Steering Wheel State Warning

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Problem

Place weighty objects on the steering wheel, which self driving car misunderstand users is grabbing the steering wheel. This can cause terrible accident when driver faces corner case.



Proposal

What about using cameras to detect whether driver is grabbing the steering wheel or not?



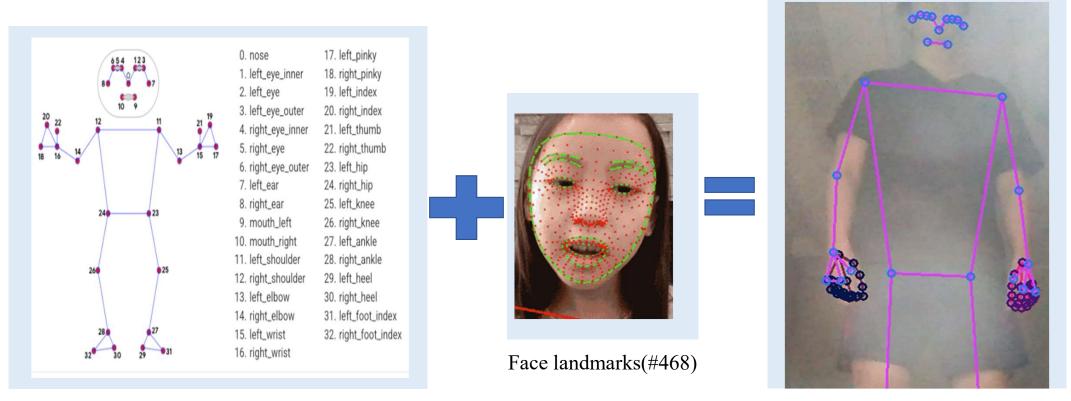
Driver grabbing the steering wheel



Driver not grabbing the steering wheel

Mediapipe Solution

Utilize Mediapipe Pose Solution to use landmarks as datasets.

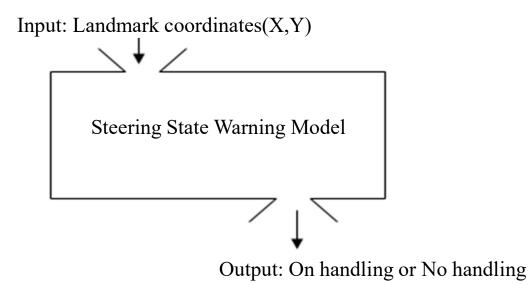


Pose landmarks (# 33)

Pose+Face landmarks(#501)

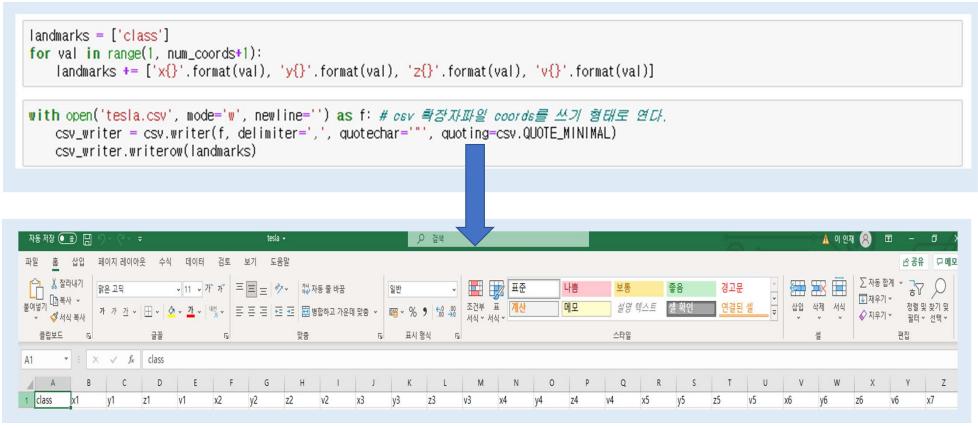
Define Machine-learning Model

Define classifier machine-learning model which gets real time coordinates of driver's landmark as input and class name as output whether driver is grabbing the steering wheel or not.



Collecting Dataset(Make CSV File)

Define coordinates of landmark as x,y,z and open tesla.csv file as write mode.



