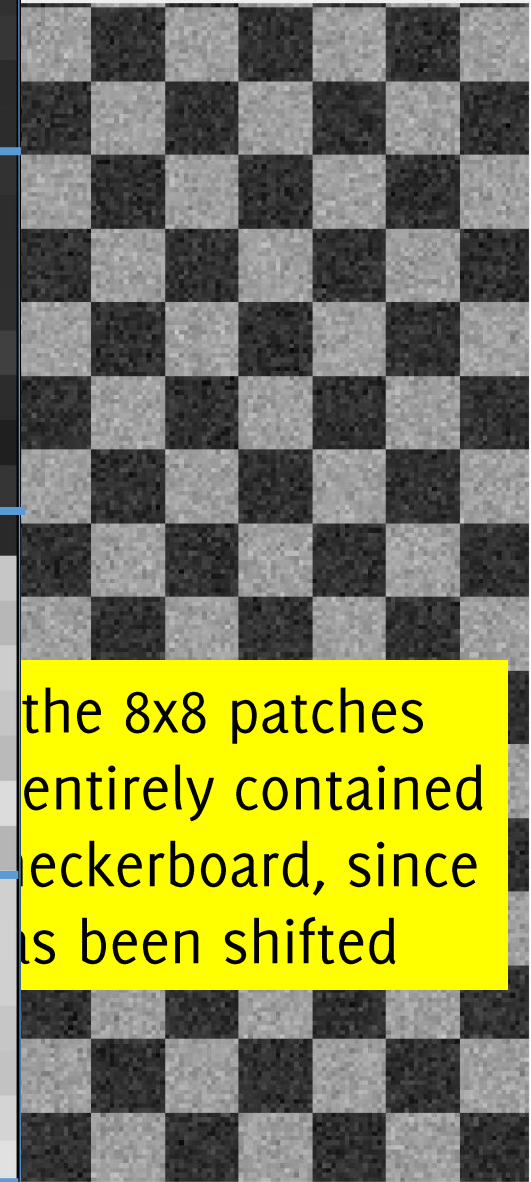
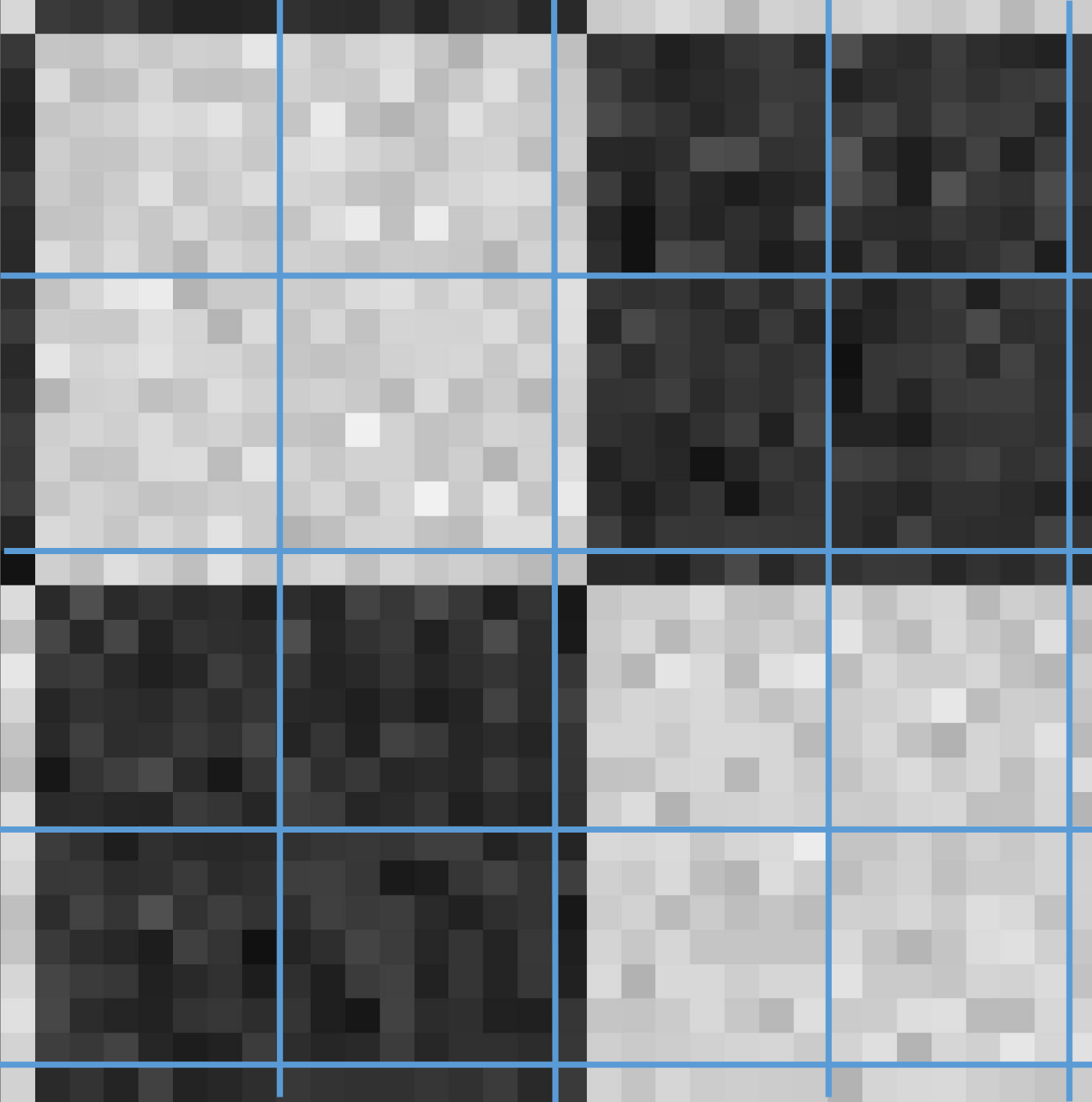
## Shift [1 row, 1 col]



