## Introducing Assignment 1: Image Processing

## COS 426: Computer Graphics (Fall 2022)

## Setup

## Same layout as A0:

- Run "python3 -m http.server" (or similar) inside the assignment directory
- Open "http : //localhost : 8000 " in web browser


## GUI

cos426 Assignment 1
Image Processing - Interactive Mode
Switch to: Writeup
Student Name <NetID>


| Push Image | $\checkmark$ History |  |  |
| :---: | :---: | :---: | :---: |
| Batch Mode | - 1: Push Image |  |  |
| Animation | image name | flower.jpg | $\bigcirc$ |
| MorphLines | Delete Below |  |  |
| - SetPixels | - 2: Brightness |  |  |
| - Luminance | brightness |  | 0 |
| Brightness | Delete |  |  |
| Contrast | Close Controls |  |  |

- Useful functions
- Push Image
- Animation: generate gif animation using (min, step, max)
- MorphLines: specify line correspondences for morphing
- BatchMode: fix current parameter settings
- Features to implement
- SetPixels: set pixels to certain colors (This was A0)
- Luminance: change pixel luminance
- Color: remap pixel colors
- Filter: convolution/box filter
- Dithering: reduce visual artifacts due to quantization $\approx$ cheat our eyes
- Resampling: interpolate pixel colors
- Composite: blending two images
- Misc


## Features

## Luminance

- Brightness
- Contrast
- Gamma
- Vignette
- Histogram equalization

Color

- Grayscale
- Saturation
- White balance
- Histogram matching

Filter

- Gaussian
- Sharpen
- Edge detect
- Median
- Bilateral filter

Dithering

- Quantization
- Random dithering
- Floyd-Steinberg error diffusion
- Ordered dithering

Resampling

- Bilinear sampling
- Gaussian sampling
- Translate
- Scale
- Rotate
- Swirl


## Composite

- Composite
- Morph

A few reminders...

- Don't try to exactly replicate example images.
- Choose parameters in your code which give you best looking results.
- Have fun!


## Changing Contrast

- GIMP formula
- value $=($ value -0.5$) *(\tan (($ contrast +1$) * P l / 4))+0.5 ;$
- "Difference above mid-value times contrast multiplier, plus mid-value"
- When contrast=1, $\tan (\mathrm{PI} / 2)$ is infinite, think about limit and what is reasonable
- Clamp pixel to [0, 1] after computing the value.
- Apply to each channel separately.



## Gamma correction

- $R=R^{\wedge}$ gamma, $G=G^{\wedge}$ gamma, $B=B^{\wedge}$ gamma
- $R, G, B$ are typically in $[0,1]$ (default in the code base)
- Second arg of gammaFilter(image, logOfGamma) is log(gamma)
- So use gamma = Math.exp(logOfGamma)
- Exponentiation in JS is "Math.pow(base, exponent)" or (ES7 / ES2017+) "base**pow"
- Your browser might not support ES7



## Vignette

- Pixels within inner radius remain unchanged
- Pixels outside outer radius are black
- Pixels between innerR and outerR should be multiplied with a value in [0, 1]:
$-R=\operatorname{sqrt}\left(x^{\wedge} 2+y^{\wedge} 2\right) /$ halfdiag
- Multiplier = 1 - (R - innerR) / (outerR - innerR)
- Similar to soft brush


Multiplier map

## Histogram Equalization

Transform an image so that it has flat histogram of luminance values.


Before


## Histogram Matching

## Transform an image so that it has same histogram of luminance values as reference image.


reference image: town

reference image: flower

## Histogram Equalization/Matching




## Histogram Equalization/Matching

- Image: x
- Number of gray levels: L
- $p d f(i)=\frac{n_{i}}{n} \quad n_{i}=$ number of pixels of the i-th gray level
- $c d f(j)=\sum_{j=0}^{i} p d f(i)$
- Target cdf:
- Equalization:
- $c d f_{r e f}(i)=\frac{i}{L-1}$
- Matching:
- cdf of the reference image

(source:http://paulbourke.net/miscellaneous/equalisation/)


