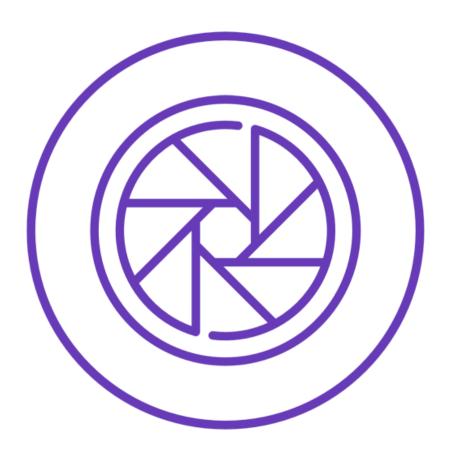
PicPro



CAB432 Assignment 2

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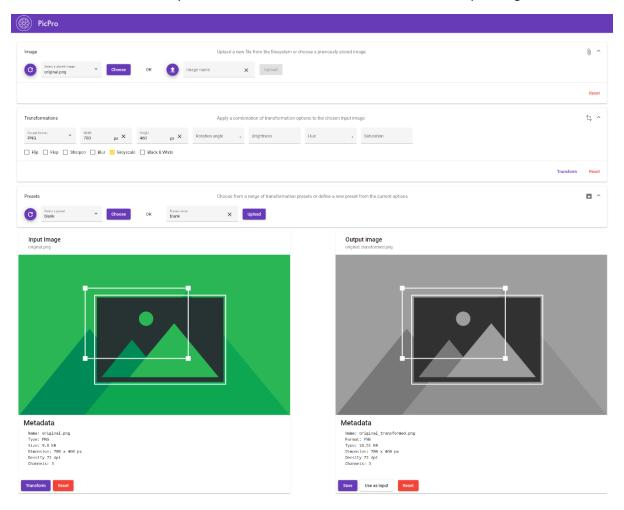
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Introduction

Purpose & description

The PicPro Image Transformer allows users to alter images using various predefined options, such as changing the filetype format, resizing the image, or adding effects like blur and greyscale. It allows new and transformed images to be stored for users that wish to use the same image multiple times, along with saving presets for transformations so that users don't need to repeatedly fill in the transformation values if they wish to use the same transformation over multiple images.



Services used

Sharp Node.js module (v0.31.1)

The Sharp module accepts and reads an image file input to allow for various transformations to be done, such as filetype format, resizing the image, or adding effects like blur and greyscale etc.

Endpoint: N/A as it is not an API, but a Node.js module

Docs: https://sharp.pixelplumbing.com/

Use cases

US 1

As a	graphic designer
I want	to transform images like changing the format type or adding filters
So that	I can manipulate images to suit my design.

US 2

As a	graphic designer
I want	save previously uploaded images
So that	I don't need to reupload the same image and can access it from the service with
	ease.

US 3

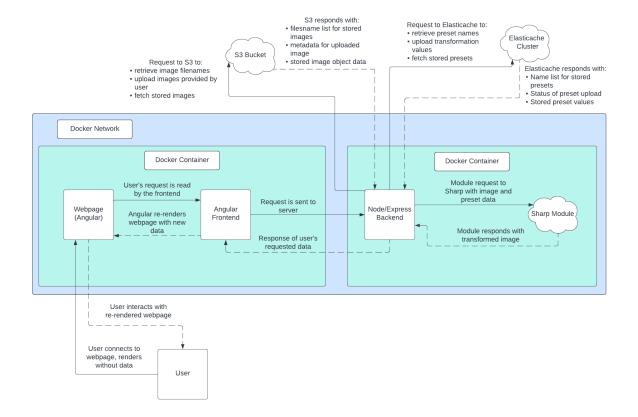
As a	graphic designer	
I want	save previous transformation presets/values	
So that	So that I can apply the same transformation to multiple images without having to fill in	
	the values every time	

Technical breakdown

Architecture

The flow of data can be seen in the context diagram below. The application consists of two docker containers, one containing the frontend, and one containing the backend. The frontend is a single page web application written with Angular, that processes and sends the user's input data to the backend. The backend is an Express application based on Node.js, receiving the input data and filtering the desired attributes that will be used for either storing/fetching images via S3, storing/fetching presets via ElastiCache, or transforming images via Sharp. Once the response is received, the backend sends the output back to the frontend to be displayed to the user.

Context Diagram



These containers are built separately with *docker compose* with the Docker file content shown in linked Appendix. The <u>backend Docker file</u> sets the work directory to the backend folder, copies the package.json file from the node folder so that it can install the necessary dependencies using npm install. Afterwards, it copies the rest of the code in the container to run the complied *server.js* file to start the backend. This gets exposed to port 3000. The <u>Dockerfile</u> of the client uses a multi-stage build. The first stage builds the Angular application and copies the artifacts. The second stage serves the app on a lightweight NGINX web server, which reduces the size of the Docker image heavily to around 25 MB. The original size, if serving the app on the development server, would have been over 2 GB. This setup follows best-practice guidelines for serving Angular applications in a production environment, and the server also acts as a reverse proxy to redirect requests between client and server in any environment. This gets exposed to port 80. These two Docker files are then used by the <u>docker compose file</u> to be built and tagged together, exposing them to the appropriate ports, 3000:3000 for the backend and 4200:80 for the frontend. Port 80 is needed so that the NGINX reverse proxy can listen on a separate port to redirect requests accordingly.

