

# Image processing with scikit-image

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**scikit-image**  
image processing in python

# The revolution of images

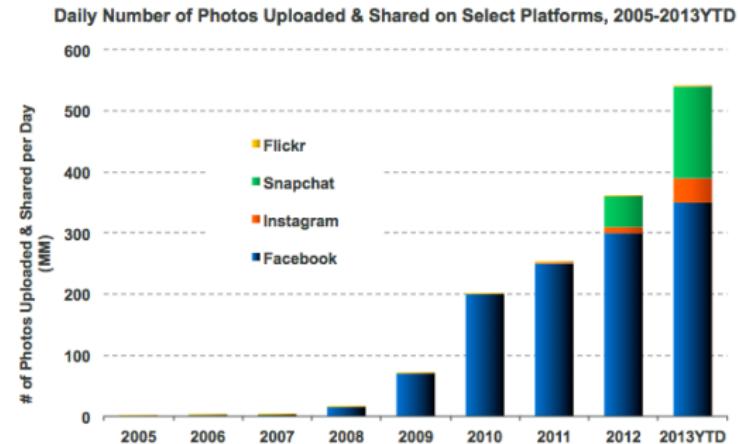


# The revolution of images



# A flood of images

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

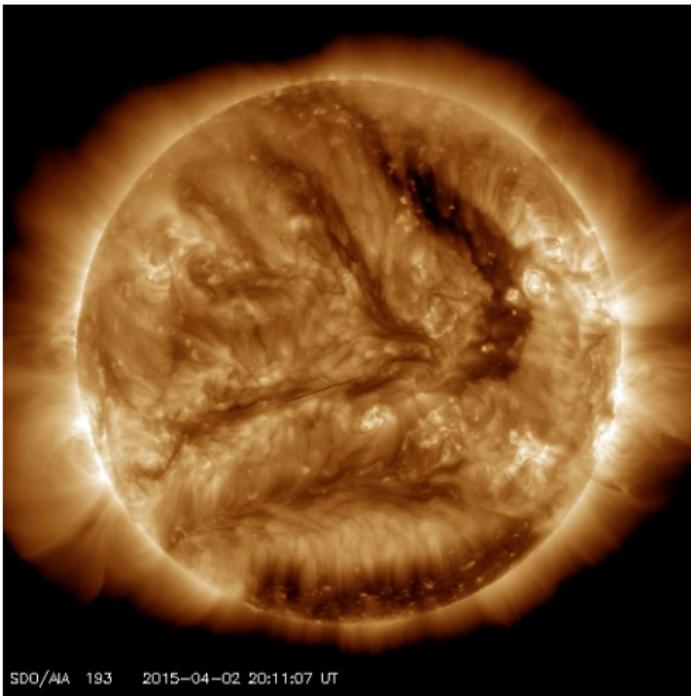


KPCB

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

several  $10^8$  images uploaded on Facebook each day

# A flood of images

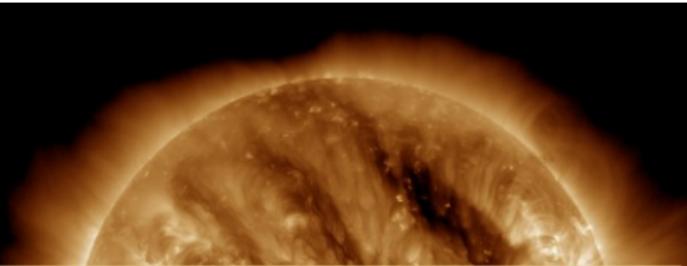


SDO/AIA 193 2015-04-02 20:11:07 UT

hundreds of terabytes of scientific data for scientific experiment

<http://sdo.gsfc.nasa.gov/>

# A flood of images



## Image processing

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

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hundreds of terabytes of scientific data for scientific experiment

<http://sdo.gsfc.nasa.gov/>

# Why use scikit-image?

Lots of excellent tools available: OpenCV (computer vision),  
ImageJ (GUI-based software), ITK, ...

