

ECE 253 Homework 4

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Problem 1 2D Sampling and Aliasing

Part(a)

```
clear; close all; clc;  
[x y] = meshgrid(0:256,0:256);  
figure, subplot(1,2,1), imshow(x, [0, 255])  
subplot(1,2,2), imshow(y, [0, 255]), title('x, y ')
```



- We are given the sampling period $T = 1$, thus the sampling frequency $F_s = 1$.
- x and y are the spatial coordinates of the original image given by the mesh grid in matlab
- u and v are the spatial frequency coordinates of the Fourier transform of the image.

```
z1 = cos ( 2 * pi * 1/32 .* x - 2 * pi * 1/128 .* y);  
figure, imshow(z1), title('z1 matrix/image');
```



Spatial Frequencies for z1.

- In x-direction = $1/32$
- In y-direction = $1/128$

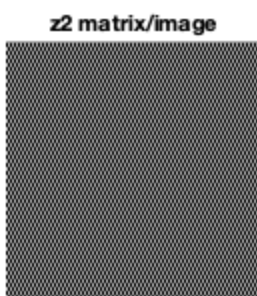
Image can be reconstructed exactly from it's samples if Cutoff freq $\leq 1/2$ sampling freq F_s

- If equality holds, then the sampling is at the Nyquist rate
- If Δx and Δy are smaller than required, the image is called oversampled
- If they are larger than required, the image is undersampled

In terms of spatial frequency and the sampling frequency,

- $2 * 1/32 = 1/16 < F_s = 1 \implies$ oversampled in x-direction
- $2 * 1/128 = 1/64 < F_s = 1 \implies$ oversampled in y-direction

```
z2 = cos ( 2 * pi * 1/4 .* x - 2 * pi * 7/8 .* y);
figure, imshow(z2), truesize, title('z2 matrix/image');
```



Spatial Frequencies for z2.

- In x-direction = $1/4$
- In y-direction = $7/8$

In terms of spatial frequency and the sampling frequency,

z3 matrix/image



- $2 * F_{xc} = 1/2 < F_s = 1 \implies$ oversampled in x-direction
- $2 * F_{yc} = 7/4 > F_s = 1 \implies$ undersampled in y-direction

```
z3 = cos ( 2 * pi * 1/2 .* x - 2 * pi * 1/2 .* y);
figure, imshow(z3), title('z3 matrix/image');
```

z3 matrix/image

Spatial Frequencies for z3.

- In x-direction = $1/2$
- In y-direction = $1/2$

In terms of spatial frequency and the sampling frequency,

- $2 * F_{xc} = 1 = F_s = 1 \implies$ critically sampled in x-direction or sampling is at the Nyquist rate
- $2 * F_{yc} = 1 > F_s = 1 \implies$ critically sampled in y-direction or sampling is at the Nyquist rate

Part(b)

Since Cosine function is periodic with period 2π . we can use any frequencies which is generated by adding any multiple of 2π to each of the xy frequency components in z1 to obtain a identical sampled function to that of the sampled function z1. Therefore let us consider $a = F_{xc} + 2\pi$ and $b = F_{yc} + 2\pi$, taking the 2π in common we get $a = 33/32$ $b = 129/128$ Any $a = 1/32 + k 2\pi$ where $k \in \mathbb{Z}$, will give us the identical sampled function.

However, we have to note that the we want the sampled function to be aliased, this means we have to consider the inequality $a > 1/2$ and $b > 1/2$

```
z4 = cos ( 2 * pi * 33/32 .* x - 2 * pi * 129/128 .* y);
figure, imshow(z4), title('z3 matrix/image');
```



In the figure above it came be seen that z4 produces the same image as sampled in z1.

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Problem 2 Pre-filtering to Reduce Aliasing before Sampling

```
close all; clc; clear;
barbara = read(Tiff('barbara.tiff', 'r'));

figure,
subplot(1,2,1), imshow(barbara), title('barbara original');

baby = read(Tiff('baby.tiff', 'r'));
subplot(1,2,2), imshow(baby), title('baby original');
```



```
close all;
barbBig = barbara(1:512,37:548);
figure,
subplot(1,2,1), imshow(barbBig), title('barbBig 512 x 512 section');
babyBig = baby(201:712,201:712);
subplot(1,2,2), imshow(babyBig), title('babyBig 512 x 512 section');
```

