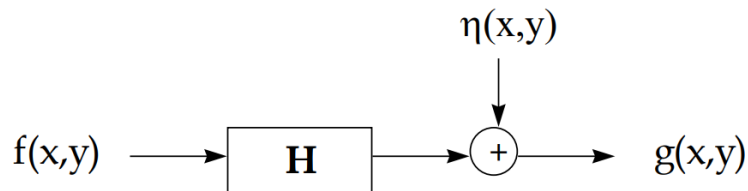


Report

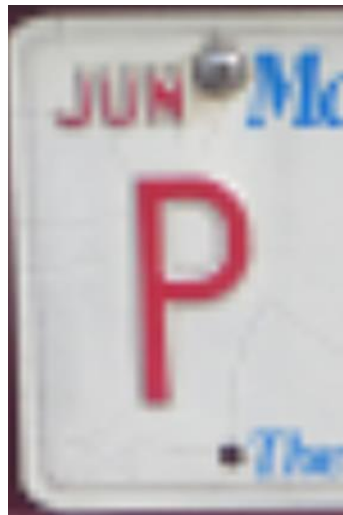
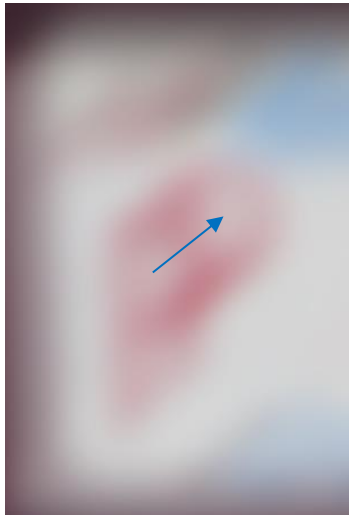
1. Blind image restoration

For the first task, we need to restore the input1.bmp without using the known clean image (input1_ori.bmp).

$$\text{Frequency Domain: } G(u, v) = H(u, v)F(u, v) + N(u, v)$$

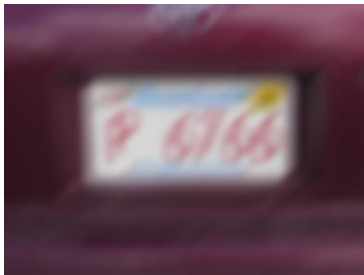


From the above, we are given only $g(x,y)$. We don't know the degradation function (H) and the noise $\eta(x,y)$ of the process. But we can observe the blurred image by human eyes to estimate the degradation process.



From the above images, we can guess that the blurring process is following the 45° direction and roughly counting the difference between two center points of P is about 20 pixels. Therefore, I use the motion blur kernel with 21 as the kernel size and 45° as the blurring direction.

On the other hand, I try the Gaussian blur kernel with different sigma values to make the deblurring effect more significant.



PSNR: 78.065 dB



PSNR: 64.747 dB

```

for(int i=0; i<Nb; i++) {
    for(int j=0; j<Nb; j++) {
        if(i == j)
            MotionBlurKernel[i][j] = 1.0 / double(Nb);
        else
            MotionBlurKernel[i][j] = 0;
    }
}

```

MotionBlurKernel with 45°

```

// generate Gaussian blur kernel
for(int x=-padding; x<padding+1; x++) {
    for(int y=-padding; y<padding+1; y++) {
        r = (x * x + y * y);
        GaussianBlurKernel[x+padding][y+padding] = (exp((-1) * r / dev)) / (M_PI * dev);
        sum += GaussianBlurKernel[x+padding][y+padding];
    }
}
// normalize kernel
for(int i=0; i<N; i++) {
    for(int j=0; j<N; j++) {
        GaussianBlurKernel[i][j] /= sum;
    }
}

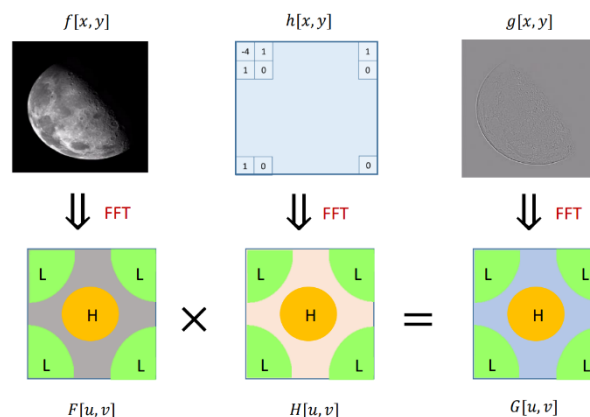
```

After preparing those kernels, we want to compute them in frequency domain.

