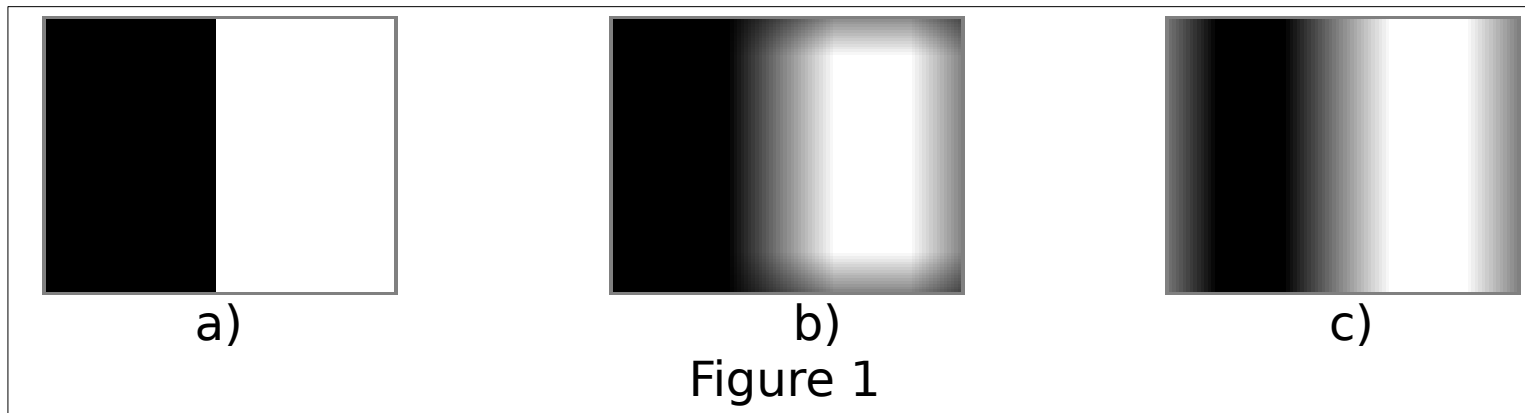


3. Exercise – Theory



A moving average filter was applied to the image in Figure 1a).

Figure 1b) shows the result, if the convolution is carried out in spatial domain, Figure 1c) if the convolution is carried out as multiplication in frequency domain.

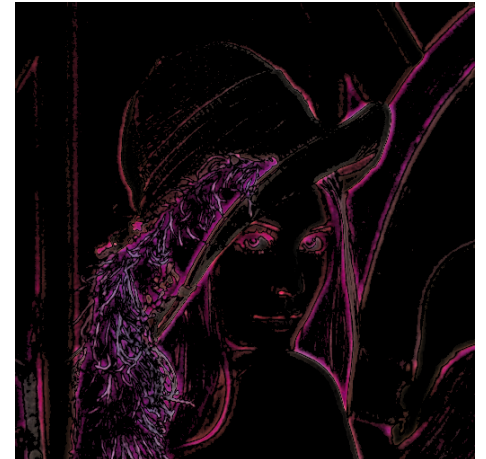
- i) Explain which assumptions lead to the “unexpected” border values in each image and why they are different for both methods. *Hint: Consider which are the grey values assumed outside the image boundaries.*
- ii) What steps are necessary for the convolution in spatial domain to produce the result in Fig. 1c)? *Hint: Consider different kinds of boundary treatment.*
- iii) What steps are necessary for the convolution by multiplication in frequency domain to produce the result in Fig. 1b)? *Hint: Consider to enlarge your image.*

3. Exercise – Given

FILE: main.cpp

```
int main(int argc, char** argv)
```

- Declares variables
- Displays and saves images
- Calls Dip3::run-function for unsharp masking, using either:
 - convolution in spatial domain (1)
 - convolution by multiplication in frequency domain (2)
- Measures and saves time to files
 - 1: convolutionSpatialDomain.txt
 - 2: convolutionFrequencyDomain.txt



3. Exercise – Given

```
Mat Dip3::mySmooth(Mat& in, int size, int type)
```

in: input image

size: size of filter kernel

type: whether or not working in spatial domain

return: smoothed image

- Applies Gaussian blurring to image
- Either working in spatial or frequency domain

```
void Dip3::test(void)
```

```
void Dip3::test_createGaussianKernel(void)
```

```
void Dip3::test_circShift(void)
```

```
void Dip3::test_frequencyConvolution(void)
```

→ Simple test function to check for basic correctness

3. Exercise – Reuse!

```
Mat Dip3::spatialConvolution(Mat& src, Mat& kernel)
```

- Parameter:
 - `src` : source image
 - `kernel` : kernel of the convolution
 - `return` : output image
- Applies convolution in spatial domain
- One method of border handling: `size(src) == size(return)`
- Do **NOT** use convolution functions of OpenCV