The way the application is deployed to AWS by pulling the images for frontend and backend from DockerHub, creating a Docker Network so that they can be accessed by container name and then running containers with the --restart unless-stopped flag set so that the containers are run on machine startup. This is needed for the dynamic creation of machines during scaling. The pm2 configuration file (ecosystem.config.js) is available in Appendix 4A. but it was not used in the end because pure Docker can achieve the same thing.

Client / server demarcation of responsibilities

The best way to describe the demarcation of responsibilities for the application is to again separate it into **frontend and backend**.

The **frontend** oversees the GUI (Graphical User Interface). What the users see on the webpage, from the header, to the three sections representing the Image, Transformation, and Preset features, along with the input/output images, are rendered from the frontend via Angular. The frontend also reads any input from the user so that it can be formatted inside a JSON request that gets sent back to the frontend, unpacking and re-rendering new output when the backend responds.

The **backend** handles the data processing. When the frontend makes a request, it sends it to the appropriate route, which can be categorised as the image and preset routes. The Appendix code has been linked below.

The **image routes** handle the requests from the Image and Transformation sections of the webpage, and there are 4 of them. The base route <u>"images/"</u> retrieves the filenames of stored S3 images so users can choose one to fetch. The <u>"iamges/fetch"</u> and <u>"images/upload"</u> routes handle storing and retrieving images from the S3 bucket, fetch responding back the chosen image information that will be displayed on the webpage while upload only responds with the metadata. The <u>"iamges/transform"</u> value again retrieves the desired image, using the transformation values sent from the frontend to apply the transformations via Sharp, before creating and sending the response containing the information for the newly edited image.

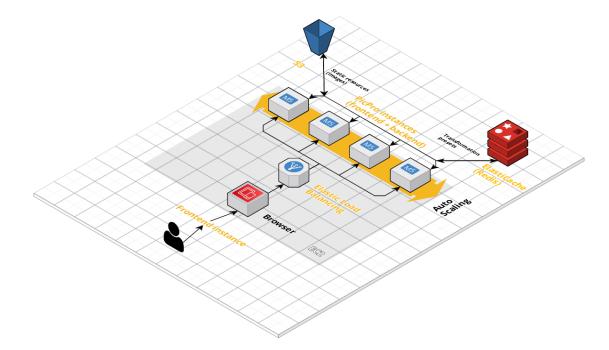
The **preset routes** handle the requests from the Preset section of the webpage, and there are 3 of them. These are similar to the image routes, and all of them interact with Redis via an ElastiCache Cluster: <u>"presets/"</u> responds with the preset names that users can choose from, <u>"presets/fetch"</u> retrieves the chosen preset that the frontend can use to fill in the Transformation section, and

<u>"presets/upload"</u> stores the current Transformation values along with a preset name provided by the user.

Looking at the architecture diagram below to explain the demarcation of responsibilities done via AWS, most of the application is contained within EC2. As can be seen, the user interacts with the frontend instance using the webpage, and this interacts with the backend via an Elastic Load Balancer. This scales the instances up and down depending on the computational load the application is receiving.

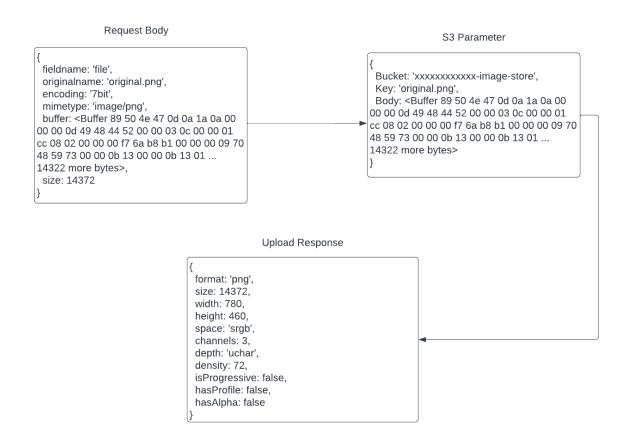
As shown in the diagram, the backend instances interact with the S3 bucket and the Elasticache Cluster when needed. S3 is a blob store, often referred to as object storage. It was chosen to store the images uploaded from the application due to how large and flexible it is, as users may upload several original and transformed images in quick succession. S3 can scale alongside the data stored in it, making it very suitable for large amounts of images.

ElastiCache for Redis is a memory-based cache and was chosen to store the presets that contain the transformation values due to its speed. As these presets are stored as JSON objects, they do not require the large flexibility in storage scaling that S3 boasts, and need to be accessed quickly, which ElastiCache provides very fast access time with very little latency, which is suitable for quickly retrieving the preset values that users will use to make their transformations.



Response filtering / data object correlation

The below diagram shows the data manipulation of the image <u>upload</u> route. The original request contains the image data, like the original name and the image buffer.



