

# Image Search Engine Resource Guide

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Adrian Rosebrock



## Image Search Engine: Resource Guide

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## Introduction

Hello! My name is Adrian Rosebrock from [PyImageSearch.com](http://PyImageSearch.com). Thanks for downloading this ***Image Search Engine Resource Guide***.

A little bit about myself: I'm an entrepreneur who has launched two successful image search engines: [ID My Pill](#), an iPhone app and API that identifies your prescription pills in the snap of a smartphone's camera, and [Chic Engine](#), a fashion search engine for the iPhone. Previously, my company ShiftyBits, LLC. has consulted with the National Cancer Institute to develop image processing and machine learning algorithms to automatically analyze breast histology images for cancer risk factors.

I have a Ph.D in computer science, with a focus in computer vision and machine learning, from the University of Maryland, Baltimore County where I spent three and a half years studying. I graduated in May of 2014.

I'm here to share the tips, tricks, and hacks I've learned in the past 8 years working in the startup field and building computer vision, machine learning, and image search engines.

After reading this guide, I would be interested to hear what you thought of it. Did you try any of the books? Did you look download any of the Python packages? Please send me an email and let me know at [adrian@pyimagesearch.com](mailto:adrian@pyimagesearch.com) or you can visit my website at [www.PyImageSearch.com](http://www.PyImageSearch.com) and leave a comment. I look forward to hearing from you soon!

-Adrian Rosebrock

## Books

Back when I was a Ph.D student, I've had to read a ton of books and papers. Some of these were reference books, some were very technical, and others simply gave a high level overview of computer vision. Having these books, whether in physical or PDF form, is invaluable -- I could quickly pull them open and get the information I needed.

While there aren't any substantial books on building image search engines, you can start by reading about *computer vision* - a larger topic that includes processing, analyzing, and understanding the content of images. Having a strong foundation of computer vision (or at least being familiar with the concept computer vision) will **dramatically** help you build image search engines of your own.

Of course, having an understanding of computer vision is not a requirement. If you are new to the field and don't have any experience, that's okay too. I like to create examples that are very hands on, that let you start building image search engines immediately, without getting lost in the details.

### My Books

I have written two books on computer vision and image processing. The first, *Practical Python and OpenCV* covers the basics of computer vision. It is a guaranteed, quick start guide to learning the fundamentals of computer vision.

The second, *Case Studies: Solving real-world problems with computer vision* applies the fundamentals of computer vision to solve problems such as face detection, object tracking, and keypoint matching using SIFT.

- To grab a copy of *Practical Python and OpenCV*, [just click here](#) and select the "Basic Bundle".
- To get my *Case Studies* eBook, [click here](#) and select the "Case Studies Bundle". This bundle also includes a copy of *Practical Python and OpenCV*!

*Note: I highly recommend the Case Studies and Premium Bundle if you're looking to not only master the fundamentals of computer vision, but apply your knowledge to solve actual real-world problems as well.*

### Beginner Books

I recommend these books for someone just starting out in the field of computer vision and image search engines:

- [Programming Computer Vision with Python: Tools and algorithms for analyzing images](#) by Jan Erik Solem

- [Practical Computer Vision with SimpleCV : The Simple Way to Make Technology See](#) by Kurt Demagd, Anthony Oliver, Nathan Oostendorp, and Katherine Scott
- [OpenCV Computer Vision with Python](#) by Joseph Howse
- [Learning OpenCV: Computer Vision with the OpenCV Library](#) by Gary Bradski and Adrian Kaehler
- [OpenCV 2 Computer Vision Application Programming Cookbook](#) by Robert Laganière
- [Mastering OpenCV with Practical Computer Vision Projects](#) by Daniel Lélis Baggio, Shervin Emami, David Millán Escrivá, Khvedchenia Ievgen, Jasonl Saragih, and Roy Shilkrot
- [SciPy and NumPy: An Overview for Developers](#) by Eli Bressert

## Textbooks

If you already have an understanding of computer vision, whether from reading previous works, taking a few courses in college, or simply working in the field already, then these books will help give you a deeper understanding of computer vision:

- [Computer Vision: A Modern Approach \(2nd Edition\)](#) by David A. Forsyth and Jean Ponce
- [Computer Vision](#) by Linda G. Shapiro and George C. Stockman
- [Computer Vision: Algorithms and Applications](#) by Richard Szeliski
- [Algorithms for Image Processing and Computer Vision](#) by J. R. Parker
- [Computer Vision: Models, Learning, and Inference](#) by Dr Simon J. D. Prince
- [Computer and Machine Vision, Fourth Edition: Theory, Algorithms, Practicalities](#) by E. R. Davies

## Conferences

If you have experience in the computer vision or Content Based Image Retrieval (CBIR) field, then you might want to consider submitting original research papers to the following conferences:

- [CVPR - Computer Vision and Pattern Recognition](#)
- [ICCV - International Conference on Computer Vision](#)
- [ECCV - European Conference on Computer Vision](#)
- [BMVC - British Machine Vision Conference](#)
- [ICIP - IEEE International Conference on Image Processing](#)

## Python Libraries

When I first became interested in computer vision and image search engines over eight years ago, I had no idea where to start. I didn't know which language to use, I didn't know which libraries to install, and the libraries I found I didn't know how to use. **I WISH** there had been a list like this one, detailing the best Python libraries to use for image processing, computer vision, and image search engines.