3. Exercise – To Do

Mat Dip3::createGaussianKernel(int kSize)

kSize: the kernel size

return: the generated filter kernel

```
Mat Dip3::createGaussianKernel(int kSize){

    Mat kernel = Mat::zeros(kSize, kSize, CV_32FC1);

    // some variables for gaussian filter kernel
    float mu_x = kernel.cols/2;    // x-coordinate of center
    float sigma_x = kernel.cols/5; // standard deviation in x direction
    float mu_y = kernel.rows/2;    // y-coordinate of center
    float sigma_y = kernel.rows/5; // standard deviation in y direction

    float val=0, norm=0;

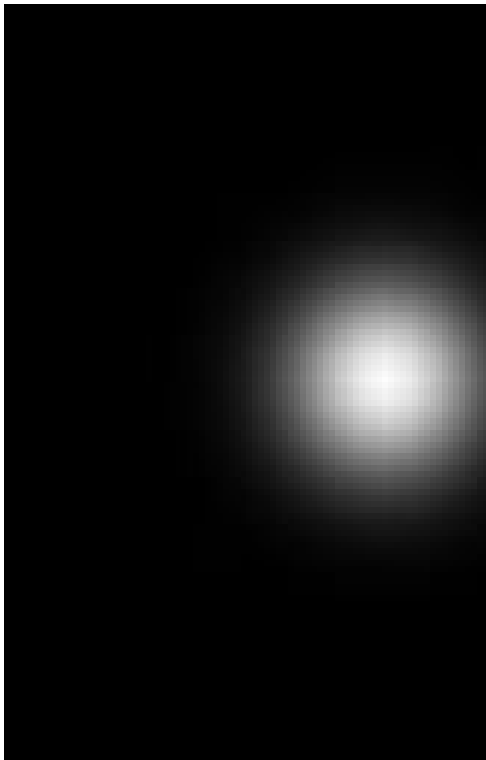
    // calculate corresponding kernel value at each position
    for(float x=0; x<kernel.cols; x++){
        for(float y=0; y<kernel.rows; y++){
            val = exp( -0.5*(((x-mu_x)/sigma_x)*((x-mu_x)/sigma_x) + ((y-mu_y)/sigma_y)*((y-mu_y)/sigma_y)) );
            norm += val;
            kernel.at<float>(y,x) = val;
        }
    }
    // ensure integration to 1
    for(float x=0; x<kernel.cols; x++){
        for(float y=0; y<kernel.rows; y++){
            kernel.at<float>(y, x) /= norm;
        }
    }
    cout << kernel << endl;
    return kernel;
}
```

3. Exercise – To Do

Mat Dip3::circShift(Mat& in, int dx, int dy)

- Performes circular shift in (dx,dy) direction

```
Mat Dip3::circShift(Mat& in, int dx, int dy){  
  
    Mat out = in.clone();  
  
    int x, y, new_x, new_y;  
  
    for(y=0; y<in.rows; y++){  
  
        // calculate new y-coordinate  
        new_y = y + dy;  
        if (new_y<0)  
            new_y = new_y + in.rows;  
        if (new_y>=in.rows)  
            new_y = new_y - in.rows;  
  
        for(x=0; x<in.cols; x++){  
  
            // calculate new x-coordinate  
            new_x = x + dx;  
            if (new_x<0)  
                new_x = new_x + in.cols;  
            if (new_x>=in.cols)  
                new_x = new_x - in.cols;  
  
            out.at<float>(new_y, new_x) = in.at<float>(y, x);  
  
        }  
    }  
    return out;  
}
```



3. Exercise – To Do

Mat Dip3::frequencyConvolution(Mat& in, Mat& kernel)

- Calculates convolution
- Forward transform:
dft(Mat, Mat, 0);
- Inverse transform
dft(Mat, Mat, DFT_INVERSE);
- Spectrum multiplication
mulSpectrums(Mat, Mat, 0);

Re Y_{00}	Re Y_{00}
Re Y_{10}	Re Y_{10}
Re Y_{20}	Re Y_{20}
...	...

```
Mat Dip3::frequencyConvolution(Mat& in, Mat& kernel){
    Mat out = Mat::zeros( in.rows, in.cols, CV_32FC1);

    // create new matrices of optimal size
    Mat dft_in = in.clone();
    Mat dft_kernel = Mat::zeros( in.rows, in.cols, CV_32FC1);
    // DFT
    dft( dft_in, dft_in, CV_DXT_FORWARD, in.rows);

    // copy kernel into new matrix
    Mat roi = dft_kernel(Rect(0,0,kernel.cols,kernel.rows));
    kernel.copyTo(roi);
    // center filterkernel
    dft_kernel = circShift( dft_kernel, -kernel.cols/2, -kernel.rows/2);

    // DFT
    dft( dft_kernel, dft_kernel, CV_DXT_FORWARD );

    //multiplication of fouriertransformed image and kernel
    mulSpectrums( dft_in, dft_kernel, dft_in, 0 );

    // inverse dft
    dft( dft_in, dft_in, CV_DXT_INV_SCALE, out.rows );

    // copy into output image
    dft_in.copyTo(out);

    return out;
}
```

3. Exercise – To Do

```
Mat Dip3::run(Mat& in, int type, int size, double thresh, double scale)
Mat Dip3::usm(Mat& in, int type, int size, double thresh, double scale)
```

in: input image
type: smoothing i
by multiplic
size: kernel size k
thresh: threshold T
scale: scale s
return: result

- Unsharp masking
- Thresholding to
- Useful function

```
// subtract smoothed version from original
tmp = in - tmp;

// threshold "edge" image
Mat tmp1, tmp2;
threshold(tmp, tmp1, thresh, 0, THRESH_TOZERO);
threshold(-tmp, tmp2, thresh, 0, THRESH_TOZERO);
tmp = tmp1 - tmp2;

// scale edge enhancement
tmp *= scale;

// add "edge enhancement" image
Mat out = in + tmp;

// constrain image values to be in [0,255]
threshold(out, out, 0, 0, THRESH_TOZERO);
threshold(out, out, 255, 255, THRESH_TRUNC);

return out;
```