

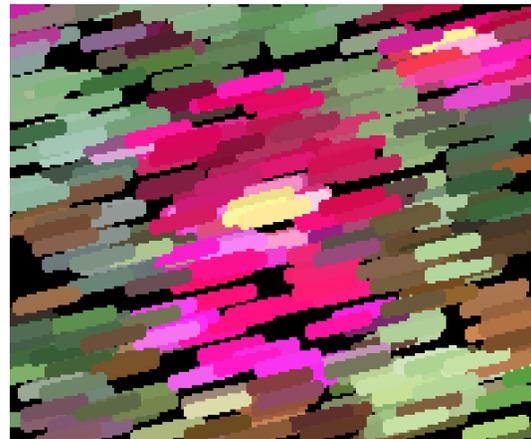
Due: Monday 23 Feb by 11:00pm

Worth: 10%

Background. In this assignment you will implement a basic system to produce a rendering of an input image in the style of a painting. Download the paper by Peter Litwinowicz on painterly rendering, available from the course webpage. The paper goes beyond what we will do in this assignment. In particular, we will not be concerned with rendering image sequences (i.e., ignore section 3.C on frame-to-frame coherence). You can also ignore brush textures, anti-aliased stroke rendering, and thin-plate spline interpolation. Nevertheless, the parts of the paper that we do not implement are both readable and motivating.

Note: in this handout, we use “edgel” to mean a pixel at which an edge point, i.e., a pixel where there is an edge in the image.

Handout Code. In `painterlyHandout.zip`, we included a test image and some simple starter Python code. Running the handout Python script in `paintrend.py` produces the result shown below.



The main block of `paintrend.py` takes an image as input, and then randomly samples image pixels and renders fat brush strokes on a new canvas. The brush strokes have the RGB values of the sampled pixel from the original image. The radius, orientation and length of each brush stroke are fixed constants (although some strokes are cropped to fit within the original image frame). Your job is to build on this handout code as described below.

What to Hand In. Include a write-up describing your implementation of the various parts of the program and its output. There is no need to repeat the details described in the handout: just describe any changes you might have made. In particular, provide the specific values of the parameters you used (for example, the Gaussian standard deviations for edge detection and for gradient orientation estimation, the thresholds on the minimum size of the gradient to be used in these two stages, and the stroke radius and maximum half-length parameters). **For each part, include the output for two images: orchid.jpg and any image that you choose to work with. Please submit your image separately as well.**

Submit your report as `pdf` and `texfiles`, and submit your python code. Submit any images that you used (Tip: do not use images that cannot fit on a page in order to keep the file size down.)

Your `__main__` block should generate the images that you are using in the report.

Part 1. Covering the Canvas: Blind Sampling (5 points)

Your first task is to ensure that the canvas is covered by painted strokes. The handout code simply uses a for loop to select a fixed number of strokes centers at random points, and as a result, it may leave some

pixels unpainted. The unpainted pixels have RGB values of -1 . You should replace this `for` loop, with a `while` loop that generates random paint strokes while there is at least one unpainted pixel left in the canvas. You could use `numpy.where` for finding pixels with the RGB value of -1 .

Part 2. Covering the Canvas: Systematic Sampling (5 points)

Next, replace the blind random sampling of center pixels by a more systematic sampling of unpainted pixels. In particular, initialize the canvas to -1 , and at each iteration locate all pixels which remain unpainted. Randomly select one of these pixels as the center of the next paint stroke, and then paint the corresponding stroke on the canvas (as done in the handout-code). As described in Litwinowicz’s paper, it is important to randomly select the next center pixel to use, rather than processing the pixels in scan-line order, in order to avoid the appearance of a systematic ordering of the strokes.

Part 3. Computing Canny Edgels (5 points)

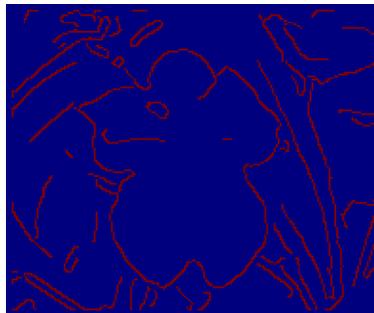
Your next task is to compute a Canny edgel image. This can be by using the code in the Python file `canny.py`. The canny code should run on a single monochrome image. Litwinowicz suggests using the intensity image

$$I(\vec{x}) = 0.30R(\vec{x}) + 0.59G(\vec{x}) + 0.11B(\vec{x}), \quad (1)$$

where $R(\vec{x})$, $G(\vec{x})$, and $B(\vec{x})$ are the three colour channels of the original image.

For the edge detection, use $\sigma = 2.0$ for the standard deviation of the Gaussian derivative filters (used by the Canny edge detector). Adjust the threshold on gradient amplitudes so that most of the edgels separating distinct regions of the image are detected but, at the same time, not too many edgels appear in other image regions. There is no perfect threshold. You may wish to adjust this threshold after the clipping code described in the next section is implemented. Include the canny edge image in your report.

In the following you see the edge image that we obtained by setting some parameters. Obviously, your results don’t have to look identical completely identical.



Part 4. Clipping Paint Strokes at Canny Edges (5 points)

Given the binary image marking the locations of Canny edgels, use this image to clip painted strokes. One approach for doing this is described in Appendix A of Litwinowicz’s paper. We will make a few simplifications to this process.

Suppose \vec{c} is the integer-valued pixel location for the center of the stroke to be painted. Let $\vec{t} = (t_1, t_2)^T$ be the tangent direction (non-integer-valued) for the stroke.

