

Image processing with scikit-image

Emmanuelle Guillard

Surface, Glass and Interfaces, CNRS/Saint-Gobain

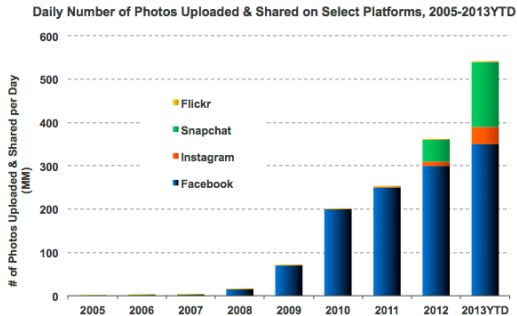
Paris-Saclay Center for Data Science



scikit-image
image processing in python

The world is getting more and more visual

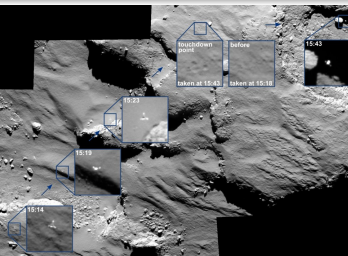
**Photos = 500MM+ Uploaded & Shared Per Day,
Growth Accelerating, on Trend to Rise 2x Y/Y...**



KPCB

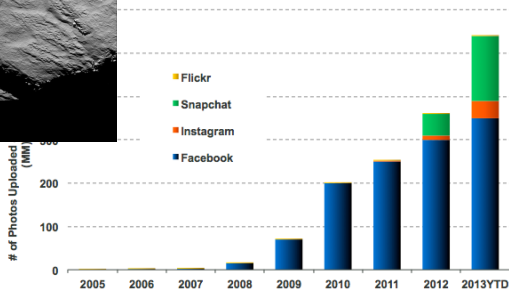
Source: KPCB estimates based on publicly disclosed company data. 14

The world is getting more and more visual



MM+ Uploaded & Shared Per Day,
operating, on Trend to Rise 2x Y/Y...

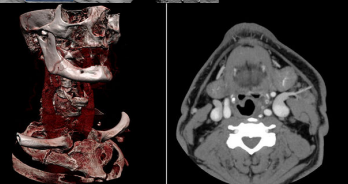
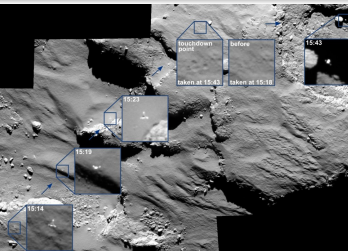
of Photos Uploaded & Shared on Select Platforms, 2005-2013YTD



KPCB

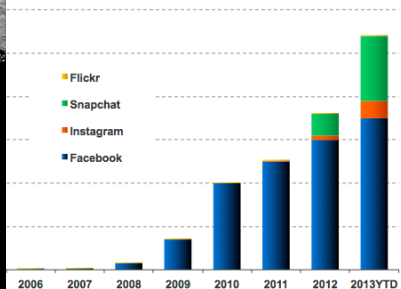
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The world is getting more and more visual



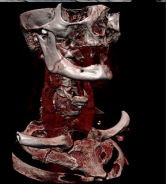
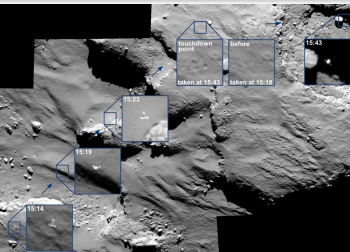
MM+ Uploaded & Shared Per Day,
generating, on Trend to Rise 2x Y/Y...