This list is by no means complete or exhaustive. It's just my favorite Python libraries that I use each and everyday for computer vision and image search engines. If you think that I've left an important one out, please leave me an email at [adrian@pyimagesearch.com](mailto:adrian@pyimagesearch.com).

### NumPy

NumPy is a library for the Python programming language that (among other things) provides support for large, multi-dimensional arrays. Why is that important? Using NumPy, we can express images as multi-dimensional arrays. Representing images as NumPy arrays is not only computational and resource efficient, but many other image processing and machine learning libraries use NumPy array representations as well. Furthermore, by using NumPy's built-in high-level mathematical functions, we can quickly perform numerical analysis on an image.

### SciPy

Going hand-in-hand with NumPy, we also have SciPy. SciPy adds further support for scientific and technical computing. One of my favorite sub-packages of SciPy is the [spatial package](#) which includes a vast amount of distance functions and a kd-tree implementation. Why are distance functions important? When we "describe" an image, we perform feature extraction. Normally after feature extraction an image is represented by a vector (a list) of numbers. In order to compare two images, we rely on distance functions, such as the Euclidean distance. To compare two arbitrary feature vectors, we simply compute the distance between their feature vectors. In the case of the Euclidean distance, the smaller the distance the more "similar" the two images are.

### matplotlib

Simply put, matplotlib is a plotting library. If you've ever used MATLAB before, you'll probably feel very comfortable in the matplotlib environment. When analyzing images, we'll make use of matplotlib, whether plotting the overall accuracy of search systems or simply viewing the image itself, matplotlib is a great tool to have in your toolbox.

### PIL and Pillow

These two packages are good and what they do: simple image manipulations, such as resizing, rotation, etc. If you need to do some quick and dirty image manipulations definitely check out PIL and Pillow, but if you're serious about learning about image processing, computer vision, and image search engines, I would *highly* recommend that you spend your time playing with OpenCV and SimpleCV instead.



## **OpenCV**

If NumPy's main goal is large, efficient, multi-dimensional array representations, then, by far, the main goal of OpenCV is real-time image processing. This library has been around since 1999, but it wasn't until the 2.0 release in 2009 did we see the incredible NumPy support. The library itself is written in C/C++, but Python bindings are provided when running the installer. OpenCV is hands down my favorite computer vision library, but it does have a learning curve. Be prepared to spend a fair amount of time learning the intricacies of the library and browsing the docs (which have gotten substantially better now that NumPy support has been added). If you are still testing the computer vision waters, you might want to check out the SimpleCV library mentioned below, which has a substantially smaller learning curve.

## **SimpleCV**

The goal of SimpleCV is to get you involved in image processing and computer vision as soon as possible. And they do a great job at it. The learning curve is substantially smaller than that of OpenCV, and as their tagline says, "it's computer vision made easy". That all said, because the learning curve is smaller, you don't have access to as many of the raw, powerful techniques supplied by OpenCV. If you're just testing the waters, definitely try this library out.

## **mahotas**

Mahotas, just as OpenCV and SimpleCV, rely on NumPy arrays. Much of the functionality implemented in Mahotas can be found in OpenCV and/or SimpleCV, but in some cases, the Mahotas interface is just easier to use, especially when it comes to their [features](#) package.

## **scikit-learn**

Alright, you got me, Scikit-learn isn't an image processing or computer vision library — it's a machine learning library. That said, you can't have advanced computer vision techniques without some sort of machine learning, whether it be clustering, vector quantization, classification models, etc. Scikit-learn also includes a handful of [image feature extraction functions](#) as well.

## **scikit-image**

Scikit-image is fantastic, but you have to know what you are doing to effectively use this library -- and I don't mean this in a "there is a steep learning curve" type of way. The learning curve is actually quite low, especially if you check out their gallery. The algorithms included in scikit-image (I would argue) follow closer to the state-of-the-art in computer vision. New algorithms right from academic papers can be found in scikit-image, but in order to (effectively) use these algorithms, you need to have developed some rigor and understanding in the computer vision field. If you already have some experience in computer vision and image processing, definitely check out scikit-image; otherwise, I would continue working with OpenCV and SimpleCV to start.

### **[ilastik](#)**

I'll be honest. I've never used ilastik. But through my experiences at computer vision conferences, I've met a fair amount of people who do, so I felt compelled to put it in this list. Ilastik is mainly for image segmentation and classification and is especially geared towards the scientific community.

### **[pprocess](#)**

Extracting features from images is inherently a parallelizable task. You can reduce the amount of time it takes to extract features from an entire dataset by using a multithreading/multitasking library. My favorite is pprocess, due to the simple nature I need it for, but you can use your favorite.

### **[h5py](#)**

The h5py library is the de-facto standard in Python to store large numerical datasets. The best part? It provides support for NumPy arrays. So, if you have a large dataset represented as a NumPy array, and it won't fit into memory, or if you want efficient, persistent storage of NumPy arrays, then h5py is the way to go. One of my favorite techniques is to store my extracted features in a h5py dataset and then apply scikit-learn's [MiniBatchKMeans](#) to cluster the features. The entire dataset never has to be entirely loaded off disk at once and the memory footprint is extremely small, even for thousands of feature vectors.

## Connect

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