Linear Filtering in Frequency Domain

We need to place the kernel values following the rule like this ->

Do 2D-FFT to get the frequency response of the kernel. The H from the formula in this task is combining the Gaussian blur kernel and the Motion blur kernel, which is simply multiplying in frequency domain. However, when we want



to restore the image to get \hat{F} , we cannot directly divide $G(u,v)$ by $H(u,v)$, this is because $H(u,v)$ may be close to zero at some point which is often happened at high frequency component. Hence, we need the K term to compensate the effect.

$$\hat{F}(u,v) = \left[\frac{1}{H(u,v)} \frac{|H(u,v)|^2}{|H(u,v)|^2 + K} \right] G(u,v)$$

```

// do fft
fft2d(Bplane, 1);
fft2d(Gplane, 1);
fft2d(Rplane, 1);

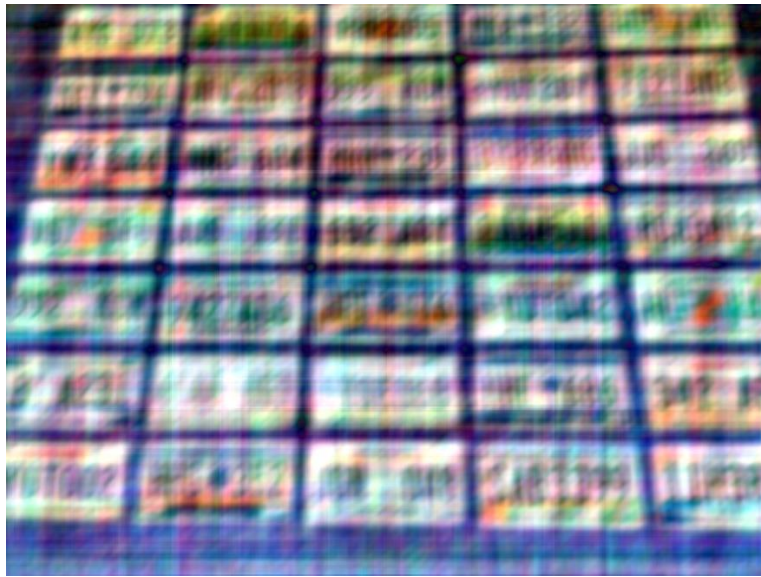
fft2d(GaussianBlurKernel_F, 1);
fft2d(MotionBlurKernel_F, 1);

for(int y=0; y<pad_height; ++y) {
    for(int x=0; x<pad_width; ++x) {
        complex<double> H = GaussianBlurKernel_F[y][x] * MotionBlurKernel_F[y][x];
        complex<double> wf = (1.0 / H) * (pow(abs(H), 2) / (pow(abs(H), 2) + eps));
        Bplane[y][x] = Gplane[y][x] * wf;
        Gplane[y][x] = Gplane[y][x] * wf;
        Rplane[y][x] = Rplane[y][x] * wf;
    }
}

// do ifft
fft2d(Bplane, -1);
fft2d(Gplane, -1);
fft2d(Rplane, -1);

```

As for deblurring using both input1.bmp and input1_ori.bmp to estimate the degradation function. I fail on making good estimation to restore the image.



Input2.bmp using blind method

| | | | | |
|---------|----------|----------|---------|---------|
| WYG 573 | 11FH756 | PHP 2455 | MKA 532 | 405 ZMU |
| MAY 794 | AFV 2018 | 993 KCM | YUT207 | 7121AN8 |
| YMX 644 | MMG 604 | MKM 239 | 378984K | JJS 269 |
| V67 SFL | JJS 131 | 552 AOY | 2AA4510 | RCA3412 |
| 992 KCM | 9427A06 | HPR 476 | YUT 042 | HLF V4 |
| 8 A231 | 4144 AGH | YSE068 | MHF 686 | 342 A8 |
| YUT002 | HHG352 | JGN 048 | SAB3399 | 11H38 |