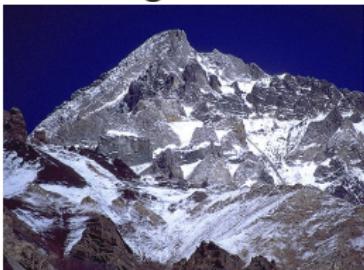
- Native NumPy compatibility



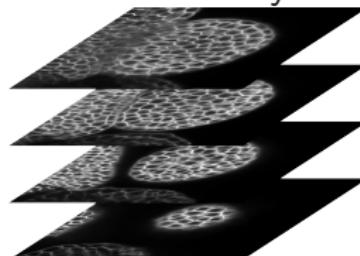
- Variety of features



- Gentle learning curve



- Versatile - 3D-friendly



# Datasheet

## Package statistics

- <http://scikit-image.org/>
- Release 0.11 (1 - 2 release per year)
- Available in most Scientific Python Distributions: Canopy, Anaconda, ...
- Packaged on Ubuntu/Debian
- Among 1000 best ranked packages on PyPi



## Development model

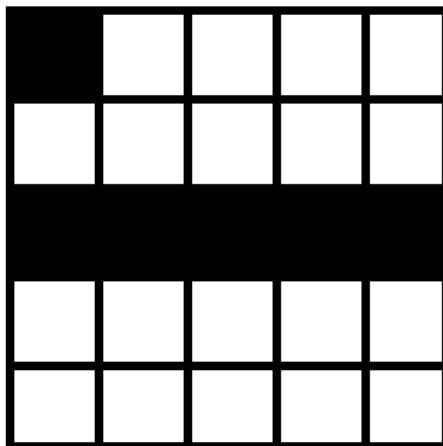
- Mature algorithms
- Only Python + Cython code for easier maintainability
- Thorough code review by others: readability, PEP8, efficiency, ...
- Core team of 5 – 10 persons (close to applications)



# Manipulating images as numerical (numpy) arrays

- Pixels are arrays elements

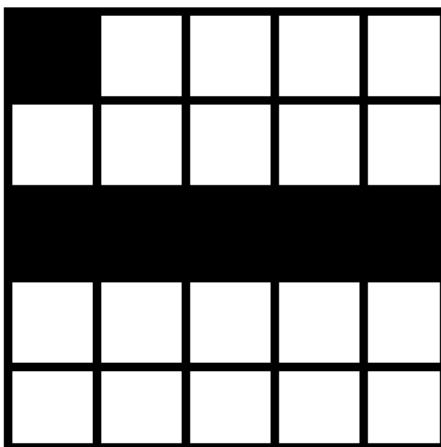
```
import numpy as np  
image = np.ones((5, 5))  
image[0, 0] = 0  
image[2, :] = 0  
*
```



# 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  
*
```



```
>>> coffee.shape  
(400, 600, 3)  
>>> red_channel =  
coffee[..., 0]  
>>> image_3d =  
np.ones((100, 100, 100))
```

# NumPy-native: images as NumPy arrays

- NumPy arrays as arguments and outputs

```
>>> from skimage import io, filters  
>>> camera_array = io.imread('camera_image.png')  
>>> type(camera_array)  
<type 'numpy.ndarray'>  
>>> camera_array.dtype  
dtype('uint8')  
>>> filtered_array = filters.gaussian_filter(  
    camera_array, sigma=5)  
>>> type(filtered_array)  
<type 'numpy.ndarray'>  
>>> import matplotlib.pyplot as plt  
>>> plt.imshow(filtered_array, cmap='gray')  
x
```



# An API relying mostly on functions

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

Multi-dimensional Gaussian filter

Parameters

-----

image : array-like

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.

# Getting started: finding documentation

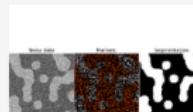


scikit-image  
image processing in python™

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## 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

[Download](#)

### Development

pre-0.11

[Download](#)[8+](#)

307

[Star](#)

522

### Links

[Issue tracker](#)[Mailing list](#)[Test results](#)

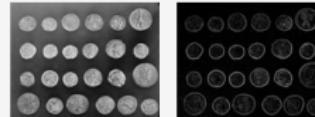
### Related Projects

[OpenCV](#)[Scikit-learn](#)[Mahotas](#)[SimpleCV](#)[Ilastik](#)

## 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()
```



If you find this project useful, please cite:

[\[BIBTeX\]](#)

Stefan van der Walt, Johannes L. Schönberger, Juan Nunez-Iglesias, François Boulogne, Joshua D. Warner, Neil Yager, Emmanuelle Gouillart, 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>

# Gallery of examples



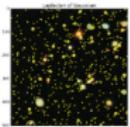
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## General examples

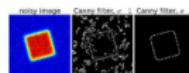
General-purpose and introductory examples for the scikit-image.