When open the made tesla.csv file you can See x1,y1 to x501 y501 which means landmarks coordinates

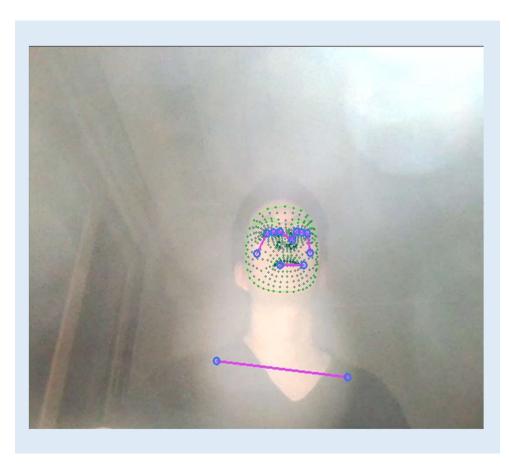
Collect Dataset(No Handling Code)

Collecting No Handling Dataset Code

```
class_name = "no handling" #classname 설정
cap = cv2.VideoCapture(0)
# Initiate holistic model
with mp_holistic.Holistic(min_detection_confidence=0.5, min_tracking_confidence=0.5) as holistic:
   while cap.isOpened():
       ret, frame = cap.read()
       # webcam을 처음 읽으면 bgr로 인식이 되므로 mediapipe를 사용하려면 rgb순으로 바꿔줘야 한다.
        image = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
        image.flags.writeable = False
       # Make Detections
       results = holistic.process(image)
       # rendering을 위해 rgb를 bgr로 바꿔준다.
       image.flags.writeable = True #detection 후에 True로 바꿔준다.
       image = cv2.cvtColor(image, cv2.COLOR_RGB2BGR)
       # 1. 얼굴
       mp_drawing.draw_landmarks(image, results.face_landmarks, mp_holistic.FACE_CONNECTIONS,
                               mp_drawing.DrawingSpec(color=(80,110,10), thickness=1, circle_radius=1),
                               mp_drawing.DrawingSpec(color=(80,256,121), thickness=1, circle_radius=1)
       # 2. 오른손
       mp_drawing.draw_landmarks(image, results.right_hand_landmarks, mp_holistic.HAND_CONNECTIONS,
                               mp_drawing.DrawingSpec(color=(80,22,10), thickness=2, circle_radius=4),
                               mp_drawing.DrawingSpec(color=(80,44,121), thickness=2, circle_radius=2)
       #3 84
       mp_drawing.draw_landmarks(image, results.left_hand_landmarks, mp_holistic.HAND_CONNECTIONS,
                               mp_drawing.DrawingSpec(color=(121,22,76), thickness=2, circle_radius=4),
                               mp_drawing.DrawingSpec(color=(121,44,250), thickness=2, circle_radius=2)
       # 4. Pose Detections
       mp_drawing.draw_landmarks(image, results.pose_landmarks, mp_holistic.POSE_CONNECTIONS,
                               mp_drawing.DrawingSpec(color=(245,117,66), thickness=2, circle_radius=4),
                               mp_drawing.DrawingSpec(color=(245,66,230), thickness=2, circle_radius=2)
```

```
# Export coordinates
          # pose landmark를 추출 후 numpy array 형태로 변환해준다. 이 때 flatten을 사용해 1차원으로 만들어준다.
          pose = results.pose_landmarks.landmark
          pose_row = list(np.array([[landmark.x, landmark.y, landmark.z, landmark.visibility] for landmark in pose]).flatten())
           # Extract Face landmarks
           face = results.face landmarks.landmark
           face_row = list(np.array([[landmark.x, landmark.y, landmark.z, landmark.visibility] for landmark in face]).flatten())
           # pose 열과 face 열의 정보들을 이어준다.
          row = pose_row+face_row
           # 위에서 정의한 class name을 입력해준다.
          row.insert(0, class name)
          # OSV파일로 좌표값 추출
          with open('tesla.csv', mode='a', newline='') as f:# 이미 만들어진 csv파일에 새로운 데이터를 입력하는 형태로 열어준다.
              csv_writer = csv.writer(f, delimiter=',', quotechar='"', quoting=csv.QUOTE_MINIMAL)
              csv writer.writerow(row)
       except:
           pass
       cv2.imshow('Raw Webcam Feed', image)
       if cv2.waitKey(10) & OxFF == ord('q'):
          break
cap.release()
cv2.destroyAllWindows()
```

Collecting Dataset(No handling)



- 1. Run the code then collect no grabbing steering wheel datasets using web cam.
- 2. You should change the angle of face and upper body to get accurate machine-learning model.

Collect Dataset(On Handling Code)