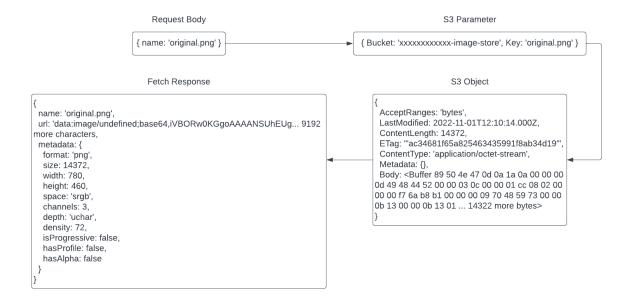
Buffer is used due to many file readers like the Sharp module often manipulating files as Buffer data, as they are suitable for fine-grained data manipulation, according to Digital Ocean (2020, para. 3), along with its ease of conversion to other data types like a base64 URL, used to re-render the images to the webpage so they can be displayed back to the user.

The upload parameters contain the bucket name along with the filename from the request body and the image buffer from the image data. This gets uploaded to S3, while the response contains the metadata that is obtained using Sharp, as shown in the code segment below.

```
sharp(file.buffer)
    .metadata()
    .then(metadataResult => {
```

<u>Fetch</u> can be seen as going backwards from upload path. The original request is the filename of the image the user chose to retrieve. This filename is placed within a parameter JSON for S3, again containing the bucket name, which is used to retrieve the data. The object that is returned holds various information, like the last record of it being modified, and the body which contains the image buffer.

As metadata is not stored in S3 and is therefore blank, the route again uses Sharp to obtain it, which gets placed in the fetch response, also containing the filename and the image buffer which has been converted to a base64 URL.



<u>Transform</u> has a very similar path to fetch. Using the image name to obtain the desired image from S3, the request body containing the transformation values are sent to the transform function, that calls Sharp to apply the transformations using several if-statements to check through the request and apply the necessary transformations when values are supplied or marked as true.

The function first creates the Sharp pipeline by reading the image buffer and resizing based on the supplied width and height. If they haven't been changed from the original image, then it just stays the same size. Every if statement afterwards stacks a prebuilt sharp function to the pipeline to apply the transformation, below is only a short example of the full function. The full example is shown in Appendix 12A and 13A. At the end, it returns the pipeline containing all the transformations converted back into an image buffer.

```
function transform(imgBuffer, transformation) {
                               let transformationPipeline =
                           sharp(imgBuffer).resize(transformation.width, transformation.height);
                                // greyscale
                               if (transformation.greyscale) {
                                    transformationPipeline.greyscale();
    Request Body
                                // brightness
outputType: 'PNG',
                               if (transformation.brightness) {
flip: false,
                                    transformationPipeline.modulate({
flop: false
sharpen: false,
                                        brightness: transformation.brightness,
blur: false,
greyscale: false,
blackwhite: false.
imageName: 'original.png',
width: 780,
height: 460
                               return transformationPipeline.toBuffer();
```

The transform function returns the new image buffer for the transformed image. As seen in the code below, the original filename is split at the full stop so that '_transformed' can be added to differentiate the image from the original. This is only a default naming and when the user decides to download or save the image, they will be queried to supply a custom file name. The response is the same as previously shown in the fetch diagram, returning the new name, converted image URL and metadata obtained from Sharp.

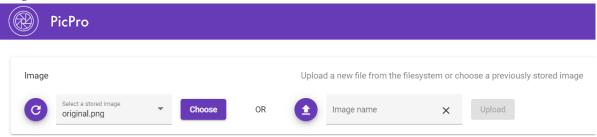
```
const fileNameParts = inputImageName.split('.');
const outputImageName = fileNameParts[0] + '_transformed.';
```

Presets stay largely the same, a JSON object containing the transformation values, alongside a unique name or key. The values are displayed on the frontend, <u>uploading</u> retrieving currently filled in values, and <u>fetching</u> filling in values from the preset. Examples of the preset JSON are explained in the <u>Presets</u> section of Features.

Features

The flow of the data in our project, as well as its architecture, can be best described by the three main features of PicPro: Image, Transformations, and Presets

Image



The **Image** section allows users to either upload an image or choose from a set of stored images. Uploading an image creates a POST request to store them as a Buffer object in an S3 bucket. When accessing stored images, the keys are retrieved from S3 with a GET request so users can select which one to fetch (which uses the user-defined file names as keys, like "original.png"). Only fetching the file names to show in the selection element instead of the whole image data is intentional to reduce network load if the user decides to upload a new file instead. Thus, the actual data – which can be quite large depending on the picture – is only loaded once the user makes the intentional decision to load an image by clicking the "Choose" button. Once an image is chosen, a POST request is made to fetch the actual image data from S3, which is converted to a base64-encoded data URL and sent as a JSON response along with the filename key and metadata from the Sharp module. This gets rerendered and displayed by the frontend to display the output to the user.

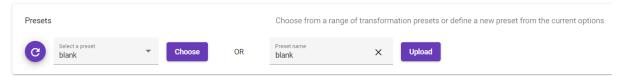
Transformations



{
 outputType: 'PNG',
 flip: false,
 flop: false,
 sharpen: true,
 blur: true,
 greyscale: false,
 blackwhite: false,
 imageName: 'original.png'
 width: 5,
 height: 5,
 rotationAngle: 5,
 brightness: 5
}

The **Transformations** section allows users to select how they want to alter the chosen image. The values adjusted here are stored in a JSON request that gets sent alongside the data of your chosen image. Referring to the code example to the left, the transformation JSON contains the "imageName" which just stores a unique key for the requested image, and "transformations" that stores the values. The checkboxes are stored as Boolean values to identify whether they are to be applied to the image, and the text values get stored as string and number values as they control transformations that can be adjusted or have multiple options. Numerical values are optional so if the user does not define values there, they are also not sent to the server. Users can use the buttons on the bottom right to either apply the transformations or reset the values.