of Photos Uploaded & Shared on Select Platforms, 2005-2013YTD



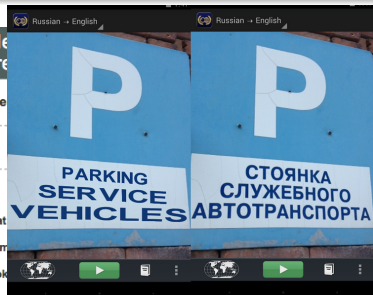
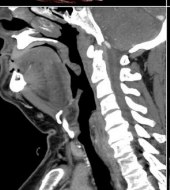
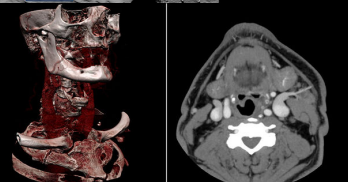
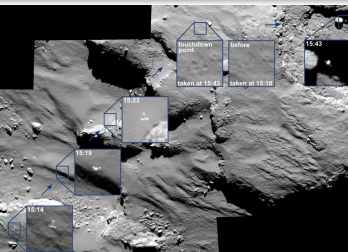
Source: KPCB estimates based on publicly disclosed company data. 14

The world is getting more and more visual



Source: KPCB estimates based on publicly disclosed company data. 14

The world is getting more and more visual



The world is getting more and more visual

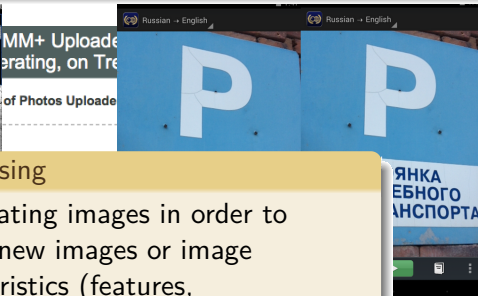
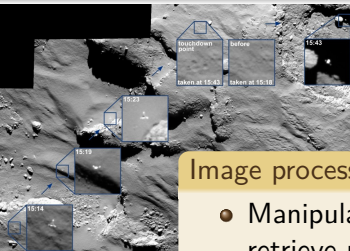
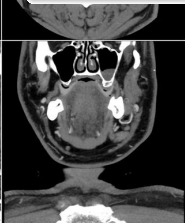
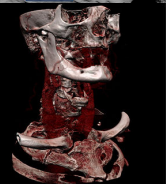


Image processing

- Manipulating images in order to retrieve new images or image characteristics (features, measurements, ...)
- Often combined with machine learning



<http://scikit-image.org/>

A module of the Scientific Python stack

- Language: Python
 - Core modules: NumPy, SciPy, matplotlib
 - Application modules: scikit-learn, scikit-image, pandas, ...

A general-purpose image processing library

- open-source (BSD)
- not an application (ImageJ)
- less specialized than other libraries (e.g. OpenCV for computer vision)



scikit-image
image processing in python

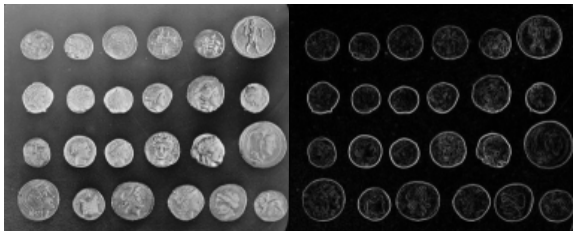
1 Principle

1 First steps

```
from skimage import data, io, filter

image = data.coins() # or any NumPy array!
edges = filter.sobel(image)
io.imshow(edges)
io.show()
```

x

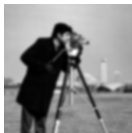


My environment: IPython interpreter + text editor
Ipython notebook nice for demos/trial and error

1 Manipulating images as numpy arrays

- numpy arrays as arguments and outputs

```
>>> from skimage import io, filter
>>> camera_array = io.imread('camera_image.png')
>>> type(camera_array)
<type 'numpy.ndarray'>
>>> camera_array.dtype
dtype('uint8')
>>> filtered_array = filter.gaussian_filter(
    camera_array, sigma=5)
>>> type(filtered_array)
<type 'numpy.ndarray'>
>>> filtered_array.dtype
dtype('float64')
```



