

# advanced\_lane\_finding

April 2, 2021

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In [1]: import numpy as np
import cv2
import glob
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

%matplotlib qt

In [78]: class CalibrateCamera:
    """
    CALIBRATE CAMERA
    This class calculates camera calibrates coefficients using a set of chessboard images
    """
    def __init__(self, image_dir=None):
        self.image_dir = image_dir or './camera_cal/calibration*.jpg'
        self.mtx = None # Camera calibration matrix
        self.dist = None # Camera distortion coefficient

        self._calibrate()

    def _calibrate(self):
        """
        CALIBRATE
        This function uses a set of chessboard calibration images to calculate the
        calibration matrix and distortion coefficients for the camera.
        """
        # Prepare 3D object points, like (0,0,0), (1,0,0), (2,0,0) ...., (6,5,0)
        objp = np.zeros((6*9,3), np.float32)
        objp[:, :2] = np.mgrid[0:9,0:6].T.reshape(-1, 2)

        # Initialize arrays to store object points and image points
        objpoints = [] # 3d points in real world space
        imgpoints = [] # 2d points in image plane

        # List of chessboard calibration test images
        test_images = glob.glob(self.image_dir)
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        # Find chessboard corners for each image
        for filename in test_images:
            image = cv2.imread(filename)
            gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)

            # Find the chessboard corners
            ret, corners = cv2.findChessboardCorners(gray, (9,6), None)

            # If corners successfully found, append object points and image points
            if ret == True:
                objpoints.append(objp)
                imgpoints.append(corners)

        # Find calibration matrix and distortion coefficient using cv2.calibrateCamera
        _, mtx, dist, _, _ = cv2.calibrateCamera(objpoints, imgpoints, gray.shape[::-1])

        self.mtx = mtx
        self.dist = dist

    return self

class FindLane:
    """
    FIND LANE
    This class takes in an input image and camera calibration data) to find and
    draw detected lane lines back onto the image, with included road and vehicle stats.
    """
    def __init__(self, image, mtx, dist):
        # Input image
        self.image = image
        self.height = image.shape[0]
        self.width = image.shape[1]

        # Camera calibration
        self.mtx = mtx    # Calibration matrix
        self.dist = dist  # Distortion coefficient

        # Perspective transform matrix
        self.M = np.matrix([])

        # Sobel transform parameters
        self.s_thresh = (120, 255) # S channel threshold
        self.sx_thresh = (10, 150) # x-distance threshold

        # Pixels to meters conversion
        self.y_meters_per_px = 30/720 # meters per pixel in y dimension
        self.x_meters_per_px = 3.7/700 # meters per pixel in x dimension

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        # Initialize empty images
        self.undistorted_image = np.array([])
        self.binary_image = np.array([])
        self.birds_eye_image = np.array([])
        self.output_image = np.array([])

        # Execute the finding algorithm and return image
        self._find()

def _find(self):
    """
    FIND LANE
    Warp the detected lane boundaries onto original input image,
    and display the road curvature and vehicle distance from center.
    """
    # Find left and right lane line pixels
    if not self.output_image.any():
        leftx, lefty, rightx, righty = self._find_lane_line_pixels()

        # Fit polynomials to both left and right lane lines
        left_fit, left_fitx, fity = self._fit_polynomial(leftx, lefty, 'pixels')
        right_fit, right_fitx, fity = self._fit_polynomial(rightx, righty, 'pixels')

        # Recast the x and y points into usable format for cv2.fillPoly()
        pts_left = np.array([np.transpose(np.vstack([left_fitx, fity]))])
        pts_right = np.array([np.flipud(np.transpose(np.vstack([right_fitx, fity])))]
        pts = np.hstack((pts_left, pts_right))

        # Create a empty image and draw the lane onto it
        lane_image = np.zeros_like(self._undistorted_image()).astype(np.uint8)
        cv2.fillPoly(lane_image, np.int_([pts]), (10, 250, 10))

        # Warp lane image back to original perspective using inverse perspective map
        warped = cv2.warpPerspective(lane_image, self.M.I, (self.width, self.height))

        # Combine the warped result with the original input image
        combined = cv2.addWeighted(self.image, 1, warped, 0.3, 0)

        # Add road curvature and vehicle center stats to combined image
        self.output_image = self._render_stats(combined, left_fit, right_fit)