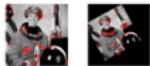
[Blob Detection](#)



[BRIEF binary descriptor](#)



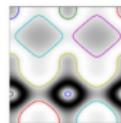
[Canny edge detector](#)



[CENSURE feature detector](#)



[Circular and Elliptical Hough Transforms](#)



[Contour finding](#)



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[Blob Detection](#)

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[Longer examples and demonstrations](#)

### Versions

[skimage dev](#)

[skimage 0.10.x](#)

[skimage 0.9.x](#)

[skimage 0.8.0](#)

[skimage 0.7.0](#)

[skimage 0.6](#)

[skimage 0.5](#)

[skimage 0.4](#)

[skimage 0.3](#)

# Getting started: finding documentation

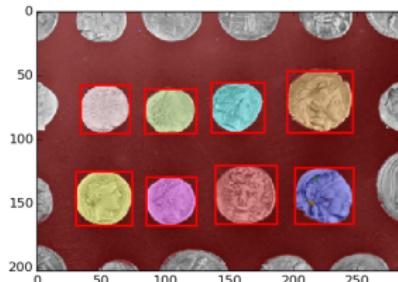


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## Label image regions

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)

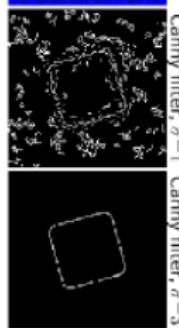
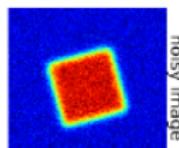
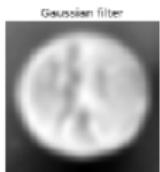
for region in regionprops(label_image):

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

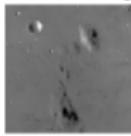
    # draw rectangle around segmented coins
    minc, minr, maxc, maxr = region.bbox
    rect = mpatches.Rectangle(minc, minr, maxc - minc, maxr - minr,
                             fill=False, edgecolor='red', linewidth=2)
    ax.add_patch(rect)