Collecting On Handling Dataset Code

```
class_name = "no handling" #classname 설정
cap = cv2.VideoCapture(0)
# Initiate holistic model
with mp_holistic.Holistic(min_detection_confidence=0.5, min_tracking_confidence=0.5) as holistic:
   while cap.isOpened():
       ret, frame = cap.read()
       # webcam을 처음 읽으면 bgr로 인식이 되므로 mediapipe를 사용하려면 rgb순으로 바꿔줘야 한다.
        image = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
        image.flags.writeable = False
       # Make Detections
       results = holistic.process(image)
       # rendering을 위해 rgb를 bgr로 바꿔준다.
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                               mp_drawing.DrawingSpec(color=(80,110,10), thickness=1, circle_radius=1),
                               mp_drawing.DrawingSpec(color=(80,256,121), thickness=1, circle_radius=1)
       # 2. 오른손
       mp_drawing.draw_landmarks(image, results.right_hand_landmarks, mp_holistic.HAND_CONNECTIONS,
                               mp_drawing.DrawingSpec(color=(80,22,10), thickness=2, circle_radius=4),
                               mp_drawing.DrawingSpec(color=(80,44,121), thickness=2, circle_radius=2)
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          # OSV파일로 좌표값 추출
           with open('tesla.csv', mode='a', newline='') as f:# 이미 만들어진 csv파일에 새로운 데이터를 입력하는 형태로 열어준다.
              csv_writer = csv.writer(f, delimiter=',', quotechar='"', quoting=csv.QUOTE_MINIMAL)
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       except:
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       cv2.imshow('Raw Webcam Feed', image)
       if cv2.waitKey(10) & OxFF == ord('q'):
          break
cap.release()
cv2.destroyAllWindows()
```

Collecting Dataset(On Handling)



- 1. Run the code then collect grabbing steering wheel datasets using web cam.
- 2. You should change the angle of face and location of arms(one arm grabbing steering wheel, two arm grabbing steering wheel) to get accurate machine-learning model.

Collected Dataset

no handlii	0.595236	0.576571	-0.93765	1	0.618763	0.536841	-0.85181	1	0.635369	0.53954	-0.85182	0.99996	0.652407	0.542128	-0.85201	1	0.569141	0.532694	-0.85725	1	0.551423	0.532158	-0.85777	0.999966	0.5334
no handlii	0.58438	0.573156	-0.93565	1	0.60835	0.532458	-0.84604	1	0.625217	0.534966	-0.84608	0.999976	0.642504	0.53732	-0.8463	1	0.558337	0.529472	-0.85306	1	0.540731	0.529387	-0.85355	0.999974	0.52292
7 no handlir	0.574261	0.572546	-0.92309	1	0.600984	0.530829	-0.8387	1	0.617996	0.533453	-0.83888	0.999944	0.635432	0.535919	-0.83927	1	0.551778	0.527324	-0.84683	1	0.534588	0.526897	-0.84731	0.999927	0.517
no handlii	0.563843	0.574177	-0.74441	1	0.590342	0.532708	-0.67104	1	0.607803	0.53499	-0.67118	0.999977	0.625651	0.537012	-0.67149	1	0.540712	0.529619	-0.66765	1	0.523671	0.529191	-0.66811	0.999959	0.5064
9 no handlii	0.570083	0.572127	-0.8168	1	0.592938	0.530205	-0.734	1	0.609353	0.532127	-0.73411	0.999992	0.626171	0.533803	-0.73436	1	0.544564	0.528447	-0.73875	1	0.527514	0.528557	-0.73919	0.999987	0.510
o no handlii	0.569993	0.570542	-0.7804	1	0.592165	0.528869	-0.696	1	0.60848	0.530522	-0.69617	0.999994	0.625188	0.531925	-0.69646	1	0.543756	0.527923	-0.70217	1	0.526701	0.528299	-0.70259	0.999989	0.5094
no handlii	0.572754	0.568344	-0.76409	1	0.594994	0.526259	-0.68082	1	0.611422	0.527803	-0.68095	0.999993	0.62824	0.529105	-0.68114	1	0.546112	0.52574	-0.68608	1	0.528865	0.526358	-0.68649	0.999987	0.5114
2 no handlir				1	0.595733	0.525359	-0.64505	1	0.612242	0.526975	-0.64517	0.999991	0.629137	0.528361	-0.64533	1	0.546621	0.524719	-0.65197	1	0.529281	0.525338	-0.65236	0.999984	0.5117
13 no handlii	0.572703	0.565885	-0.657	1	0.596686	0.523700	-0 57738	1	0.613302	0.525561	-0 5775	0 000000	0.630313	0.527000	-0 5777	1	0.547515	0.522866	-0.5852	1	0.53018	0.523478	-0.58566	0 0000021	0.5126
no handlii	0.577269	0.565328	-0.68016	1	0.598697	0.522609	-0.59999	1	0.615084	0.524496	-0.60012	0.999981	0.631851	0.526153	-0.60035	1	0.549486	0.521313	-0.60753	1	0.531829	0.521808	-0.60799	0.99997	0.5140
no handlir	0.577788	0.565799	-0.73208	1	0.598793	0.523236	-0.64678	1	0.615187	0.525129	-0.64687	0.999995	0.63196	0.526793	-0.64701	1	0.549571	0.521981	-0.65618	1	0.532021	0.522518	-0.65664	0.999991	0.5143
no handlii	0.582799	0.561642	-0.873	1	0.60252	0.520399	-0.7766	1	0.618848	0.522375	-0.77667	0.999983	0.635556	0.524134	-0.77678	1	0.552584	0.519146	-0.78888	1	0.534691	0.519788	-0.78928	0.999973	0.5166
no handli	0.583142	0.55484	-0.73526	1	0.602776	0.514548	-0.64845	1	0.619229	0.516874	-0.64853	0.999977	0.636036	0.518976	-0.64873	1	0.552589	0.512581	-0.66273	1	0.534721	0.513044	-0.66321	0.999967	0.5167
no handlii	0.586015	0.55462	-0.87288	1	0.603401	0.514242	-0.78427	1	0.619389	0.516432	-0.78432	0.999953	0.636145	0.518426	-0.78447	1	0.554296	0.512378	-0.80129	1	0.536515	0.512829	-0.80165	0.999938	0.5185
on handli	0.565637	0.629227	-0.21332	1	0.583925	0.599419	-0.13704	1	0.59735	0.601554	-0.13691	1	0.611019	0.603371	-0.13655	1	0.543591	0.59893	-0.1529	1	0.529386	0.599747	-0.15331	1	0.5152
on handli	0.566504	0.627656	-0.14485	1	0.587023	0.601657	-0.06727	1	0.600617	0.604934	-0.06726	1	0.614473	0.607847	-0.06713	1	0.546451	0.598178	-0.07784	1	0.532408	0.598184	-0.07827	1	0.5183