    #return self.output_image
    return self

def _undistorted_image(self):
    """
    UNDISTORTED IMAGE

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Returns an undistorted image using calculated camera calibration matrix and distortion coefficients.
'''
if not self.undistorted_image.any():
    # Use camera calibration parameters to undistort image
    self.undistorted_image = cv2.undistort(self.image, self.mtx, self.dist, None,

return self.undistorted_image

def _binary_image(self):
    '''
    BINARY IMAGE
    Convert image into a binary representation using thresholding.
    a. Use a Sobel X transformation on "L" channel to find vertical lines.
    b. Use thresholding on the "S" channel to pick out lane lines.
    '''
    if not self.binary_image.any():
        image = np.copy(self._undistorted_image())

        # Convert to HLS color space and separate into L and S channels
        hls_image = cv2.cvtColor(image, cv2.COLOR_RGB2HLS)
        l_channel = hls_image[:, :, 1]
        s_channel = hls_image[:, :, 2]

        # Apply Sobel to l_channel
        sobelx = cv2.Sobel(l_channel, cv2.CV_64F, 1, 0) # Take the derivative in the x direction
        abs_sobelx = np.absolute(sobelx) # Absolute x derivative to accentuate lines
        scaled_sobel = np.uint8(255*abs_sobelx / np.max(abs_sobelx))

        # Create composite binary image using threshold parameters
        binary = np.zeros_like(scaled_sobel)
        binary[(scaled_sobel >= self.sx_thresh[0]) & (scaled_sobel <= self.sx_thresh[1])] = 1
        binary[(s_channel >= self.s_thresh[0]) & (s_channel <= self.s_thresh[1])] = 1

        self.binary_image = binary

    return self.binary_image

def _birds_eye_image(self):
    '''
    BIRDS EYE IMAGE
    Applies a perspective transform to create a "birds-eye view" of the road.
    '''
    if not self.birds_eye_image.any():
        # Define source transform points
        top_left = (596, 450)
        top_right = (685, 450)
        bottom_right = (1070, 700)

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        bottom_left = (245, 700)
        src = np.float32([top_left, top_right, bottom_right, bottom_left])

        # Define output destination points
        offset      = 300
        top_left     = (offset, 0)
        top_right    = (self.width-offset, 0)
        bottom_right = (self.width-offset, self.height)
        bottom_left  = (offset, self.height)
        dst = np.float32([top_left, top_right, bottom_right, bottom_left])

        # Calculate the perspective transform matrix
        self.M = np.matrix(cv2.getPerspectiveTransform(src, dst))

        # Generate warped, birds-eye-view image using perspective transform matrix
        self.birds_eye_image = cv2.warpPerspective(self._binary_image(), self.M,
                                                    (self.width, self.height))

    return self.birds_eye_image

def _find_lane_line_pixels(self):
    '''
    FIND LANE PIXELS
    Calculate pixel coords for lane lines using sliding window.
    '''
    image = self._birds_eye_image()

    # Grab only the bottom half of the image
    # Lane lines are likely to be mostly vertical nearest to the car
    bottom_half = image[self.height//2:, :]

    # Sum across image pixels vertically
    # The highest areas of vertical lines should be larger values
    histogram = np.sum(bottom_half, axis=0)

    # Find left and right x-coord to use as the starting point in our search
    # by slicing the histogram and calculating the max value for each lane.
    mid = np.int(self.width//2)
    leftx_current = np.argmax(histogram[:mid]) # Slice left
    rightx_current = np.argmax(histogram[mid:]) + mid # Slice right

    # Search window parameters
    nwindows = 9 # The number of sliding windows
    margin = 100 # Width of the windows +/- margin
    minpix = 50 # Minimum number of pixels found to re-center window

    # Set height of search window
    window_height = np.int(self.height//nwindows)

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# Identify the x and y positions of all nonzero (i.e. activated) pixels in the
nonzero = image.nonzero()
ys = np.array(nonzero[0])
xs = np.array(nonzero[1])

# Initialize lists for left and right lane pixel indices
left_lane_inds = []
right_lane_inds = []

# Step through the windows one by one
for window in range(nwindows):
    # Set window height boudaries and window left/right width boudaries
    win_y_lo = self.height - (window+1) * window_height
    win_y_hi = self.height - window * window_height
    win_xleft_lo = leftx_current - margin
    win_xleft_hi = leftx_current + margin
    win_xright_lo = rightx_current - margin
    win_xright_hi = rightx_current + margin

    # Identify nonzero pixels inside window boundaries
    y_lo = ys >= win_y_lo
    y_hi = ys < win_y_hi
    x_left_lo = xs >= win_xleft_lo
    x_left_hi = xs < win_xleft_hi
    x_right_lo = xs >= win_xright_lo
    x_right_hi = xs < win_xright_hi

    good_left_inds = (y_lo & y_hi & x_left_lo & x_left_hi).nonzero()[0]
    good_right_inds = (y_lo & y_hi & x_right_lo & x_right_hi).nonzero()[0]

    left_lane_inds.append(good_left_inds)
    right_lane_inds.append(good_right_inds)