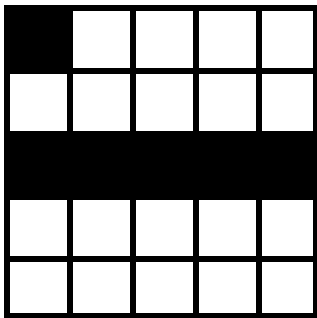
<http://www.numpy.org/>
scipy-lectures.github.io/intro/numpy/index.html

1 Manipulating images as numerical (numpy) arrays

- Pixels are arrays elements

```
import numpy as np
image = np.ones((5, 5))
image[0, 0] = 0
image[2, :] = 0
```

x



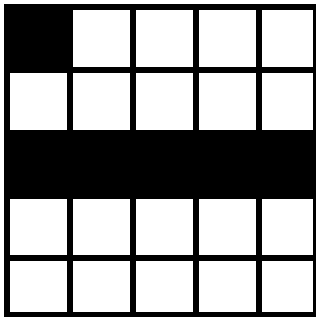
(use matplotlib for visualization: `matplotlib.pyplot.imshow`)

1 Manipulating images as numerical (numpy) arrays

- Pixels are arrays elements

```
import numpy as np
image = np.ones((5, 5))
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```

x



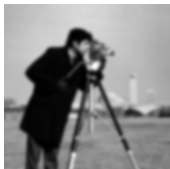
(use matplotlib for visualization: `matplotlib.pyplot.imshow`)

1 Some magics inside

Don't let yourself be tricked by integer / float conversion!

```
>>> from skimage import data, filter
>>> camera_array = data.camera()
>>> camera_array.dtype
dtype('uint8')
>>> filtered_array = filter.gaussian_filter(
    camera_array, sigma=5)
>>> filtered_array.dtype
dtype('float64')
>>> camera_array.min(), camera_array.max()
(0, 255)
>>> filtered_array.min(), filtered_array.max()
(0.031287384322526979, 0.8560994897846772)
```

x



1 An API relying mostly on functions

```
skimage.filter.gaussian_filter(image, sigma, output=
    None, mode='nearest', cval=0, multichannel=None)
```

Multi-dimensional Gaussian `filter`

Parameters

`image` : array-like

`input` image (grayscale **or** color) to `filter`.

`sigma` : scalar **or** sequence of scalars

standard deviation **for** Gaussian kernel. The
standard

deviations of the Gaussian `filter` are given **for**
each axis as a

sequence, **or** as a single number, **in** which case it
is equal **for**

`all` axes.

`output` : array, optional

The “output” parameter passes an array **in** which
to store the

`filter` output.

1 Images and dimensions

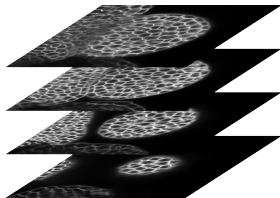


```
>>> data.camera().shape  
(512, 512)
```

- Most functions suitable for 2-D gray- or color-scale images
- Some functions work with 3D images as well
- Check out `scipy.ndimage` for n-d functionality.



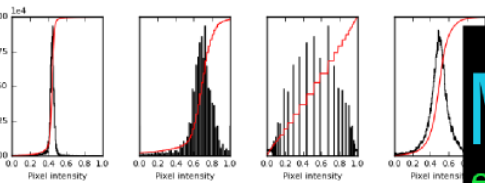
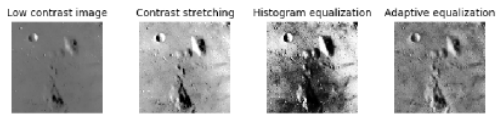
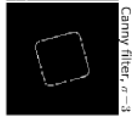
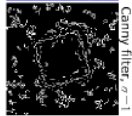
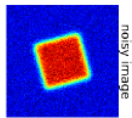
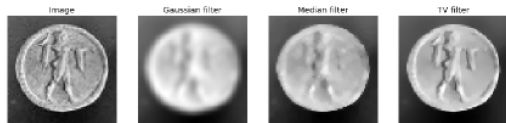
```
>>> coffee.shape  
(400, 600, 3)  
>>> red_channel =  
coffee[..., 0]
```



(d_0, d_1, d_2)

2 Some features

2 Filtering: transforming image data



denoising sobel
equalize wiener
Median
Gaussian canny
enhance _contrast
total_variation

`skimage.filter`, `skimage.exposure`, `skimage.restoration`

2 From very simple/classical algorithms

noisy



Gaussian filter



Median filter