plt.show()
```

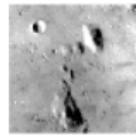
# Filtering: transforming image data



Low contrast image



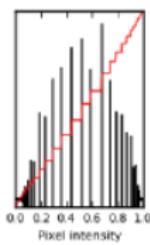
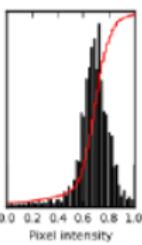
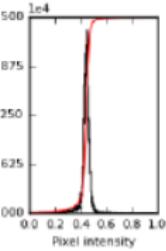
Contrast stretching



Histogram equalization



Adaptive equalization



denoising sobel  
equalize wiener  
**Median**  
Gaussian canny  
enhance\_contrast  
total\_variation

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

# Extracting features

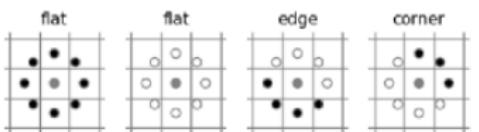
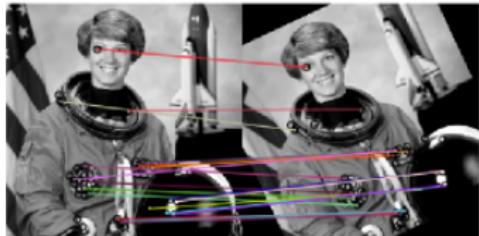
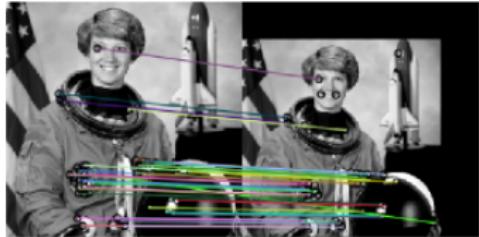
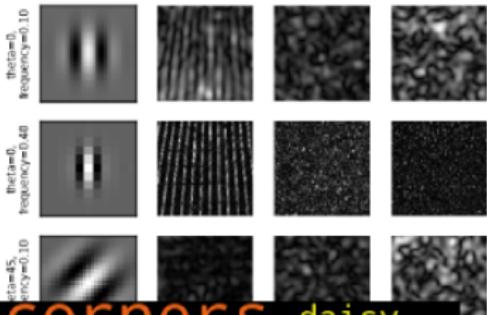


Image responses for Gabor filter kernels



corners  
local\_maxima  
Gabor  
canny  
Harris  
hog  
hough  
cooccurrence

`skimage.feature`, `skimage.filter`

# 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



# Geometrical transformations

`skimage.transform`

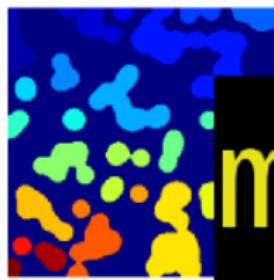
scale, zoom, rotate, swirl, warp, ...



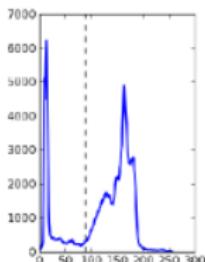
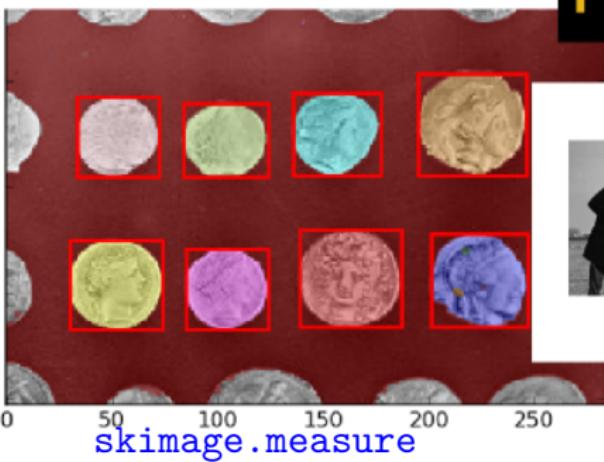
# Segmentation: labelling regions



# Measures on images

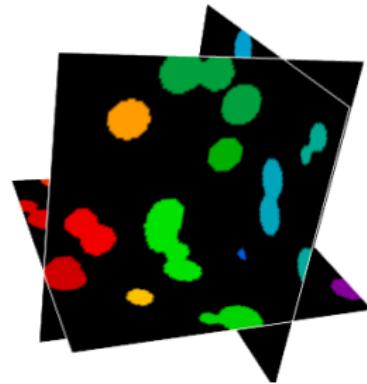
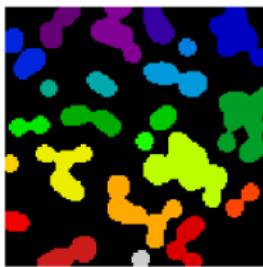
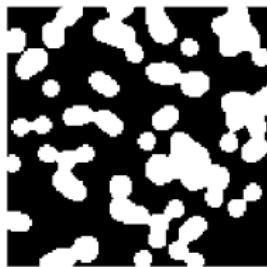


size  
label  
**measure**  
histogram  
regionprops



# Versatile use for 2D, 2D-RGB, 3D...

```
>>> from skimage import measure  
>>> labels_2d = measure.label(image_2d)  
>>> labels_3d = measure.label(image_3d)
```



Try it out! <http://scikit-image.org/>



**scikit-image**  
image processing in python

Feedback welcome

[github.com/scikit-image/scikit-image](https://github.com/scikit-image/scikit-image)

Please cite the paper ☺

scikit-image: image processing in Python

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François Boulogne<sup>4</sup>, Joshua D. Warner<sup>5</sup>, Neil Yager<sup>6</sup>, Emmanuelle Gouillart<sup>7</sup>, Tony Yu<sup>8</sup>,

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