    # If you found > minpix pixels, recenter next window on their mean position
    if len(good_left_inds) > minpix:
        leftx_current = np.int(np.mean(xs[good_left_inds]))
    if len(good_right_inds) > minpix:
        rightx_current = np.int(np.mean(xs[good_right_inds]))

# Concatenate the arrays of indices (previously was a list of lists of pixels)
left_lane_inds = np.concatenate(left_lane_inds)
right_lane_inds = np.concatenate(right_lane_inds)

# Extract left and right pixel positions
leftx = xs[left_lane_inds]
lefty = ys[left_lane_inds]
rightx = xs[right_lane_inds]

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        righty = ys[right_lane_inds]

    return leftx, lefty, rightx, righty

def _find_lane_line_pixels_from_prior(self, prior_left_fit, prior_right_fit):
    '''
    FIND LANE PIXELS FROM PRIOR
    Calculate pixel coords for lane lines based on prior video frame.
    '''
    image = self._birds_eye_image()

    # Width of margin to search around the prior polynomial
    margin = 100

    # Find activated pixels
    nonzero = image.nonzero()
    ys = np.array(nonzero[0])
    xs = np.array(nonzero[1])

    # Set the area of search based on activated x-values
    a, b, c = prior_left_fit
    x_left_lo = xs > a*(ys**2) + b*ys + c - margin
    x_left_hi = xs < a*(ys**2) + b*ys + c + margin
    left_lane_inds = (x_left_lo & x_left_hi)

    a, b, c = prior_right_fit
    x_right_lo = xs > a*(ys**2) + b*ys + c - margin
    x_right_hi = xs < a*(ys**2) + b*ys + c + margin
    right_lane_inds = (x_right_lo & x_right_hi)

    # Extract left and right pixel positions
    leftx = xs[left_lane_inds]
    lefty = ys[left_lane_inds]
    rightx = xs[right_lane_inds]
    righty = ys[right_lane_inds]

    return leftx, lefty, rightx, righty

def _fit_polynomial(self, xs, ys, scale='meters'):
    '''
    FIT POLYNOMIAL
    Fit a second order polynomial to a set of x and y coords.
    '''
    if scale == 'meters':
        sy = self.y_meters_per_px
        sx = self.x_meters_per_px
    else:
        sy = 1

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        sx = 1

        # Use polyfit to return coefficients a, b, c of  $x = ay^2 + by + c$ 
        # Use meters-per-pixel conversion to return real-world values
        a, b, c = np.polyfit(ys*sy, xs*sx, 2)

        # Generate sequential y coords like [0, 1, 2, 3... height-1]
        ys = np.linspace(0, self.height-1, self.height)

        # Use fit to calculate x coords
        xs = a*ys**2 + b*ys + c

        # Return coefficients and poly coords
        return (a, b, c), xs, ys

def _calculate_stats(self, left_fit, right_fit):
    '''
    CALCULATE STATS
    Calculate the road radius of curvature and vehicle distance from center of lane
    a. Calculates the curvature of the polynomial functions in pixels, convert to m
    b. Determine the vehicle position with respect to center in pixels, convert to m
    '''
    y = self.height
    mx = self.x_meters_per_px
    my = self.y_meters_per_px

    # Calculate radius of curvature for left lane
    a = left_fit[0]
    b = left_fit[1]
    c = left_fit[2]
    #left_curverad = ((1 + (2*a*y*my+b)**2)**(3/2)) / np.abs(2*a)
    left_curverad = mx / (my**2)*a*(y**2) + (mx/my)*b*y + c

    # Calculate x-coord of left lane
    left_x = a*y**2 + b*y + c

    # Calculate radius of curvature for right lane
    a = right_fit[0]
    b = right_fit[1]
    c = right_fit[2]
    #right_curverad = ((1 + (2*a*7*my+b)**2)**(3/2)) / np.abs(2*a)
    right_curverad = mx / (my**2)*a*(y**2) + (mx/my)*b*y + c