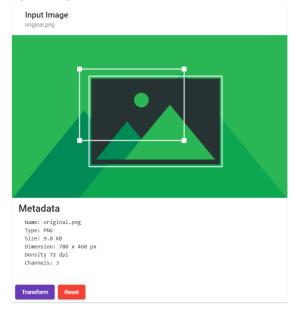
Presets

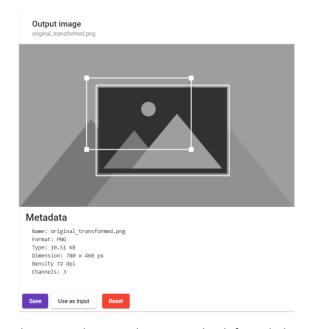


{
 name: 'blank',
 transformation: {
 outputType: 'JPEG',
 flip: false,
 flop: false,
 sharpen: true,
 blur: true,
 greyscale: false,
 blackwhite: false,
 width: 5,
 height: 5,
 rotationAngle: 5,
 brightness: 5
 }
}

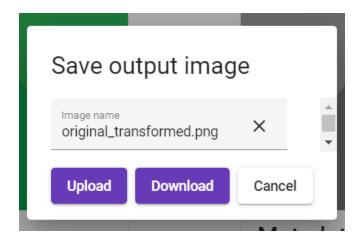
The **Presets** section allows users to store previously used transformations. Uploading a preset takes the values from the Transformation section, as can be seen in the code example to the left. The preset JSON is slightly different from the transformation JSON shown before, being separated into 2 parts: "name" which just stores a unique key for the transformation that can be used later, and "transformation" that stores the values. This is sent as a POST request to the backend that stores the preset JSON in an ElastiCache cluster using Redis. Any previously stored presets have their keys retrieved with a GET request, (like how filenames are displayed in the Image section) displaying the names of all the presets the user can choose from. The same reasoning can be applied here for the choice to only fetch names instead of the whole data as we wanted to focus on computational load while reducing network load. Once chosen, a POST fetch request is made to retrieve the desired preset, which gets sent to the frontend to replace the values in the Transformations section so that they can be used on an image.

Input/Output





The images are displayed below these 3 sections, the original input shown on the left and the transformed output on the right. When a transformation is made, users have the option to save the new image, which brings up a popup dialog (shown below) that lets you choose a filename to either download to the local machine or upload to the S3 bucket. Users can also choose to use the output image again as the input image if further transformations need to be applied to the new image, or reset to use a new image or transformation.

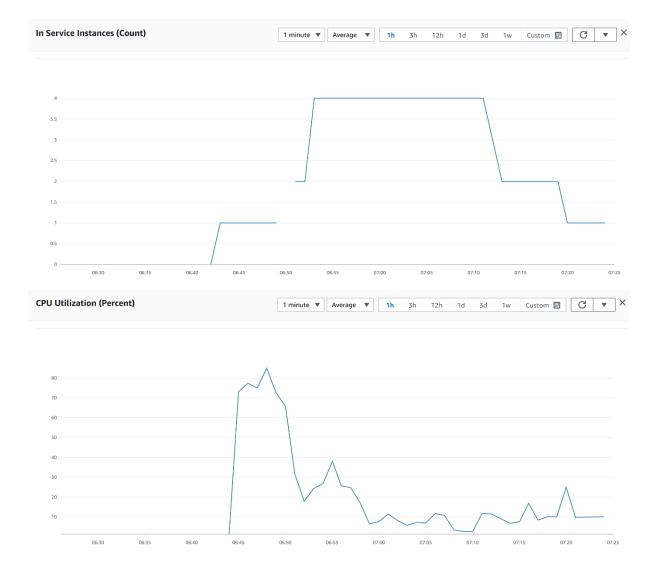


Scaling and Performance

The application receives load from the computationally expensive image transformations and the amount of requests received. A few successive requests with relatively intense transformations like transcoding a large JPEG to a GIF and resizing to e.g. 8000px x 8000px will also max out the CPU capabilities of a single machine. While Sharp is very well optimised and usually performs a transformation pipeline in a few hundred miliseconds, using large input images with high resolution will easily create enough computational load from processing to allow easy scaling.

The scaling setup on AWS is very similar to the respective prac. Machines running frontend and backend are hosted on an LoadBalancer. The scaling policy in the AutoScalingGroup is set to use the avergae CPU utilization as the scaling metric and will scale out at a threshold value of 50%. We experimented with this value a while and found it easier to set it to a lower value for the sake of presentation and testing. A more flexible real-world setting with the potential of more resources would allow increasing the target value for CPU utilization, but we encountered some problems with higher values. If the ASG scaled out to the maximum number of machines quickly and further requests were sent, then the resources were also quickly maxed out resulting in Gateway problems. Investigating these issues were deemed to be out-of-scope for this assignment.