# Original Checkerboard

Estimated

R : 23.645

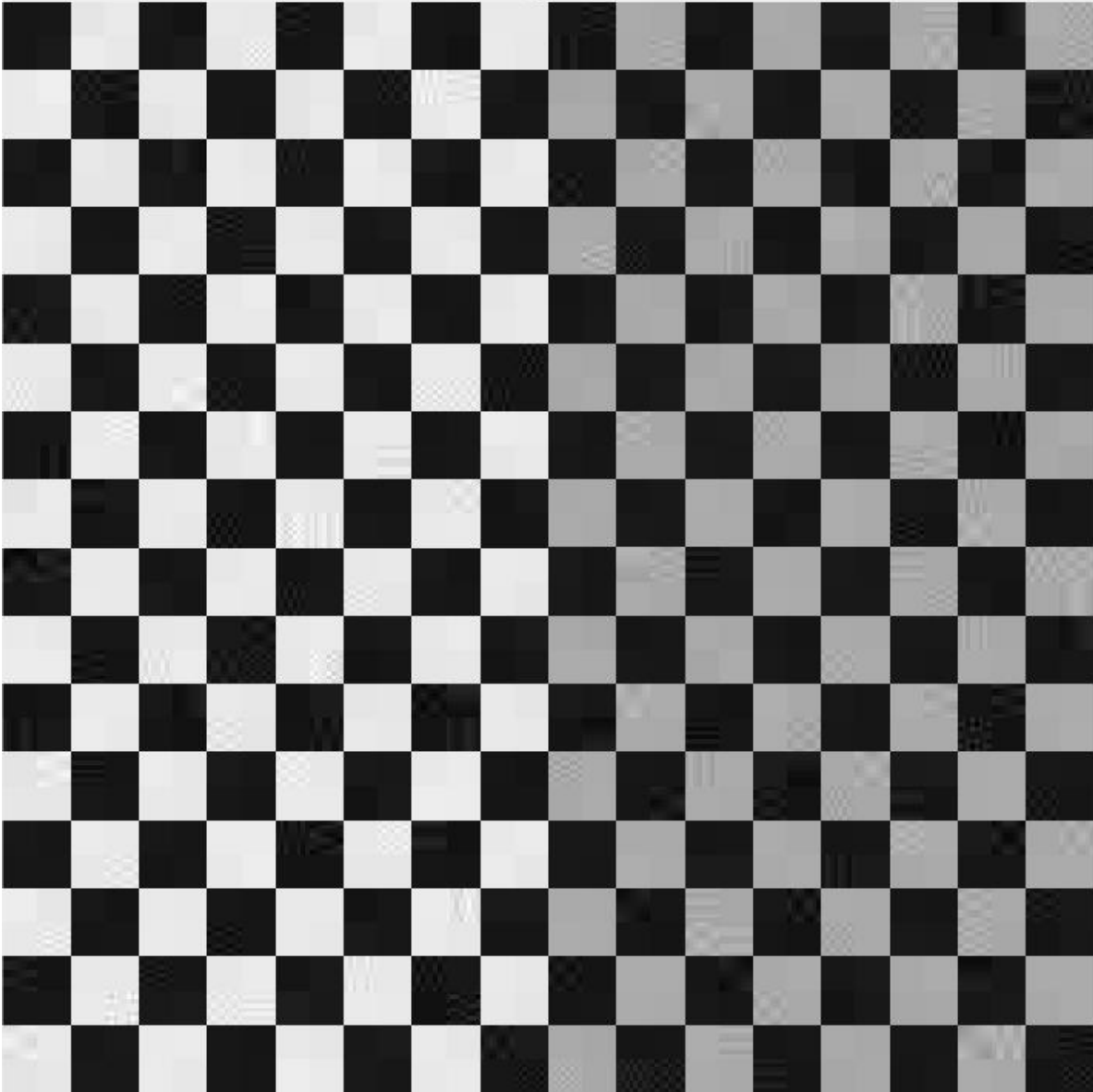


In this case  
used in slice  