```
from skimage import data, filter, color
from skimage.morphology import disk

l = data.lena()
l = color.rgb2grey(l)
l = l[230:290, 220:320]

noisy = l + 0.4 * l.std() * np.random.random(l.shape)

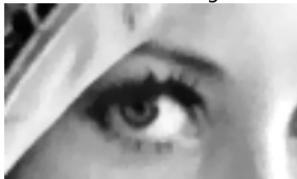
gaussian_denoised = filter.gaussian_filter(noisy,
    sigma=2)
median_denoised = filter.rank.median(noisy, disk(3))
```

2 To more advanced/recent algorithms

noisy



TV denoising



(more) TV denoising



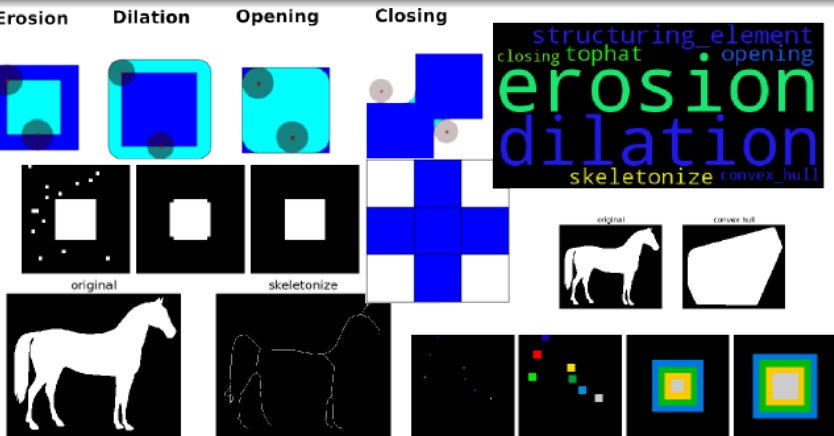
```
from skimage.filter import tv_denoise
from skimage import data

I = data.lena()
I = I[230:290, 220:320]
noisy = I + 0.4*I.std()*np.random.random(I.shape)