The screnshots shown below provide evidence of a proper deployment of our application in an elastic cloud setting with the ASG configured according to the above mentioned metrics. The first image shows the amount of in-use instances of the LoadBalancer over time. The second image shows the respective CPU utilization over the same time period. It can be observed that the CPU load increased rapidly at first before being cushioned by successful scaling-out. In this experiment, a Postman runner was setup to send 50 high-intensity transformation requests.



The cloud setup was configured according to our architecture diagram. A sidenote would be that in the diagram, the LoadBalancer is placed "behind" the browser block while technically, a single machine will run both the frontend and backend containers. In case of scaling-out, the LoadBalancer will thus also create new instances of the frontend that could be accessed via their instance address and respective port. However, this was pictured intentionally in this way to hide complexity as we will only interact with a single browser and focus very much on data processing in the backend. Thus, the architecture diagram is shown from a user perspective.

Test plan

Task	Expected Outcome	Result	Screenshot (Appendix B)
Image Select	Drop down displayed currently stored images	PASS	<u>1</u>
Image Upload	Loading bar shows during request, image file displays below	PASS	2
Image Fetch	Loading bar shows during request, stored image displays below	PASS	3
Image Section Reset	Input/Output images are empty again	PASS	<u>4</u>

Transform Values	User input is displayed in transformation	PASS	<u>5</u>
	fields/checkboxes		
Transform Image	Transformed image appears as output below	PASS	<u>6</u>
Transform Reset	Transformation values are reset to the input's	PASS	7
	original values		
Preset Select	Drop down displayed currently stored presets	PASS	<u>8</u>
Preset Upload	Transformation values are stored, now available to	PASS	<u>9</u>
	select		
Preset Fetch	Preset values fill in the above transformation values	PASS	<u>10</u>
Input Transform	Transformed image appears as output below	PASS	<u>6</u>
Input Reset	Input/Output images are empty again	PASS	<u>4</u>
Output Upload	Popup appears, image is saved with new name and is	PASS	<u>11</u> , <u>12</u>
	now available to select		
Output Download	Popup appears, image is downloaded with new name	PASS	<u>11</u> , <u>13</u>
Output to Input	Input image is replaced with output image	PASS	<u>14</u>
Output Reset	Output image is empty again	PASS	<u>15</u>
No Image Input	Transform button is not highlighted and can't be	PASS	<u>16</u>
Disables Transform	clicked/triggered		
No Image Chosen	Choose button is not highlighted and can't be	PASS	<u>17</u>
Disables Fetch	clicked/triggered		
No File Chosen	Upload button is not highlighted and can't be	PASS	<u>18</u>
Disables Upload	clicked/triggered		
No Preset Chosen	Choose button is not highlighted and can't be	PASS	<u>19</u>
Disables Fetch	clicked/triggered		

Please refer to the images provided in Appendix B to confirm the passing of the test cases. Some are hard to prove with a simple screenshot. We will be happy to show that these work in the demo.

Difficulties and Reflexion

The setup of the application was fairly straight-forward. We re-used most of the template Matthias implemented for the first assignment with regard to the frontend and backend projects, the dockerization especially for the NGINX webserver and reverse-proxying and the basic setup of the routing in the backend were re-used. The key points in implementation were to figure out in what way the image data can be stored efficiently in S3 so that few steps are necessary to upload and retrieve it. The connection and upload/download from and to S3 was also largely re-used. For Redis and ElastiCache it was mostly the connection and understanding how the cluster works that posed a slight challenge but was quickly resolved.

There aren't any notable differences when comparing the proposal to the final product. The features and use cases mentioned in the proposal were successfully implemented.

When images or presets are uploaded, if the same name is used as a previously stored object, the stored object will be overridden by the new image or preset. There currently isn't a warning in place to let users know they are about to overwrite a stored object, and this could be something to add in a future version.

As no external APIs were used, there weren't any constraints there. The sharp module used to apply the transformations and retrieve metadata was relatively simple to learn and use. The application is however limited to what Sharp can do, so something like applying a mosaic blur would most likely require another module or API to add additional features for future versions.

When uploading presets, users can save them with a blank name, the current application does not require users to fill in the upload text field. This is something that should be fixed in a future version, detecting if there is text in the name field before allowing users to upload a preset

Extensions

These are a few extensions that could possibly be added onto future versions of the application

- Allow users to have a private storage, uploading images/presets that can't be accessed by others
- Add additional image transformation features, like mosaic blur, background removal, etc
- Allow the input and output image to be downloaded as a photoshop psd file, separating some
 of the transformations as separate layers for further editing
- Presets provide a preview of what transformations values them contain, so users don't need to fetch them to see what they are
- Transformations provide a preview of what the image looks like after it's transformed, so users can make more fine-tuned edits

User guide

The "Features" section explains the components of our application in a visual way. In general, a user should either choose a previously uploaded image or upload a new one in the "Image" panel. Then the "Transformation" panel can be opened and settings can be made. The "Transform" button will trigger the processing and display the output image. The transformation settings can be saved in the "Presets" panel. Generally, the UI explains where the functionalities are contained and how it should be used.

For a guide on how to run the Docker containers, please refer to the README file(s) in the repository.