barbBig 512 x 512 section



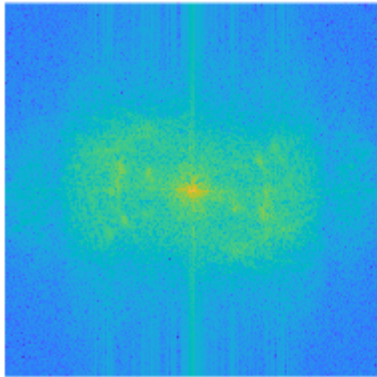
babyBig 512 x 512 section



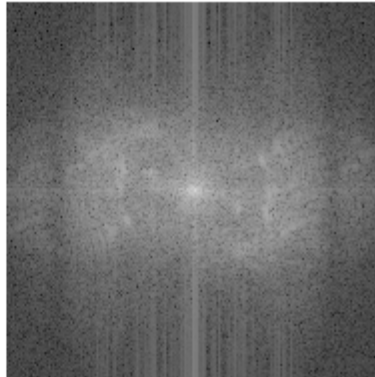
```
close all;
barbfft_1 = fftshift(abs(fft2(barbBig)).^2);
barbspectrum = log(barbfft_1 + 1);
figure,
subplot(1,2,1),
imshow(barbspectrum / max(max(barbspectrum))),
colormap("default"),
title('Power Spectrum of barbBig w/ color');

subplot(1,2,2),
imshow(barbspectrum / max(max(barbspectrum))),
title('Power Spectrum of barbBig');
```

Power Spectrum of barbBig w/ color



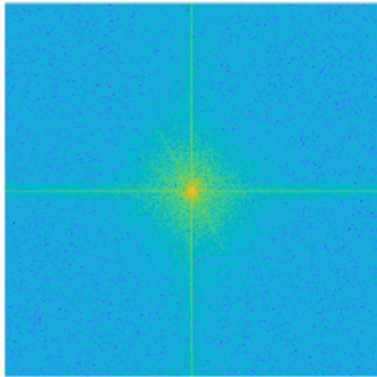
Power Spectrum of barbBig



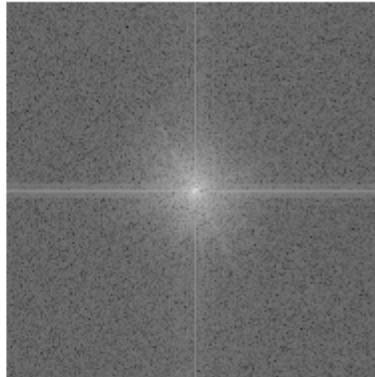
```
close all;
babyfft_1 = fftshift(abs(fft2(babyBig)).^2);
babyspectrum = log(babyfft_1 + 1);
figure, subplot(1,2,1),
imshow(babyspectrum / max(max(babyspectrum))),
colormap("default"),
title('Power Spectrum of babyBig w/ color');

subplot(1,2,2),
imshow(babyspectrum / max(max(babyspectrum))),
title('Power Spectrum of babyBig');
```


Power Spectrum of babyBig w/ color



Power Spectrum of babyBig



Part(a)

As can be seen from the above Power Spectrum the DC coefficient is located right at the center. Note that it would initially be in the corner but the `fftshift` brings it to the center. First `fft` transforms the signal to frequency domain, then the shift is basically rearranges the fourier transform by shifting the DC to center. Before the `fftshift`, the DC coefficient is located in the (1,1) position, and the low frequencies are out in the four corner areas. The `fftshift` swaps the quadrants, making the corner of each into the center such that the quadrants are symmetric at the center. The way we can check this is through the matlab documentation on how the `fftshift` function shifts the DC coefficient to be at the center of the 2D space, such that the spectrum looks symmetric.

Part(b)

By just doing the visual comparison, we can say that `barbBig` has a higher frequency content because of the vertical and horizontal lines that we see in the image and power spectrum. Looking at the `babyBig` and its power spectra, there are some high frequency components, however not as high as `barbBig`. We just see the white circle in the middle which is a low frequency component. `Barbara` will have more of the aliasing problem from downsampling because of the high frequency components in the image.