tv_denoised = tv_denoise(noisy, weight=10)
```

x

2 Mathematical morphology



2 Extracting features

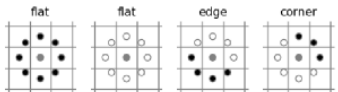
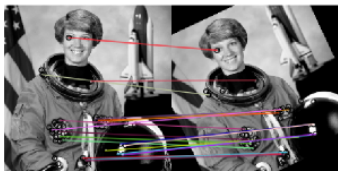
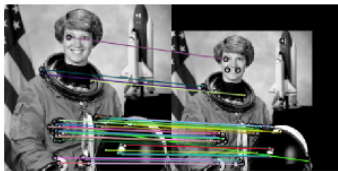
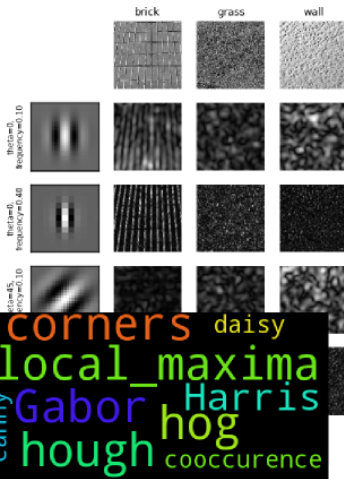


Image responses for Gabor filter kernels



`skimage.feature`, `skimage.filter`

2 Geometrical transformations

`skimage.transform`

scale, zoom, rotate, swirl, warp, ...



2 Segmentation: labelling regions



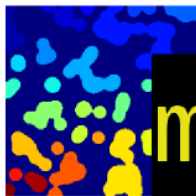
2 Feature extraction followed by classification

Combining `scikit-image` and `scikit-learn`

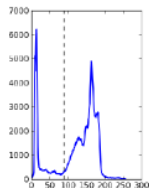
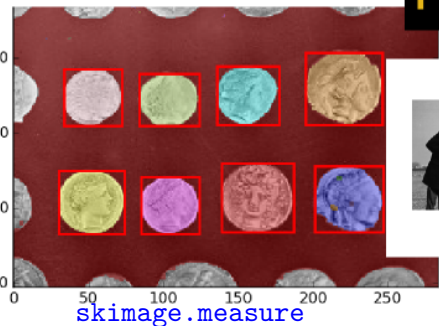
- Extract features (`skimage.feature`)
 - Pixels intensity values (R, G, B)
 - Local gradients
 - More advanced descriptors: HOGs, Gabor, ...
- Train classifier with known regions
 - here, random forest classifier
- Classify pixels



2 Measures on images

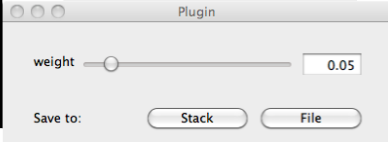
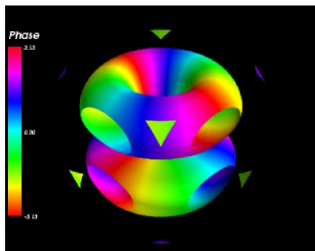
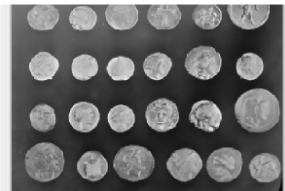
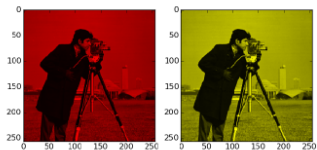


size label
measure
histogram
regionprops



2 Visualizing images and more

matplotlib, mayavi



Image

Sobel+Watershed

SLIC superpixels

Join



2 Development of scikit-image

- Mature algorithms
- Only Python + Cython code for easier maintainability
- Hosted on GitHub
- Thorough code review by others: readability, PEP8, efficiency, ...
- 1-2 releases per year
- Core team of 5 persons (+ GSoc students)



2 Getting started: installing scikit-image

<http://scikit-image.org/docs/dev/install.html>

- Packaged on Ubuntu/Debian
- Shipped with all major Scientific Python distributions:
Enthought Canopy, Anaconda, Python(x,y)

2 Getting started: finding documentation



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Search documentation ...

Image processing in Python

scikit-image is a collection of algorithms for image processing. It is available **free of charge and free of restriction**. We pride ourselves on high-quality, peer-reviewed code, written by an active **community of volunteers**.

[Download](#)



Stable

0.10.1 - June 2014

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Development

pre-0.11

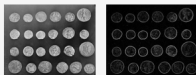
[Download](#)

Getting Started

Filtering an image with *scikit-image* is easy! For more examples, please visit our [gallery](#).