References

Stack Abuse. (2020). Using Buffers in Node.js (Introduction). Digital Ocean. Retrieved October 28, 2022, from https://www.digitalocean.com/commun1.ity/tutorials/using-buffers-in-node-js

Appendices

Appendix A

Appendix 1A – Dockerfile Frontend

```
# Stage 1: Build
FROM node:16-alpine as build
WORKDIR /usr/src/app
COPY package.json package-lock.json ./
RUN npm install
COPY . .
RUN npm run build --prod
# Stage 2: Run
FROM nginx:stable-alpine
# Remove default nginx website
RUN rm -rf /usr/share/nginx/html/*
# Copy dist folder from build stage to nginx public folder
COPY --from=build /usr/src/app/dist/frontend /usr/share/nginx/html
# Copy nginx config file
COPY nginx.conf /etc/nginx/conf.d/default.conf
# Expose port
EXPOSE 80
```

Appendix 2A – Dockerfile Backend

```
FROM node:16-alpine

# Set the work directory to app folder.

# We will be copying our code here

WORKDIR /backend

# Copy package.json file in the node folder inside container

COPY package.json .

# Install the dependencies in the container

RUN npm install

# Copy the rest of the code in the container

COPY . .

# Run the node server with compiled server file

CMD ["node", "server.js"]

# Expose port

EXPOSE 3000
```

Appendix 3A – docker-compose.yml

```
version: '2.2'
services:
backend:
build: ./backend
container_name: backend
image: meder96/picpro:backend
ports:
    - "3000:3000"
frontend:
build: ./frontend
container_name: frontend
image: meder96/picpro:frontend
ports:
    - "4200:80"
```

Appendix 4A – pm2 ecosystem.config.js

```
module.exports = {
    apps: [{
        name: "docker network",
        script: "sudo docker network create net"
    },
    {
        name: "backend",
        script: "sudo docker run --network net --name backend -p 3000:3000 -t
meder96/picpro:backend"
    },
    {
        name: "frontend",
        script: "sudo docker run --network net --name frontend -p 4200:80 -t
meder96/picpro:frontend"
    }]
}
```

Appendix 5A – Image "/" route

```
router.post('/fetch', (req, res) => {
    console.log("

Received request to /images/fetch");
    const fileName = req.body.name;
    const params = {
        Bucket: s3BucketName,
       Key: fileName
    s3.getObject(params).promise()
        .then(data => {
            const imgBuffer = data.Body;
            sharp(imgBuffer).metadata()
                .then(metadataResult => {
                    console.log(` Successfully fetched image
\'${fileName}\' from \'${s3BucketName}\'`);
                    res.send({
                        name: fileName,
                        url: getDataUrlFromBuffer(imgBuffer,
metadataResult.mimeType),
                        metadata: metadataResult
                    });
                })
                .catch(error => {
                    console.error(`X Error during image analysis for
metadata: ${error}`);
                });
        })
        .catch(error => {
            console.error(`X Error during image retrieval of file
\'${fileName}\' from S3: ${error}`);
        });
});
```

```
router.post('/upload', upload, (req, res) => {
    console.log(" A Received request to /images/upload");
    const file = req.file;
    const fileName = req.body.name;
    sharp(file.buffer)
       .metadata()
        .then(metadataResult => {
            // use multipart upload for potentially large files
            var upload = new AWS.S3.ManagedUpload({
                params: {
                    Bucket: s3BucketName,
                    Key: fileName,
                    Body: file.buffer
            });
            upload.promise()
                .then(() => {
                    console.log(` ✓ Uploaded image \'${fileName}\' to
\'${s3BucketName}\'`);
                    res.send(metadataResult);
                })
                .catch(error => {
                    console.error(`X Error during image upload to S3:
${error}`);
                });
        })
        .catch(error => {
            console.error(`X Error during image analysis for metadata:
${error}`);
       });
```

```
router.post('/transform', (req, res) => {
    // main endpoint for image transformation, use sharp here
    console.log("

Received request to /images/transform");
    const inputImageName = req.body.imageName;
    const params = { Bucket: s3BucketName, Key: inputImageName };
    s3.getObject(params).promise()
        .then(data => {
            const imgBuffer = data.Body;
            transform(imgBuffer, req.body)
                .then(outputImgBuffer => {
                    const url = getDataUrlFromBuffer(outputImgBuffer);
                    const fileNameParts = inputImageName.split('.');
                    const outputImageName = fileNameParts[0] +
 transformed.';
                    sharp(outputImgBuffer).metadata()
                        .then(outputMetadata => {
                            res.send({
                                name: outputImageName +
outputMetadata.format.toLowerCase(),
                                url: url,
                                metadata: outputMetadata
                            });
                        })
                        .catch(metadataError => {
                            console.error(`X Error during image analysis for
metadata: ${metadataError}`);
                            res.status(500).send(metadataError);
                        });
                })
                .catch(transformationError => {
                    console.error(`X Error during image transformation:
${transformationError}`);
                    res.status(500).send(transformationError);
                });
        })
        .catch(retrievalError => {
            console.error(`X Error during image retrieval of file
\'${inputImageName}\' from S3: ${retrievalError}`);
            res.status(500).send(retrievalError);
        });
```

```
router.get('/', (_, res) => {
    console.log(" Received request to /presets/");

// return the "name" attributes of all stored preset objects in redis

const keyNames = [];
    redisClient.keys('*').then((result) => {
        for (var i = 0; i < result.length; i++) {
            keyNames.push(result[i]);
        }
        console.log(`i Retrieved ${result.length} presets from redis`);
        res.send(keyNames);
    })
    .catch((err) => {
        console.error(`X Error during preset listing from redis:
${err}`);
        });
})
```