## Histogram Equalization/Matching

- Target cdf:
- Equalization:
- $c d f_{r e f}(i)=\frac{i}{L-1}$
- Matching:
- cdf of the reference image
- Implementation
- Equalization
- $x^{\prime}=(c d f(x) *(L-1)) /(L-1)$
- Matching
- $x^{\prime}=\arg \min _{i}\left|c d f(x)-c d f_{r e f}(i)\right|$
- Convert back to gray level: $x^{\prime}=\frac{x^{\prime}}{L-1}$



## Saturation

- pixel = pixel + (pixel - gray(pixel)) * ratio
- Do clamp()



## White balance

whitebalance(image, $r g b_{w}$ )
$\left[L_{w}, M_{w}, S_{w}\right.$ ] = rgb2lms $\left(r g b_{w}\right)$
for each pixel x in image
$[L, M, S]=$ rgb2lms(image $(x)$ )
$\mathrm{L}=\mathrm{L} / L_{w}$
$\mathrm{M}=\mathrm{M} / M_{w}$
$\mathrm{S}=\mathrm{S} / S_{w}$
image_out(x) $=\operatorname{lms} 2 \operatorname{rgb}(\mathrm{~L}, \mathrm{M}, \mathrm{S})$

- Hints:
- Use rgbToXyz(), xyzToLms(), ImsToXyz(), xyzToRgb()
- Do clamp()


## Convolution (Gaussian/Sharpen/Edge)



## Convolution (Gaussian/Sharpen/Edge)

- Weights can be normalized depending on the application
- Variety of ways to handle edges
- Mirror boundary
- Zero padding
- Use part of the kernel only


## Gaussian filter

- Create a new image to work on
- Weights should be normalized to sum to 1 , otherwise average color changes

$$
G(x)=\frac{1}{\sqrt{2 \pi \sigma^{2}}} e^{-\frac{x^{2}}{2 \sigma^{2}}} \quad \frac{1}{16}\left[\begin{array}{lll}
1 & 2 & 1 \\
2 & 4 & 2 \\
1 & 2 & 1
\end{array}\right]
$$

- $x=$ distance to the center of the kernel
- Linear separation optimization:
- First apply a 1D Gaussian kernel vertically and then a 1D Gaussian kernel horizontally


## Edge

Kernel:

| -1 | -1 | -1 |  |  |
| :--- | ---: | :--- | ---: | ---: |
| -1 | 8 | -1 |  |  |
| -1 | -1 | -1 | 3 | -1 |
| Inside boundary | -1 | -1 |  |  |

- Weights sum to 0
- Optional to invert the edge map for visualization: pixel = 1 - pixel


## Sharpen

- Kernel:

| -1 | -1 | -1 |  |  |
| :--- | :--- | :--- | ---: | :--- |
| -1 | 9 | -1 | 4 | -1 |
| -1 | -1 | -1 | -1 | -1 |
| Inside boundary |  |  |  |  |
| ghts sum to 1 |  |  |  |  |
| At boundary |  |  |  |  |

Edge Filter vs Sharpen Filter

| -1 | -1 | -1 | -1 | -1 | -1 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| -1 | 8 | -1 | -1 | 9 | -1 |
| -1 | -1 | -1 | -1 | -1 | -1 |
|  | Edge Filter |  |  |  |  |
| Convolution(Image, Sharpen Filter) $=$ Convolution(Image, Edge Filter) + Image |  |  |  |  |  |

## Median

- Use a window (similar to convolution)
- Choose the median within the window
- Sorting: sort by RGB separately / sort by luminance
- Optimization: use quick-select to find median
- Gives median in linear time


RGB Example

## Bilateral

- Combine Gaussian filtering in both spatial domain and color domain
- Weight formula of filter for pixel (i, j): Spatial distance component Color distance component

- Similar color -> large weights, Different color -> smaller weights



## Sampling \& Frequencies

- Real-world is continuous, Sensors are discrete
- How many samples do we need to measure real world?
- Too few samples = aliasing
- Nyquist rate says that we need to sample at $\geq 2 \times$ the highest frequency for perfect reconstruction
- Aliasing is when signal $X$ masquerades as signal $Y$
$-Y$ is the alias of $X$



## Fourier Transform



1st Harmonic


2nd Harmonic


Nyquist Frequency


Maps signal from time domain to frequency domain
Use low-pass filter to remove high frequencies and prevent aliasing