    # Calculate x-coord of right lane
    right_x = a*y**2 + b*y + c

    # Average the left and right lane curvatures
    curvature = int((left_curverad + right_curverad) / 2)

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        # Find the vehicle distance offset from center of lane
        midpoint = (right_x + left_x) / 2
        distance = round(mx * ((self.width/2) - midpoint), 2)
        position = 'right' if distance > 0 else 'left'

        return curvature, abs(distance), position

def _render_stats(self, image, left_fit, right_fit):
    '''
    RENDER STATS
    Use cv2.putText to include stats on final image.
    '''
    radius, center, position = self._calculate_stats(left_fit, right_fit)

    radius_text = f"Radius of Curvature is {radius}m"
    center_text = f"Vehicle is {center}m {position} of center"

    for i, text in enumerate([radius_text, center_text]):
        pos = (70, 70 + (50*i))
        font = cv2.FONT_HERSHEY_TRIPLEX
        image = cv2.putText(image, text, pos, font, 1.5, (245, 245, 245), 2)

    return image

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In [90]: # View output image
         i = 1
         # #filename = f"./test_images/straight_lines{i}.jpg"
         filename = f"./test_images/test{i}.jpg"
         image = cv2.imread(filename)

         cc = CalibrateCamera()
         fl = FindLane(image, cc.mtx, cc.dist)
         output = fl.output_image

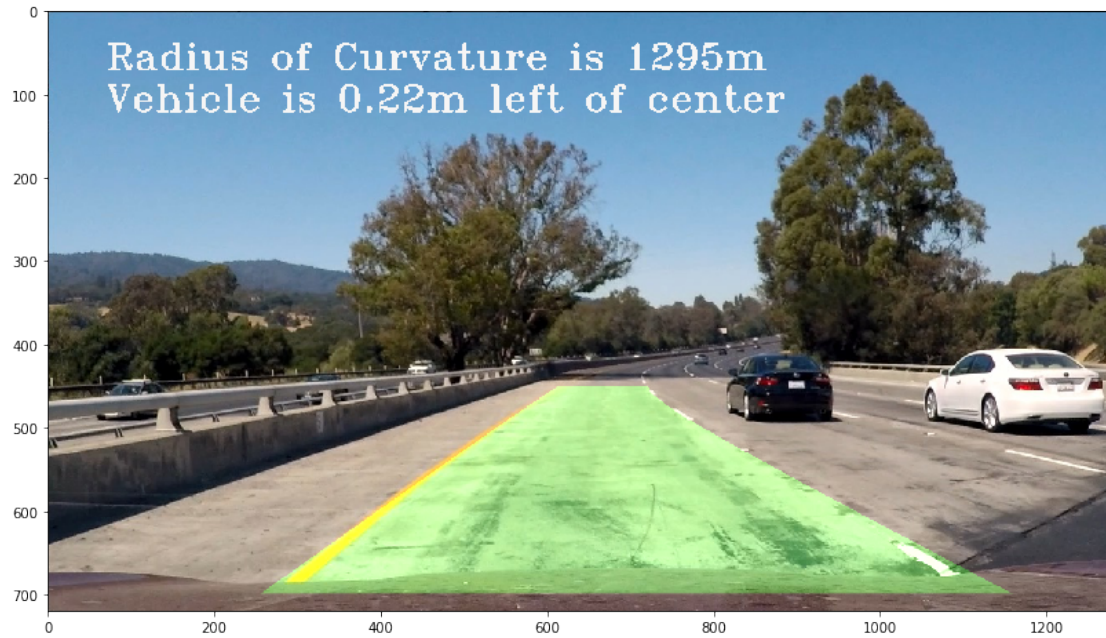
         output = cv2.cvtColor(output, cv2.COLOR_BGR2RGB)
         plt.figure(figsize = (18, 8)) # View output images larger
         plt.imshow(image_to_display)

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Out[90]: <matplotlib.image.AxesImage at 0x7fd299b91a20>

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In [92]: # Save test images
import os
from PIL import Image

for name in os.listdir('test_images'):
    if not name == '.ipynb_checkpoints':
        input = mpimg.imread(f"test_images/{name}")

        fl = FindLane(input, cc.mtx, cc.dist)
        image = fl.output_image

        output = Image.fromarray(image)
        output.save(f"output_images/final_output_{name}")
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In [89]: # Save test videos
from moviepy.editor import VideoFileClip
from IPython.display import HTML

cc = CalibrateCamera()

def process_image(image):
    fl = FindLane(image, cc.mtx, cc.dist)
    return fl.output_image

output = 'output_videos/project_video.mp4'
clip1 = VideoFileClip('project_video.mp4')
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final = clip1.fl_image(process_image)

%time final.write_videofile(output, audio=False)

[MoviePy] >>> Building video output_videos/challenge_video.mp4
[MoviePy] Writing video output_videos/challenge_video.mp4

100%|| 485/485 [01:25<00:00, 5.54it/s]

[MoviePy] Done.
[MoviePy] >>> Video ready: output_videos/challenge_video.mp4

CPU times: user 40.8 s, sys: 1.1 s, total: 41.9 s
Wall time: 1min 28s
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