If there is an edgel at pixel \vec{c} , paint the stroke with a length of 0 using `paintStroke(...)`. This will paint a disk of the stroke radius centered on \vec{c} . Otherwise, we wish to walk along the line in direction \vec{t}

from pixel \vec{c} until we either find an edgel in the Canny edge map or we have gone more than the distance `halfLen` from the center pixel. One endpoint of the paint stroke is determined by the first edgel found along this line or, if no edgel is found, then the last pixel within the maximum distance `halfLen` of the start pixel \vec{c} is used. The other endpoint of the segment is determined in the similar way, walking in the opposite direction \vec{t} from \vec{c} . (You can use loops to find these endpoint positions.)

One remaining detail is specifying which discrete pixels are visited when we walk along the line segment starting at \vec{c} in the direction \vec{t} . Suppose $|t_1| \geq |t_2|$, so the desired stroke is within 45 degrees of horizontal. Then let $\vec{s} = \vec{t}/|t_1|$, so the first component of \vec{s} is ± 1 and the second component satisfies $|s_2| \leq 1$. Then for $k = 1$ up to K we visit pixels

$$\vec{x}_k = \vec{c} + \text{round}(k\vec{s}). \quad (2)$$

Notice \vec{x}_k is integer-valued and, since $s_1 = \pm 1$, we always step exactly one pixel to the left or right each time we increment k . The maximum value K is determined by the largest k such that $\|\vec{x}_k - \vec{c}\| \leq \text{halfLen}$. For directions which are closer to being vertical, so $|t_2| > |t_1|$, use $\vec{s} = \vec{t}/|t_2|$ instead. In this case, $s_2 = \pm 1$ and \vec{x}_k increments one pixel up or down with each step in k . Note we are simply using the Canny edgel positions to crop the stroke, rather than looking for a decrease in the gradient amplitude, as described in Litwinowicz's paper. Also, notice that we only check for edgels along the center line of the painted stroke, as determined by the pixels \vec{x}_k in Equation (2), ignoring any edgels that may intersect the stroke off of this line (recall the stroke has a nonzero radius and may be significantly wider than this center line).

In order to make sure that your code for this part is correct, reduce the paint stroke radius to 1 (`rad` variable) and check the result. Include the output with `rad = 1` in your report too.

Our implementation with the provided random seed resulted in the following output for `orchid.jpg`:



Part 5. Orienting the Paint Strokes (5 points)

So far the strokes all have the same orientation θ . We wish to set the orientation of each stroke to be normal to the gradient of image intensity at the initial pixel \vec{c} of the stroke. This will cause the strokes to be roughly aligned with the edge orientation, and with contours of constant intensity.

In order to pre-compute a smooth image of gradient directions, first filter the intensity image with derivatives of Gaussians, as in the Canny edge detector. However, use a larger standard deviation σ for these Gaussians than was used to compute the Canny edgels (eg. use $\sigma = 4$ instead of $\sigma = 2$). Also, in order to cover most of the image with estimated gradient directions, use a much smaller threshold on the minimum size of the gradient to be considered. The combined effect of using a larger value of σ and a lower threshold on the gradient size should provide a gradient orientation estimate at most pixels.

Python code for computing such a gradient direction image can be extracted from `canny.py`. In particular, see the computation of the array of gradient directions `grad`. The gradient angle `theta[x, y]` at a pixel `[x, y]` is computed. Gradients lower than a threshold are cleared in the `grad` and `theta` array.

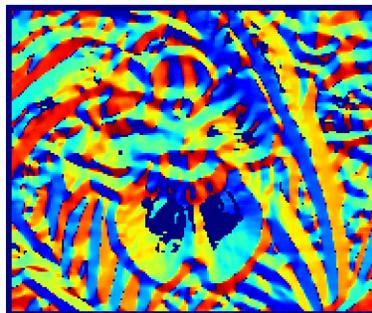
A paint stroke centered at pixel \vec{c} should then be drawn with orientation $\theta(\vec{c}) + \pi/2$, where $\theta(\vec{c})$ is the gradient orientation as described in the previous paragraphs. That is, the tangent to the stroke is defined to be $\vec{t} = (\cos(\theta(\vec{c} + \pi/2)), \sin(\theta(\vec{c} + \pi/2))$. The addition of $\pi/2$ makes the stroke orthogonal to the gradient, and therefore roughly parallel to curves of constant intensity. The stroke should then be clipped by any intervening edgels, as described in the previous section.

One remaining detail is to decide what to do at pixels which have gradients whose length is below the minimum threshold. The paper by Litwinowicz suggests interpolating neighbouring values of the gradient using something called a thin-plate spline. This is beyond the scope of this assignment. Instead, here we will use a constant default direction θ_0 for any stroke centered at these pixels.

Our implementation with the provided random seed with `halfLen 5`, resulted in the following outputs for `orchid.jpg`:



We set the threshold on theta to be very low to produce these results. Here is the visualization of the `theta` array (before adding $\pi/2$ to it) using `colorImSave()`:



Part 6. Random Perturbations (5 points)

Finally, in the subsection titled Random Perturbations in Section 3.A of Litwinowicz's paper, the addition of random variations to both the colour and the stroke orientation is described. Implement these random variations. Note that here our RGB values are in the range $[0, 1]$ while in the paper they are assumed to be $[0, 255]$. Thus instead of perturbing the colour coefficients by random amounts in the range $[15, 15]$, use $[15/255.0, 15/255.0]$ instead. Also perturb the intensity and the stroke orientation, as described in this subsection of the Litwinowicz's paper.