## Fourier Transform



Maps signal from time domain to frequency domain
Use low-pass filter to remove high frequencies and prevent aliasing


## 1D to 2D

- 2D signals follow the same analysis as 1D signals

Real world 2D image is sampled by sensor
Aliasing for 2D signals


Inadequate sampling
(Barely) adequate sampling

## 1D to 2D

- 2D signals follow the same analysis as 1D signals

Fourier Analysis for 2D signals

Spatial domain


C


Frequency domain


If image resolution is low

- E.g. image compression

Then need to apply band-limiting filter to avoid aliasing

- E.g. Triangle, Gaussian

Note that these filters are "finite" filters, they act as approximations to a perfect low pass filter

## Resampling

- Gaussian interpolation
- Weights:

$$
G(d, \sigma)=e^{-d^{2} /\left(2 \sigma^{2}\right)}
$$

- Weights need to be normalized, so that sum up to 1
- Use windowSize = 3*sigma - Sigma can be 1
- Window can be square



## Resampling

- Bilinear interpolation

$$
\begin{aligned}
f(x, y)=\frac{1}{\left(x_{2}-x_{1}\right)\left(y_{2}-y_{1}\right)} & \left(f\left(Q_{11}\right)\left(x_{2}-x\right)\left(y_{2}-y\right)+f\left(Q_{21}\right)\left(x-x_{1}\right)\left(y_{2}-y\right)\right. \\
& \left.+f\left(Q_{12}\right)\left(x_{2}-x\right)\left(y-y_{1}\right)+f\left(Q_{22}\right)\left(x-x_{1}\right)\left(y-y_{1}\right)\right)
\end{aligned}
$$

(from wikipedia)


Quantization

- Quantize a pixel within $[0,1]$ using $n$ bits - round $\left(p^{*}\left(2^{\wedge} n-1\right)\right) /\left(2^{\wedge} n-1\right)$

$\mathrm{n}=1$ example


## Random dithering

- Before quantization:
- $p=p+($ random() -0.5$) /\left(2^{\wedge} n-1\right)$
- n is number of bits per channel

Reduce banding with intentional noise


## Ordered dithering

## Pseudo code for $n$-bit case:

i $=x \bmod m$
$j=y \bmod m$
err = I(x, y) - floor_quantize(I(x, y)))

$$
\mathbf{m}=4, \mathrm{D}=\left[\begin{array}{cccc}
15 & 7 & 13 & 5 \\
3 & 11 & 1 & 9 \\
12 & 4 & 14 & 6 \\
0 & 8 & 2 & 10
\end{array}\right]
$$

threshold $=(\mathrm{D}(\mathrm{i}, \mathrm{j})+1) /\left(\mathrm{m}^{\wedge} 2+1\right)$
if err > threshold

$$
P(x, y)=\text { ceil_quantize(I }(x, y)))
$$

else

$$
P(x, y)=\text { floor_quantize(I }(x, y)))
$$

- floor_quantize(p)

$$
=\mathrm{floor}\left(\mathrm{p} *\left(2^{\wedge} \mathrm{n}-1\right)\right) /\left(2^{\wedge} \mathrm{n}-1\right)
$$

- ceil_quantize(p)

$$
=\operatorname{ceil}\left(p *\left(2^{\wedge} n-1\right)\right) /\left(2^{\wedge} n-1\right)
$$


$\mathrm{n}=1$ example

## Floyd-Steinberg error diffusion

- Loop over pixels line by line
- Quantize pixel
- Compute quantization error (the difference of the original pixel and the quantized pixel)
- Spread quantization error over four unseen neighboring pixels with weights (see left figure below)
- Results look more natural


Q\&A

## Transformation (translate/scale/rotate/swirl)

- Inverse mapping

input


Inverse mapping guarantees that every pixel in the transformed image is filled!