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- [Part\(d\)](#)

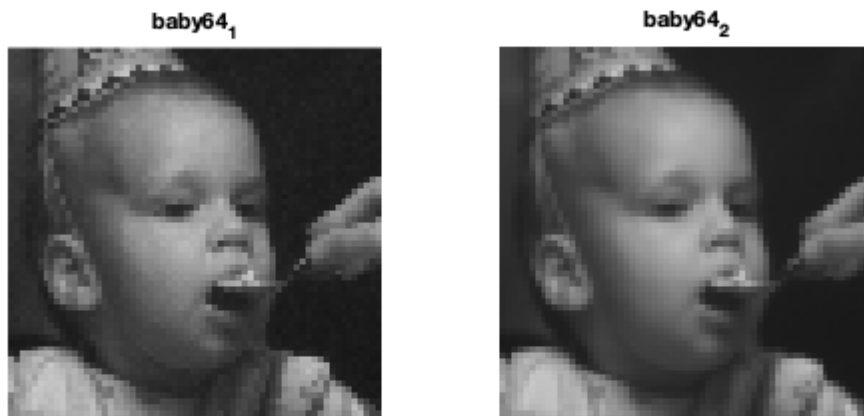
Part(c)

```
close all; clear; clc;
barbara = read(Tiff('barbara.tiff', 'r'));
baby = read(Tiff('baby.tiff', 'r'));

barbBig = barbara(1:512,37:548);
babyBig = baby(201:712,201:712);

baby64_1 = imresize(babyBig,0.125,'nearest');
figure,
subplot(1,2,1), imshow(baby64_1), title("baby64_1");

babylow = filter2(ones(8)/64,babyBig);
baby64_2 = imresize(babylow,0.125,'nearest');
subplot(1,2,2),
% imshow(baby64_2, [0 255]),
imshow(uint8(baby64_2)), title("baby64_2");
```



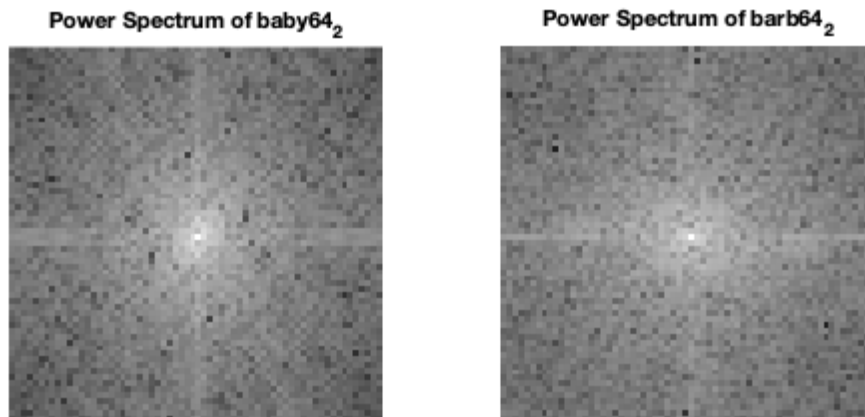
```
barb64_1 = imresize(barbBig,0.125,'nearest');
figure, subplot(1,2,1),
imshow(barb64_1), title("barb64_1")
```

```
barblow = filter2(ones(8)/64,barbBig);  
barb64_2 = imresize(barblow,0.125,'nearest');  
subplot(1,2,2),  
imshow(barb64_2, [0 255]), title("barb64_2");
```



```
baby64fft_1 = fftshift(abs(fft2(baby64_2)).^2);  
baby64spectrum = log(baby64fft_1 + 1);  
figure, subplot(1,2,1),  
imshow(baby64spectrum / max(max(baby64spectrum))),  
title('Power Spectrum of baby64_2');
```

```
barb64fft_1 = fftshift(abs(fft2(barb64_2)).^2);  
barb64spectrum = log(barb64fft_1 + 1);  
subplot(1,2,2),  
imshow(barb64spectrum / max(max(barb64spectrum))),  
title('Power Spectrum of barb64_2');
```



By visually comparing the two 64x64 versions of baby and for barbara, looks like the pre-filtering helped barbara more because of the higher frequency components and aliasing as observed before. barb64_1 looks bad, like around the scarf area where we have vertical lines in the original image with a lot of aliasing because it has high freq which we suppress using low freq filter and thus get a better image barb64_2 baby64_1 and baby64_2 looks identical, however, 1 is still little blurry compared to 2. similarity between them is likely due to the low freq nature of the image. It didn't have aliasing which required the pre-filtering, and thus it didn't make any difference.

