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## TP 1: Texture Synthesis

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You can find my source code here on my repository.

The repository is the same for all the TPs.

## **Implementation**

I have implemented the *Efros-Leung Agorithm* in C++ after having tried in octave. The main reason is the time it takes to compute the algorithm with octave whereas in C++ it is significantly faster.

You can find the implementation at src/tp1.cpp. To compile it, being located at the root of the project:

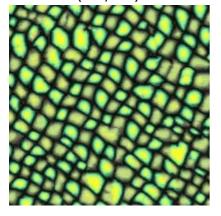
```
mkdir build cd build cmake ..
```

Compile and run (in the build directory):

```
make -j$(nproc) && ./tp1
```

Here is the synthesis with the sample data/synthese/text0.png.

The size is (200, 200) with n=9 and epsilon=0.05. It took 2 minutes and 17 sec.



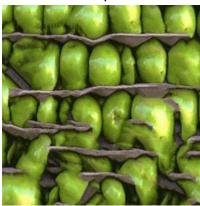
## Discussion

Tweaking the parameters is important to obtain images similar to te sample.

- n define the size of the patch that we use to compute the similarity between a portion of I, the synthesis image and Ismp, the sample we're trying to extend.
- epsilon define how similar a patch used to fill a pixel is to the closest patch of the portion of I we are looking.

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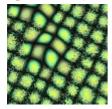
With an appropriate epsilon and n, we get a result like the image above. Otherwise, with n too big, we compare big portion of images and this could lead to pixels changing abruptly, if the patterns on the sample are not big enough. With n to little, we loose the 'sense' of the image, logic of the pattern and it becomes chaos (and monstruous) ((200,200) n=7 eps=0.05):



For epsilon, since i have normalize my distance between 0 and 1, 0 means always choosing the best fit and 1 is choosing completely randomly between all the patchs of Ismp. The closer we get to 1, the more noisy the image can be.

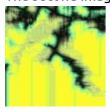
But those 2 parameters works together, even with a high epsilon, we can still get a plausible image because n allow to keep 'sense' to the image, but the inverse isn't true.

Also having both parameters high, or low gives bad results: The first image is (100, 100) with n=21 and epsilon=1,



We can notice the first patch copy past, which is clean, and the noise all around.

The second images is the same but with n=1 and epsilon=0,



## **Improvements**

We could use the algorithm *Patch Match* to improve the research for closest patch, but having a high epsilon would increase time since we will have to compute more distances.