Advanced Lane Finding Project

The goals / steps of this project are the following:

- Compute the camera calibration matrix and distortion coefficients given a set of chessboard images.
- · Apply a distortion correction to raw images.
- Use color transforms, gradients, etc., to create a thresholded binary image.
- Apply a perspective transform to rectify binary image ("birds-eye view").
- Detect lane pixels and fit to find the lane boundary.
- Determine the curvature of the lane and vehicle position with respect to center.
- Warp the detected lane boundaries back onto the original image.
- Output visual display of the lane boundaries and numerical estimation of lane curvature and vehicle position.

Camera Calibration

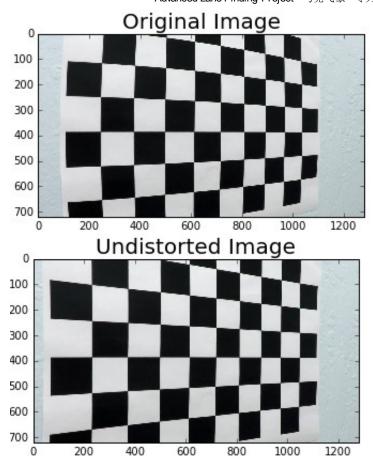
The code for this step is contained in the file called camera_calibration_using_chessboard_images.py.

1. Extract object points and image points for camera calibration.

I start by preparing "object points", which will be the (x, y, z) coordinates of the chessboard corners which size is setted to 9x6. Here I am assuming the chessboard is fixed on the (x, y) plane at z=0, such that the object points are the same for each calibration image. Thus, objp is just a replicated array of coordinates, and objpoints will be appended with a copy of it every time I successfully detect all chessboard corners in a test image. impoints will be appended with the (x, y) pixel position of each of the corners in the image plane with each successful chessboard detection.

I then used the output objpoints and impoints to compute the camera calibration and distortion coefficients using the

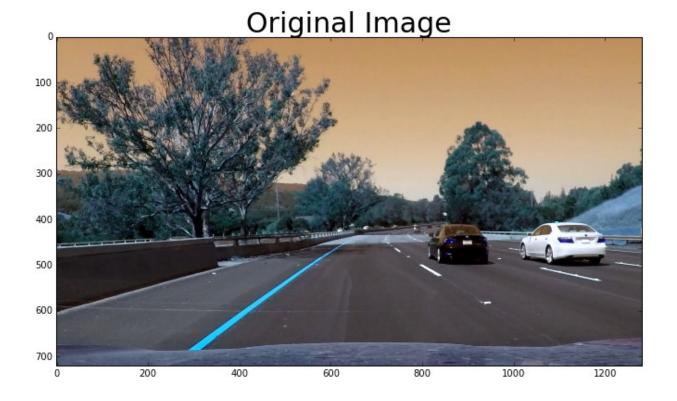
camera_calibration_using_chessboard_images.py function.lapplied this
distortion correction to the test image using the cv2.undistort() function and obtained
this result:

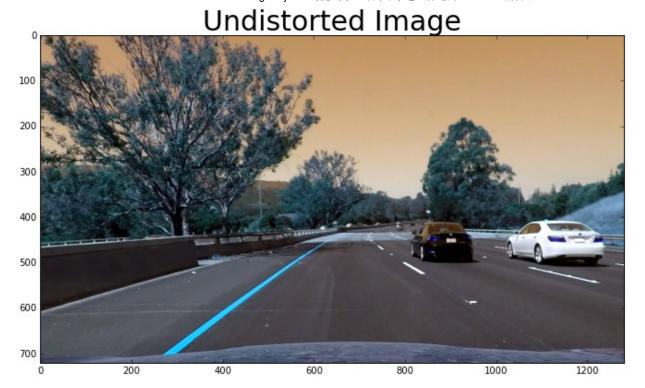


Pipeline (single images)

1. An example of a distortion-corrected image.

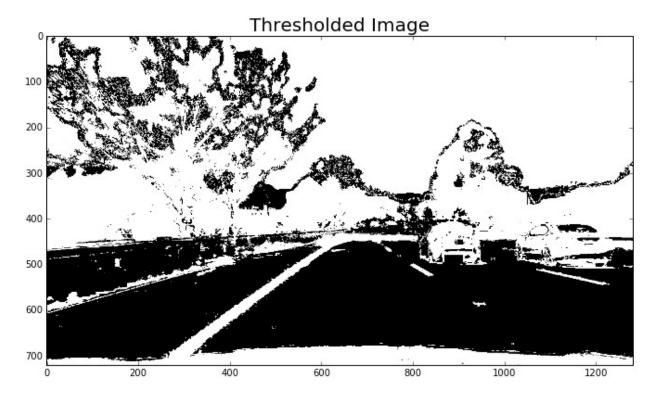
To demonstrate this step, I will describe how I apply the distortion correction to one of the test images like this one:





2. Use color transforms, gradients or other methods to create a thresholded binary image.

I used a combination of color ,sobel operator, magnitude of the gradient and direction of the gradient thresholds to generate a binary image (thresholding steps at lines 136 through 140 in instrument_function.py). Here's an example of my output for this step.



