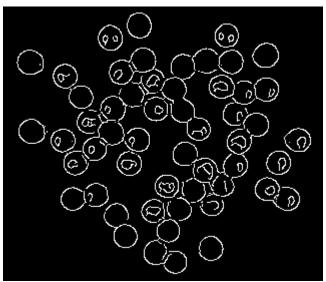
Many objects can be segmented from the background just using color. Working with multiple colored objects can be tricky to keep track of and quantify. This tutorial shows two possible solutions for segmenting M&M candy from the background and provides a summary count of what is found in the image.

## Technique #1

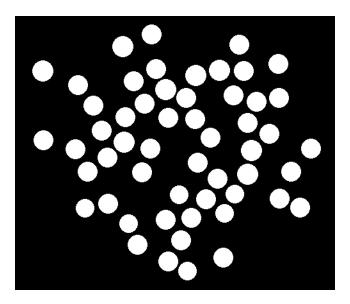
The first technique is to use the circular nature of the M&Ms to first segment them from the background. We do this by first running the Canny edge detector, remember blurring helps if you do that first, I also messed around with dilating and eroding a bit first, not sure if it's going to be in my final solution or not.



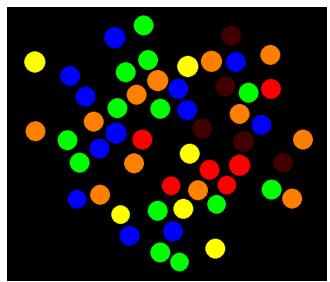
OpenCV has a circle function, returns circles that it can find in the image. We then can use the Circle function to identify circle, my picture of peanut M&Ms are oblong but so far I have been able to grab a few of those.

As I'm writing this I have written my code up to this point......I haven't dialed in all the M&Ms on the different images I'm using.....I have most, or a few, but not all yet. So the rest of this tutorial is taken from a tool that does this stuff for you, a Photoshop type tool.....no code to be written for the tool.

We next will annotate the location of each circle using a white disk 5 pixels smaller than the originally found circle. This ensure that circles are NOT overlapping. This ensures successive analysis will consider each M&M individually.



Now that we have a mask for each M&M we replace each disk by the color nearest to the mean color of all the pixels that exist in the same location in the original image. This performs a crucial step. When the replacement happens, the module also generates an array of the X,Y location of the colored circle AND includes a label that we create along with those coordinates. The label is important since each green M&M can have very different colors. In order to count all the 'green' M&M's we need to force green colored M&M's into a single label. The nearest color search does this. In other words a blob with mean of 0,245,33 and 4,246,2 which are both green in color will be labeled with 'green'.



Because we now have an array of colored entities and their labels we can use a basic script to loop through all the elements in the array in order to count the occurrence of each color.

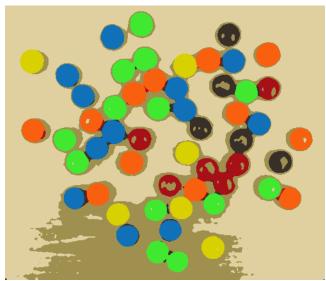
Then we can use OpenCV drawing functions to write the answer out on the bottom something like this:



## Technique #2

The first technique works well by utilizing the circle module to find the initial M&M's. This works well for circular objects but suppose we had other objects that could not be so easily discovered by their shape. In the second technique we assume the objects are not circular but instead JUST rely on color.

The first step is to reduce the actual number of colors in the image. Any image made up by 1000s of pixels will have very subtle color changes which are not exactly the same. By using the Reduce\_Colors (a module the tool has, but you don't, so you need to write the code to do it) we effectively reduce the image into blobs of similar colors. These blobs will never be perfect in terms of circular shapes but we have to start somewhere.



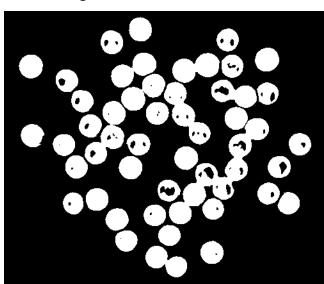
Now that the image has been reduced to a few colors, we want to remove those parts of the image that are not relevant to our analysis. This includes the tan background

(originally the wooden table) and parts of the image that make up the shadows of the M&M's. The Circle detection module in Technique #1 avoided this problem.