```
from skimage import data, io, filter
```

```
image = data.coins() # or any NumPy array!  
edges = filter.sobel(image)  
io.imshow(edges)  
io.show()
```



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Links

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Related Projects

[OpenCV](#)

[Scikit-learn](#)

[Mahotas](#)

[SimpleCV](#)

[Ilastik](#)

If you find this project useful, please cite:

[BIBTeX]

Stéfan van der Walt, Johannes L. Schönberger, Juan Nunez-Iglesias, François Boulogne, Joshua D. Warner, Neil Yager, Emmanuelle Guillard, Tony Yu and the scikit-image contributors. *scikit-image: Image processing in Python*. *PeerJ* 2:e453 (2014) <http://dx.doi.org/10.7717/peerj.453>

Announcements

2 Getting started: finding documentation



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API Reference

- `skimage`
 - Subpackages
 - Utility Functions
 - `dtype_limits`
 - `img_as_bool`
 - `img_as_float`
 - `img_as_int`
 - `img_as_ubyte`
 - `img_as_uint`
 - `test`
- Module: `color`
 - `combine_stains`
 - `convert_colorspace`
 - `deltaE_cie76`
 - `deltaE_ciede2000`
 - `deltaE_ciede94`
 - `deltaE_cmc`
 - `gray2rgb`
 - `guess_spatial_dimensions`
 - `hed2rgb`
 - `hsv2rgb`
 - `lab2lch`
 - `lab2rgb`
 - `lab2xyz`
 - `label2rgb`
 - `lch2lab`
 - `luv2rgb`
 - `luv2xyz`
 - `rgb2gray`
 - `rgb2grey`
 - `rgb2hed`
 - `rgb2hsv`
 - `rgb2lab`
 - `rgb2luv`

threshold_otsu

`skimage.filter.threshold_otsu(image, nbins=256)`

Return threshold value based on Otsu's method.

Parameters:	image : array Input image.
	nbins : int, optional Number of bins used to calculate histogram. This value is ignored for integer arrays.
Returns:	threshold : float Upper threshold value. All pixels intensities that less or equal of this value assumed as foreground.

References

[Wikipedia, http://en.wikipedia.org/wiki/Otsu's_Method](http://en.wikipedia.org/wiki/Otsu's_Method)

Examples

```
>>> from skimage.data import camera
>>> image = camera()
>>> thresh = threshold_otsu(image)
>>> binary = image <= thresh
```

threshold_yen

`skimage.filter.threshold_yen(image, nbins=256)`

Return threshold value based on Yen's method.

Parameters:	image : array Input image.
	nbins : int, optional

2 Gallery of examples



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General examples

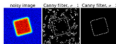
General-purpose and introductory examples for the scikit.



Blob Detection



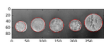
BRIEF binary descriptor



Canny edge detector



CENSURE feature detector



Circular and Elliptical Hough Transforms



Contour finding



Convex Hull



Corner detection



Dance DAVIS feature

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[skimage 0.7.0](#)
[skimage 0.6](#)
[skimage 0.5](#)
[skimage 0.4](#)
[skimage 0.3](#)

2 Getting started: finding documentation

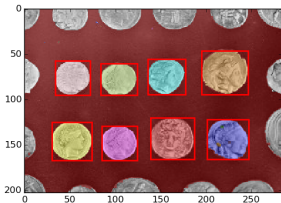


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Label image regions

This example shows how to segment an image with image labelling. The following steps are applied:

1. Thresholding with automatic Otsu method
2. Close small holes with binary closing
3. Remove artifacts touching image border
4. Measure image regions to filter small objects



```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches

from skimage import data
from skimage.filter import threshold_otsu
from skimage.segmentation import clear_border
from skimage.morphology import label, closing, square
from skimage.measure import regionprops
from skimage.color import label2rgb

image = data.coins()[50:-50, 50:-50]

# apply threshold
thresh = threshold_otsu(image)
bw = closing(image > thresh, square(3))

# remove artifacts connected to image border
cleared = bw.copy()
clear_border(cleared)

# label image regions
label_image = label(cleared)
borders = np.logical_xor(bw, cleared)
label_image[borders] = -1
image_label_overlay = label2rgb(label_image, image=image)

fig, ax = plt.subplots(ncols=1, nrows=1, figsize=(6, 6))
ax.imshow(image_label_overlay)

for region in regionprops(label_image):

    # skip small images
    if region.area < 100:
        continue

    # draw rectangle around segmented coins
    minr, minc, maxr, maxc = region.bbox
    rect = mpatches.Rectangle((minc, minr), maxc - minc, maxr - minr,
                              fill=False, edgecolor='red', linewidth=2)
    ax.add_patch(rect)

plt.show()
```

2 Conclusions

scikit-image

- An image processing Python module relying on NumPy arrays
- Trade-off between performance and usability
- More and more features
- Try it out!
 - <http://scikit-image.org/>
 - <https://www.youtube.com/watch?v=SE7h0IWD93Y>
(and others)
 - <http://scipy-lectures.github.io/packages/scikit-image/>