## Transformation (translate/scale/rotate/swirl)

- To fill in a pixel in the target image, apply the inverse transform to the pixel location and look it up in the input image (with resampling technique) for pixel value.
- i.e. For translation of $x^{\prime}=x+t x, y^{\prime}=y+t y:$

$$
I^{\prime}\left(x^{\prime}, y^{\prime}\right)=I\left(x^{\prime}-t x, y^{\prime}-t y\right)
$$

- i.e. For scale of $x^{\prime}=x^{*} s x, y^{\prime}=y^{*}$ sy:
$l^{\prime}\left(x^{\prime}, y^{\prime}\right)=I\left(x^{\prime} / s x, y^{\prime} / s y\right)$


## Composite

- output = alpha * foreground + (1-alpha) * background
- alpha is the alpha channel foreground

backgroundImg

foregroundImg

foregroundImg(alpha channel)


Result

## Morph

- Basic concepts
- transform the background image to the foreground image
- alpha $=0$ : show background
- alpha = 1: show foreground
- alpha is the blending factor / timestamp
- General approach
- specify correspondences (morphLines.html)
- create an intermediate image with interpolated correspondences (alpha)
- warp the background image to the intermediate image
- warp the foreground image to the intermediate image
- blend using alpha


## Morph

GenerateAnimation(Image $_{0}, L_{0}[\ldots]$, Image $\left._{1}, L_{1}[\ldots]\right)$ begin
foreach intermediate frame time t do for $\mathrm{i}=0$ to number of line pairs do
$L[i]=$ line $t$-th of the way from $L_{0}[i]$ to $L_{1}[i]$ end
Warp $_{0}=$ WarpImage $^{\left(\text {Image }_{0}, L_{0}, L\right)}$
Warp $_{1}=$ WarpImage $^{\left(\text {Image }_{1}, L_{1}, L\right)}$ foreach pixel $p$ in Finallmage do

Result $(\mathrm{p})=(1-\mathrm{t}) \mathrm{Warp}_{0}+\mathrm{t} \mathrm{Warp}_{1}$ end
end
end

## Warp Image

- $u=\frac{(X-P) \cdot(Q-P)}{\|Q-P\|^{2}}$
- $v=\frac{(X-P) \cdot \operatorname{Perpendicular}(Q-P)}{}$ unit vector $\quad \operatorname{Perpendicular}(Q-P)=(y,-x)$
- $X^{\prime}=P^{\prime}+u \cdot\left(Q^{\prime}-P^{\prime}\right)+\frac{v \cdot \text { Perpendicular }\left(Q^{\prime}-P^{\prime}\right)}{\left\|Q^{\prime}-P^{\prime}\right\|}$ unit vector
- dist $=$ shortest distance from $X$ to $P Q$
- 0 <= $u$ <= 1: dist = |v|
- $u<0$ : dist $=\| X-P| |$
- $u>1$ : dist $=||X-Q||$
- weight $=\left(\frac{\text { length }^{p}}{a+\text { dist }}\right)^{b}$
- we use $p=0.5, a=0.01, b=2$


Contribution of line segment PQ to the warping of X's location

## Warp Image



Warped background or Pixel source (background or foreground) foreground (currently black)

## Warp Image

For each pixel $\boldsymbol{X}$ in the destination
$\boldsymbol{D S U} \boldsymbol{M}=(0,0)$
weightsum $=0$
For each line $\boldsymbol{P}_{\boldsymbol{i}} \boldsymbol{Q}_{\boldsymbol{i}}$
calculate $u, v$ based on $P_{i} Q_{i}$

calculate $X_{i}^{\prime}$ based on $\boldsymbol{u}, \boldsymbol{v}$ and $P_{i}{ }^{\prime} Q_{i}{ }^{\prime}$
calculate displacement $D_{i}=\boldsymbol{X}_{\boldsymbol{i}}{ }^{\prime}-\boldsymbol{X}_{\boldsymbol{i}}$ for this line dist $=$ shortest distance from $X$ to $P_{i} Q_{i}$
weight $=\left(\text { length }^{p} /(a+\text { dist })\right)^{b}$
DSUM $+=D_{i}$ * weight
weightsum $+=$ weight
$X^{\prime}=X+$ DSUM / weightsum
destinationImage $(\boldsymbol{X})=$ sourceImage $\left(\boldsymbol{X}^{\prime}\right)$

## Interpolate Morph Lines



Background Image


Foreground Image
current_line[i] = (1 - alpha) * background_lines[i] + alpha * foreground_lines[i]

## Blending



## Blending



Background Image
alpha $=0.5$ (also the blending factor)


Foreground Image