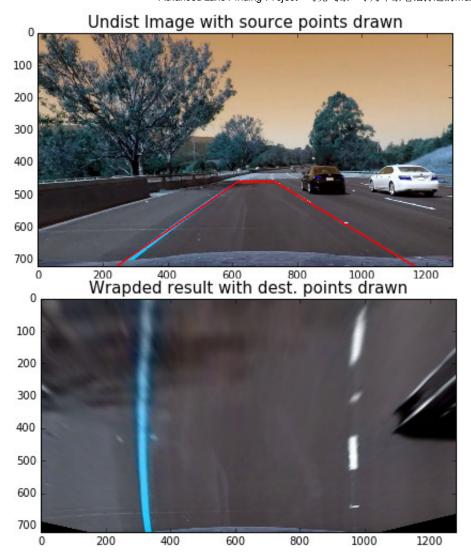
3. Performed a perspective transform to image.

After distorted, the code for my perspective transform also includes in file instrument_function.py, which appears in lines 155 through 193. The code in instrument_function.py takes as inputs an image
(./calibration_wide/test_undist.jpg), as well as source (src) and destination (dst) points. I chose the hardcode the source and destination points in the following manner:

This resulted in the following source and destination points:

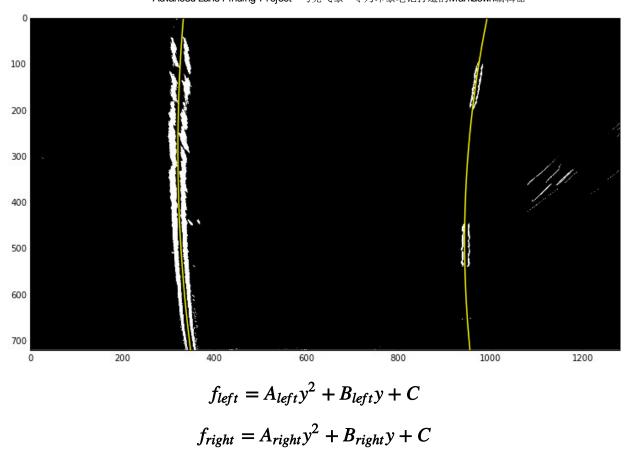
Source	Destination
585, 460.	320, 0
203.33, 720	320, 700
1121.66, 720	960, 700
700, 460	960, 0

I verified that my perspective transform was working as expected by drawing the src and dst points onto a test image and its warped counterpart to verify that the lines appear parallel in the warped image.



4. Identified lane-line pixels and fit their positions with a polynomial

Then I implemented sliding windows to find which "hot" pixels are associated with the lane lines in fuction called find_lines() in file Finding_the_Lines.py. Then I fit my lane lines with a 2nd order polynomial kinda like this:



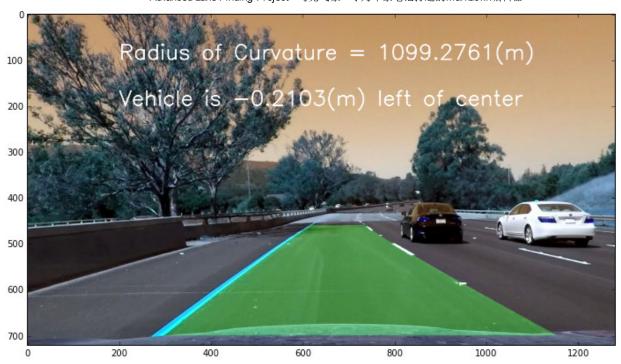
5. Calculated the radius of curvature of the lane and the position of the vehicle with respect to center.

I did this in lines 200 through 205 in my code in $instrument_function.py$, and called a function $meas_cur()$, the result is as follow:

```
left_curverad : 1099.27608597 (m) ,right_curverad : 806.926849431
(m)
Vehicle is -0.210319161451 (m) left of center
```

6. Processing result of image plotted back down onto the road

I implemented this step in lines 206 through 225 in my code in instrument_function.py. Here is an example of my result on a test image:



Pipeline (video)

I use Tracking.py to do the lane finding and here's a link to my video result

Discussion

In my premier implementation of my code, I find wobbly region appeared when the car dive in intensity sunlight where the lane line is unclear. To avoid the impact of sunlight, I tried some combination of thresholds method to generate a binary image as clear as possible. At last I chose the combination of color(with S channel), magnitude and sobel(in x orient) thresolds. Then, I creat a class Line to receive and process the characteristics of each line detection. I defined some functions to improve robust:

sanity_check() checking the detection result

- Checking that they have similar curvature or roughly parallel
- Checking that they are separated by approximately the right distance horizontally
 I chose the threshold of parallel refer to Horizontal Curvature of Highways without
 Superelevation, and width threshold of the road is set to >3.3 and < 4.5 considering
 some error in computation and estimation.

smoothing() smooth over the last n frames of video to obtain a cleaner result
If detection from some frame did not passed check of sanity_check() because of some
reasons. I average the n past results to estimate the current fit coefficients and if
successive detection lost exceed n times, reset and start searching from scratch using a
histogram and sliding window to re-establish measurement.

look_ahead_Filter

If the lane lines in one frame of video has been dectected and passed check, search within a window around the previous detection.