Part(d)

Cutoff frequency $\leq 1/2 * \text{sampling frequency}$ sampling frequency = $1/8$ therefore, the cutoff freq is $1/16$

```
barbBig = barbara(1:512,37:548);
babyBig = baby(201:712,201:712);

barbfft_1 = fftshift(abs(fft2(barbBig)).^2);
barbspectrum = log(barbfft_1 + 1);

barbfft_1 = fftshift(abs(fft2(barbBig)).^2);
barbspectrum = log(barbfft_1 + 1);
figure,
I = barbspectrum / max(max(barbspectrum));
rec = insertShape(I,"filled-rectangle",[225 225 64 64], 'Color' , 'red');
imshow(rec)
title('Power Spectrum of barbBig');

% not Aliased Energy
ERM = sum(barbfft_1(225:288,225:288),'all');

% Aliasing Energy
```

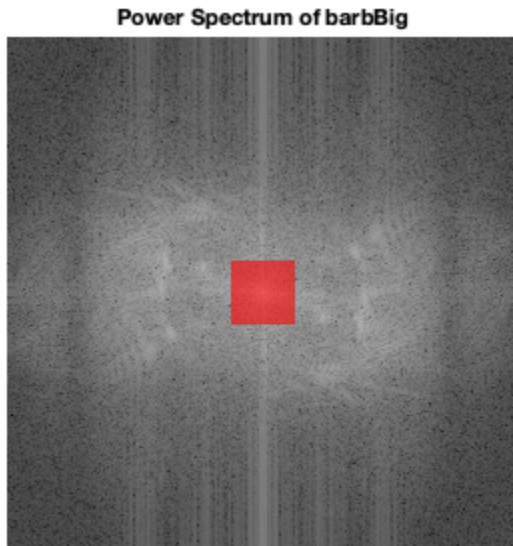
```

EAM = sum(barbfft_1,'all') - ERM;

fprintf('The not Aliased Energy is %i.\n', ERM);
fprintf('The Aliasing Energy %i.\n', EAM);

```

The not Aliased Energy is 1.022746e+15.
The Aliasing Energy 2.233135e+13.



```

% filtered!
barbfft_2 = fftshift(abs(fft2(barblow)).^2);
barbspectrum = log(barbfft_2 + 1);
figure,
I_64_2 = barbspectrum / max(max(barbspectrum));
rec = insertShape(I_64_2,"filled-rectangle",[225 225 64 64], 'Color' , 'red');
imshow(rec)
title('Power Spectrum of barb64 after filtering');

% not Aliased Energy
ER = sum(barbfft_2(225:288,225:288),'all');

% Aliasing Energy
EA = sum(barbfft_2,'all') - ER;

% Aliasing Reduction
aliasing_reduction = (EAM-EA)/EAM;

% Resolution Error
resolution_error = (ERM-ER)/ERM;

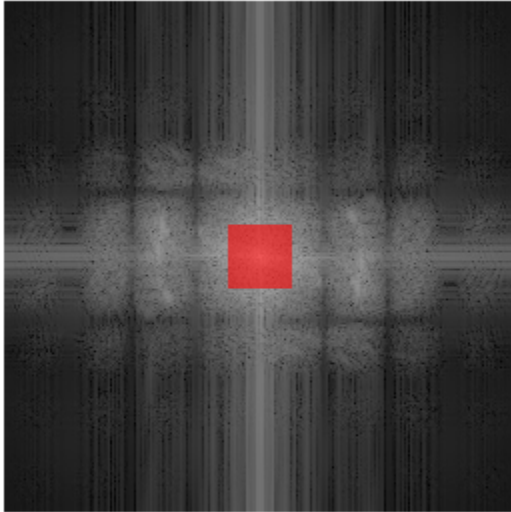
fprintf('The not Aliased Energy after filter is %i.\n', ER);
fprintf('The Aliasing Energy after filter %i.\n', EA);

```

```
fprintf('Aliasing Reduction %i.\n', aliasing_reduction);  
fprintf('Resolution Error %i.\n', resolution_error);
```

The not Aliased Energy after filter is 1.000850e+15.
The Aliasing Energy after filter 2.317592e+12.
Aliasing Reduction 8.962180e-01.
Resolution Error 2.140916e-02.