Appendix 10A – Preset "/fetch" route

```
router.post('/fetch', (req, res) => {
    console.log(" A Received request to /presets/fetch");
   // return a specific preset object by its "name" attribute that is passed
in the request body
   // retrieve from redis
    const redisKey = req.body.name;
    redisClient.get(redisKey)
        .then((result) => {
            if (result) {
               console.log(` ✓ Successfully fetched preset \'${redisKey}\'
from Redis`);
               res.send(JSON.parse(result));
        })
        .catch((err) => {
            console.error(`X Error when fetching preset \'${redisKey}\' from
Redis: ${err}`);
       });
```

```
router.post('/upload', (req, res) => {
    console.log(" A Received request to /presets/upload");
   // upload a a preset object that is passed in the request body to redis
and return a boolean if the operation was successful
   // we might change this later to a PUT method depending if we really need
the success boolean, but it's fine for now
   const redisKey = req.body.name;
    const redisJSON = req.body;
    redisClient.set(redisKey, JSON.stringify({ ...redisJSON }))
        .then((result) => {
            console.log(`  Successfully uploaded preset \'${redisKey}\' to
Redis: ${result}`);
            res.send(true);
        .catch((err) => {
            console.error(`X Error during upload of preset \'${redisKey}\'
to Redis: ${err}`);
           res.send(false);
       });
```

```
function transform(imgBuffer, transformation) {
   let transformationPipeline = sharp(imgBuffer).resize(transformation.width, transformation.height);
   // greyscale
   if (transformation.greyscale) {
        transformationPipeline.greyscale();
   if (transformation.blackwhite) {
        transformationPipeline.threshold(100);
   // brightness
   if (transformation.brightness) {
        transformationPipeline.modulate({
            brightness: transformation.brightness,
   if (transformation.saturation) {
        transformationPipeline.modulate({
            saturation: transformation.saturation,
   if (transformation.hue) {
        transformationPipeline.modulate({
            hue: transformation.hue,
   if (transformation.blur) {
        transformationPipeline.blur(transformation.blur);
   if (transformation.rotationAngle) {
        transformationPipeline.rotate(transformation.rotationAngle);
```

Appendix 13A – Transform function part 2

```
// flip
if (transformation.flip) {
    transformationPipeline.flip();
}

// flop
if (transformation.flop) {
    transformationPipeline.flop();
}

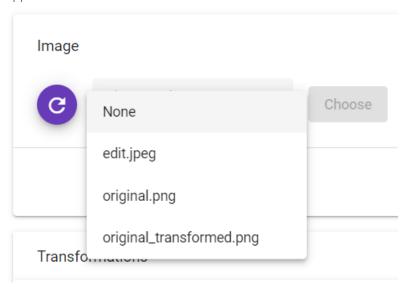
// sharpen
if (transformation.sharpen) {
    transformationPipeline.sharpen();
}

// transcode
if (transformation.outputType) {
    transformationPipeline.toFormat(transformation.outputType);
}

// generate buffer
return transformationPipeline.toBuffer();
}
```