contained in  
ch

the 8x8 patches  
entirely contained  
checkerboard, since  
has been shifted

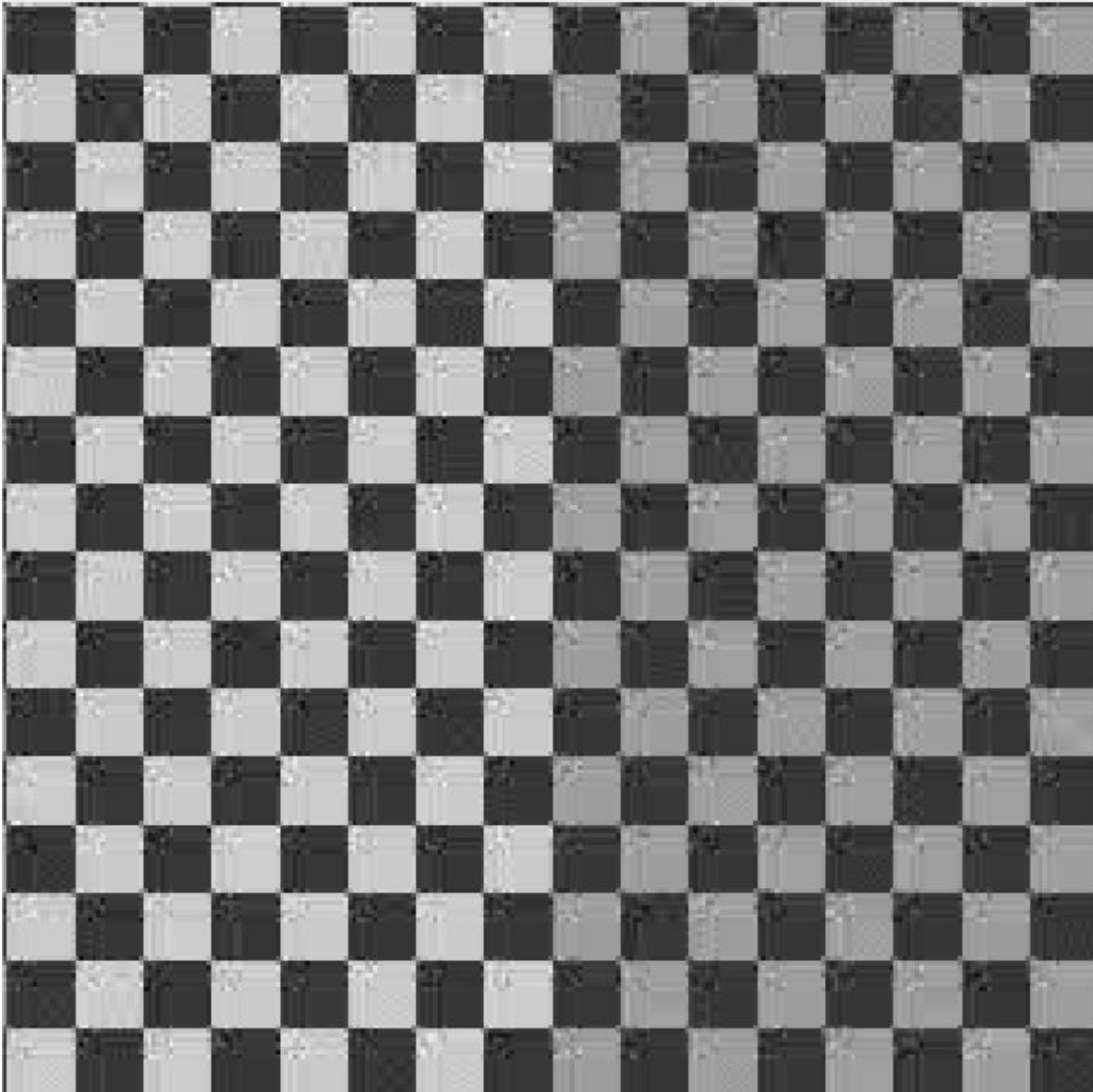