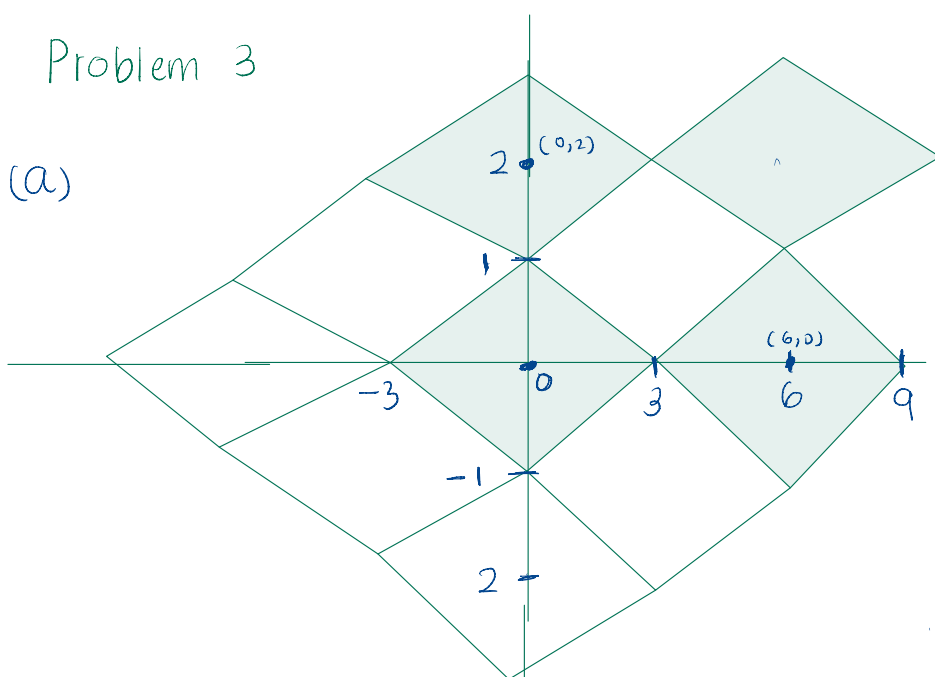
Power Spectrum of barb64 after filtering



% In the above image of the power spectra, the image where it is aliased and not-aliased is shown. The region inside the inner box with a red boarder is where the spectrum shows part which is not aliased. The region outside is aliased.

Problem 3

(a)



For the rectangular sampling lattice that has the lowest sampling density which can still ensure no aliasing (no overlap of spectral replications) has a generator matrix :

Therefore,

$$B = (A^{-1})^T$$

$$B^T = A^{-1}$$

$$(B^T)^{-1} = A$$

$$B = \begin{bmatrix} 6 & 0 \\ 0 & 2 \end{bmatrix}$$

$$B = B^T$$



each column is a spectral replication location

$$\det(A) = \frac{1}{12} \leftarrow$$

$$A = \frac{1}{12} \begin{bmatrix} 2 & 0 \\ 0 & 6 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1/6 & 0 \\ 0 & 1/2 \end{bmatrix}$$

← Generator matrix for lattice

(b) Generator Matrix for non-rectangular sampling lattice w/ lowest sampling density & Prevents spectral replications from overlapping.

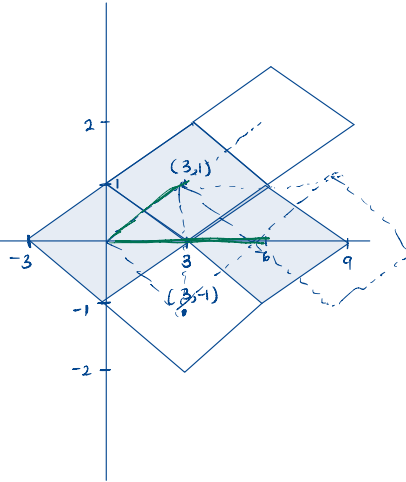
$$B^{-1} = \frac{1}{6-0} \begin{bmatrix} 1 & -3 \\ 0 & 6 \end{bmatrix}$$

$$B^{-1} = \begin{bmatrix} 1/6 & -1/2 \\ 0 & 1 \end{bmatrix}$$

$$(B^{-1})^T = A = \begin{bmatrix} 1/6 & 0 \\ -1/2 & 1 \end{bmatrix}$$

$$\det(A) = \frac{1}{6}$$

$$B = \begin{bmatrix} 6 & 3 \\ 0 & 1 \end{bmatrix}$$



Hexagonal Lattice

(c) Percentage reduction in samples by sampling on the non-rectangular grid as opposed to the rectangular grid.

$$\frac{\det(A)_{\text{rec}}}{\det(A)_{\text{NRec}}} = \frac{1/12}{1/6} = \frac{1}{2} = 0.50.$$