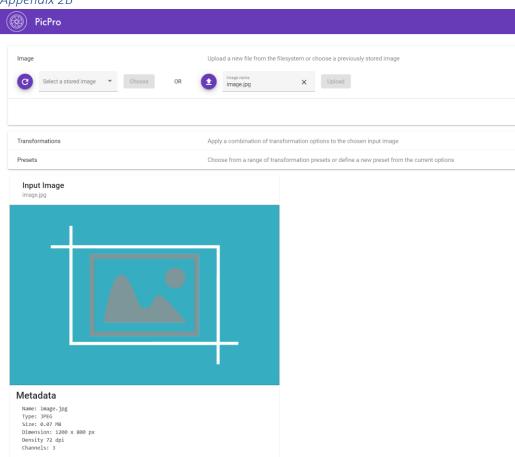
Appendix B

Appendix 1B



Appendix 2B

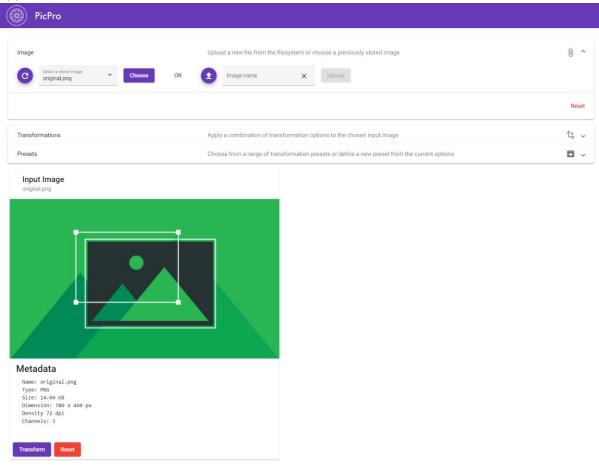
Transform Reset



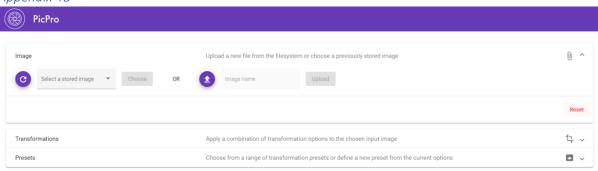
0 ^

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Appendix 3B



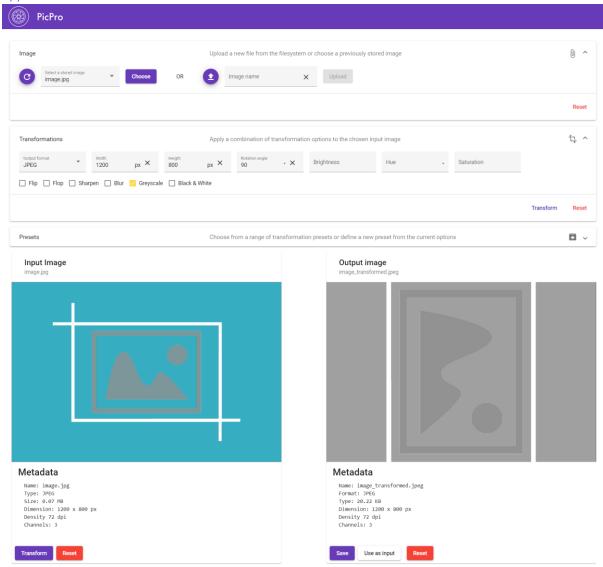
Appendix 4B



Appendix 5B



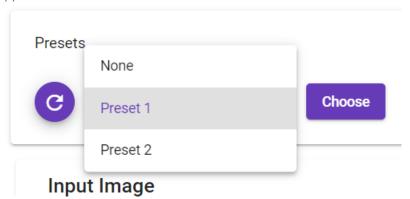
Appendix 6B



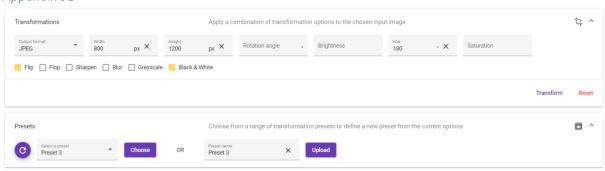
Appendix 7B



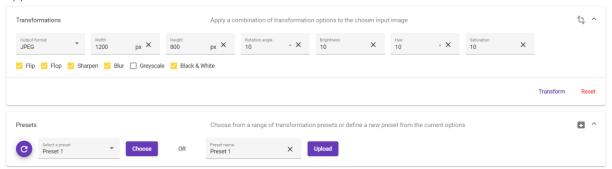
Appendix 8B



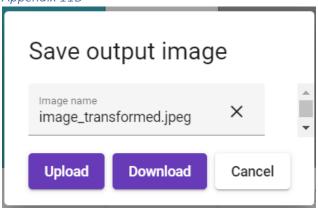
Appendix 9B



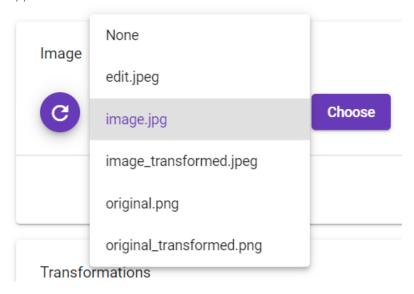
Appendix 10B



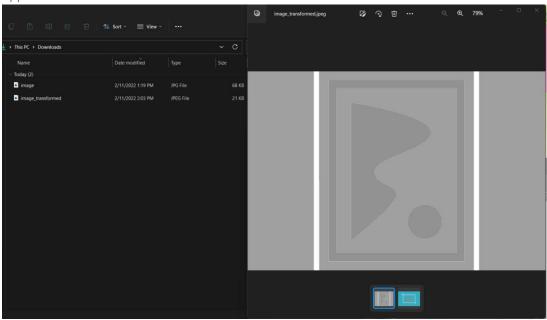
Appendix 11B



Appendix 12B



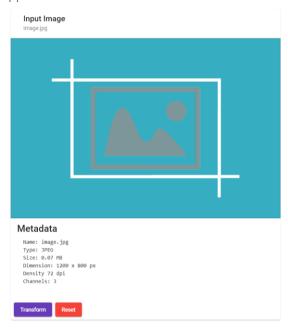
Appendix 13B



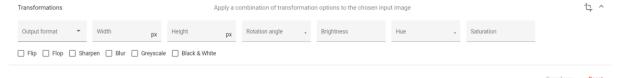
Appendix 14B



Appendix 15B

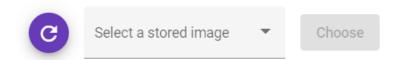


Appendix 16B



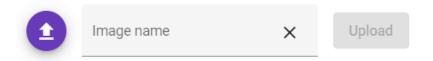
Appendix 17B

Image



Appendix 18B

Upload a new file from the filesystem or choose a previou



Appendix 19B

Presets