# Original Checkerboard

Estimated Image, PSNR : 35.747



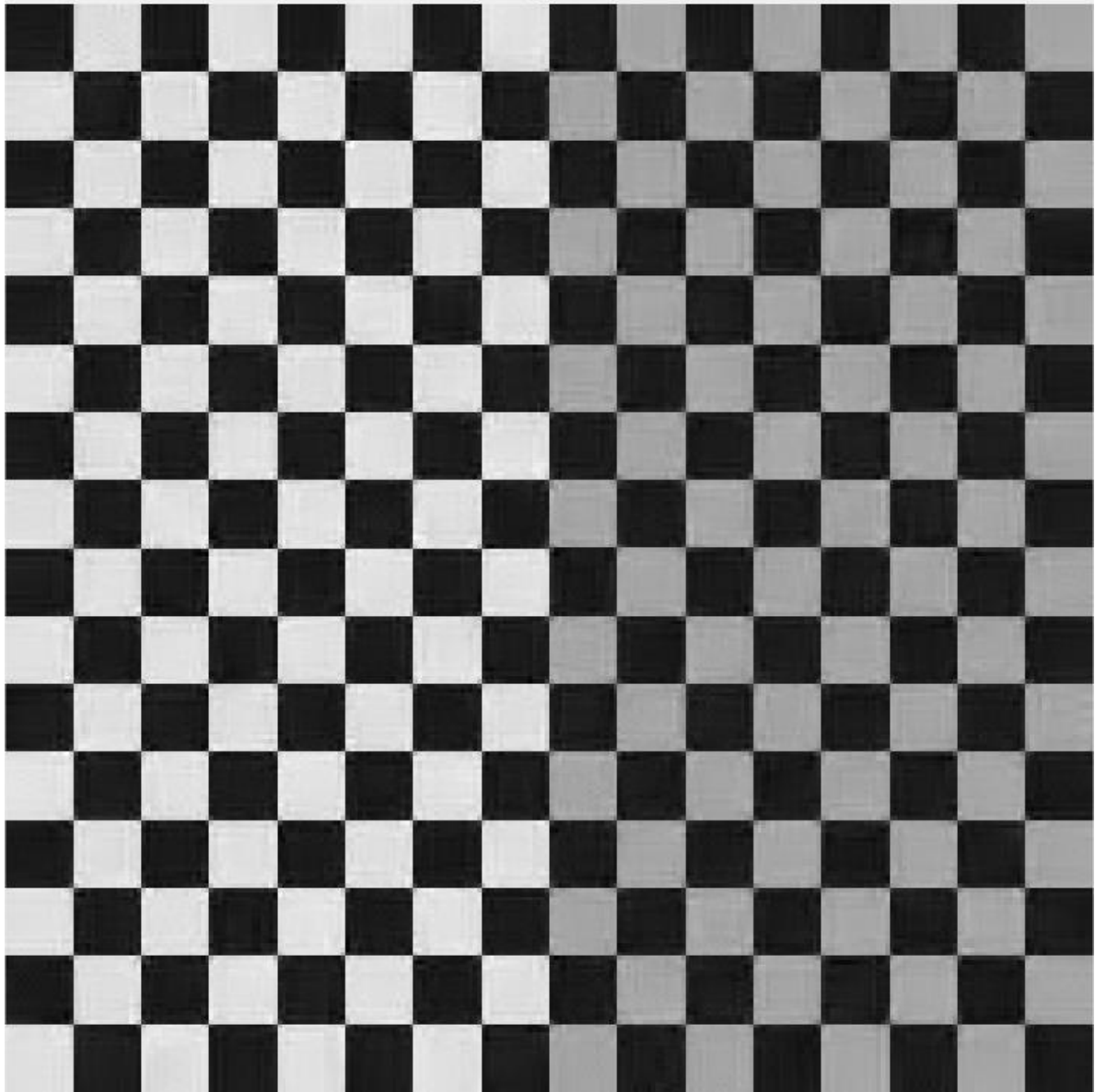
# Shift [1 row, 1 col]