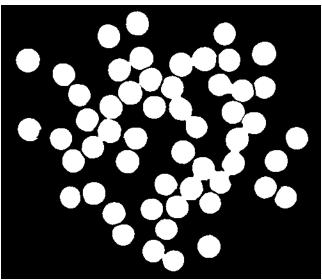
The best module for removing unwanted blobs is the Blob Filter module. This module is perhaps the largest module because it encompasses so many characteristics that can be filtered. Keep in mind that the blob filter will see each different colored group of pixels as an individual blob. The module reports that 202 blobs are found before the module runs. As we only want around 60 actual blobs we have to do a bit of work to remove the 140 that we don't want.

At this point we could also go back and use a circular shape as an attribute to remove all non-circular blobs, but we will stick with color and size to show more generic filtering. In our case we remove objects smaller than 215 pixels since our M&Ms are larger than that, remove objects larger than 5000 pixels (which eliminates the background table blob) and then remove any blob that doesn't have a strong color. Objects such as shadows will not have a strong hue and thus are removed. The problem, however, is that same filter ALSO removes the black M&M's ... since after all black is not a very strong color!

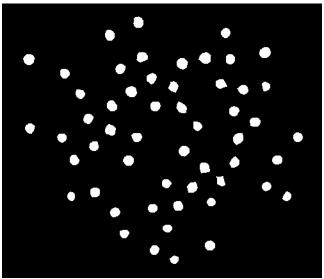
To remedy this issue we have a second filter within the BlobFilter module that selects blobs that are much darker than most in the image. This combined with a size filter successfully selects JUST the dark M&Ms. This result is then merged into the primary or Main filter to create the final image which includes both colored AND dark blobs.



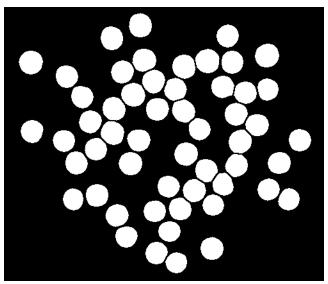
The problem now is that many of the blobs are still touching each other. In order to count them correctly we need to split them into individual blobs. Before we start splitting we need to fill in the empty holes in the blobs. This is done using the Fill module (OpenCV has a fill module)



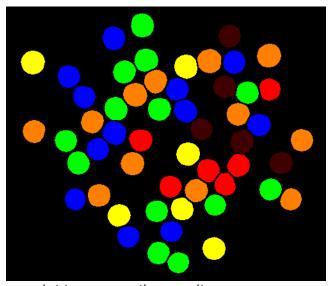
To split the blobs we separate them using the **Erode** module. By eroding the blobs by a 10 pixel perimeter which will break them into individual blobs.



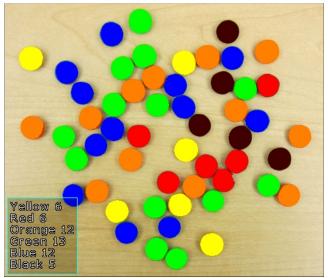
Because we want to preserve the approximate size of each M&Ms we then use the Dilate module to dilate each of the remaining blobs by 10 pixels BUT we ensure that no blobs will reconnect. This will then result in each blob being separated from each other.



We can now follow a similar process as Technique #1 by utilizing the Blob Colorize to color in each of the remaining blobs with their source mean color.



and then use the same script to sum up the results.



While the results are the same for both techniques, the pixels that determines a final M&M is a little different. You can see that the color technique does use more of the pixels to define an M&M (i.e. the final area is a bit larger) but as that was not a final measurement we were interested in either technique will work.

In terms of performance, the color technique #2 (while including more modules) is 40% faster than the circle detection technique #1 just for this specific image.

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## Advanced portion:

Put the colored m&ms on a piece of paper and move them in front of the camera like a conveyor belt and see if the count keeps working.

## The End