Estimated Image, PSNR : 23.645



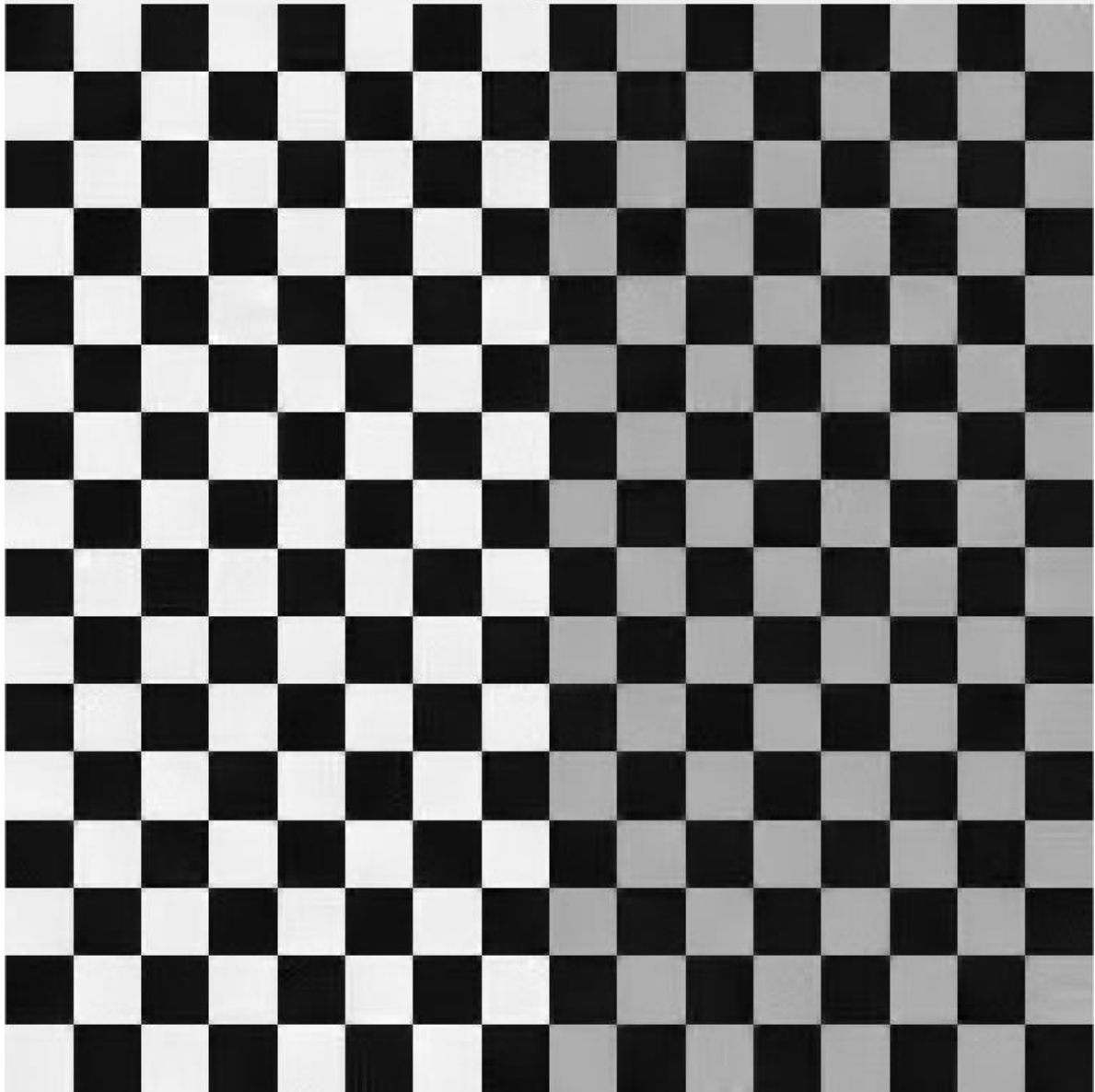
# Uniform Weights

Estimated Image, PSNR : 30.582



# Sparsity-aware

Estimated Image, PSNR : 35.656



# Assignment

1. Implement the sliding-DCT denoising using
  - no aggregation (operate on non-overlapping tiles)
  - aggregation using uniform weights
  - aggregation using weights inversely proportional to patch sparsity in DCT domain.
2. Test the three algorithms on both checkerboard and cameraman image
3. Test how much the choice of the threshold  $\tau$  influences the denoising performance. Observe the resulting image when:
  - $\tau \ll 3\sigma$
  - $\tau \gg 3\sigma$

This is very important to understand how important is the choice of the threshold

Original image



Noisy image, PSNR = 22.09





HT Estimate, PSNR = 29.15



Wiener Estimate, PSNR = 29.41