After collecting dataset, You can see each row's first column contains class names (on handling or no handling), and rests of column contains coordinates of driver's landmarks.

Train

('logisticregression', LogisticRegression())]),

('randomforestclassifier', RandomForestClassifier())]),

('gradientboostingclassifier', GradientBoostingClassifier())])}

('ridgeclassifier', RidgeClassifier())]),

'rc': Pipeline(steps=[('standardscaler', StandardScaler()),

'rf': Pipeline(steps=[('standardscaler', StandardScaler()),

'gb': Pipeline(steps=[('standardscaler', StandardScaler()),

❖ Train Code # 수집한 dataset을 train dataset과 test set를 7:3으로 random하게 나눈다. Split dataset(train: test=7:3) X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=1234) #make_pipeline(대신러님 모델에 사용할 파이프라인 구축하는 모듈),StandardScaler(dataset을 표준화 하는 모듈) import from sklearn.pipeline import make_pipeline from sklearn.preprocessing import StandardScaler from sklearn.linear_model import LogisticRegression, RidgeClassifier from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier #위에서 점의한 pipeline에 사용할 모델들 구축(좋은 결과를 내기 위해 여러가지 모델 사용) pipelines = { 'Ir':make_pipeline(StandardScaler(), LogisticRegression()), 'rc':make_pipeline(StandardScaler(), RidgeClassifier()), Define pipeline containing 4 machine-learning model 'rf':make_pipeline(StandardScaler(), RandomForestClassifier()), 'gb':make_pipeline(StandardScaler(), GradientBoostingClassifier()), list(pipelines.values())[0] #처음 data가 표준화 된 후 logisticregression 모델을 사용해서 학습한다. Pipeline(steps=[('standardscaler', StandardScaler()), ('logisticregression', LogisticRegression())]) fit_models = {} #파이프 라인들 안의 모델들을 이용해서 학습한다. for algo, pipeline in pipelines.items(): model = pipeline.fit(X_train, y_train) fit_models[algo] = model Train model fit models {'Ir': Pipeline(steps=[('standardscaler', StandardScaler()),

Test

❖ Test Code

```
# Make Detections
X = pd.DataFrame([row])
body_language_class = model.predict(X)[0] #landwark의 좌표값을 넣어서 운전상태를 나타내는 class를 body_language_c
body_language_prob = model.predict_proba(X)[0] # 예측한 운전상태(class) 정확도를 측정한다.
print(body_language_class, body_language_prob)
# 영상에 운전상태(class)와 class의 확률을 왼쪽 귀를 기준으로 동일한 곳에 표시하기 위해 왼쪽귀를 기준으로 잡는다.
coords = tuple(np.multiply(
               np.array(
                  (results.pose_landmarks.landmark[mp_holistic.PoseLandmark.LEFT_EAR].x,
                   results.pose_landmarks.landmark[mp_holistic.PoseLandmark.LEFT_EAR].y))
           , [640,480]).astype(int)) #웹캠 frame에 크기를 맞춰준다.
cv2.rectangle(image.
             (coords[0], coords[1]+5),
             (coords[0]+len(body_language_class)*20, coords[1]-30).
             (245, 117, 16), -1)
cv2.putText(image, body_language_class, coords,
           cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2, cv2.LINE_AA)
cv2.rectangle(image, (0,0), (250, 60), (245, 117, 16), -1)
# Display Class
cv2.putText(image, 'CLASS'
            (95,12), cv2.FONT_HERSHEY_SIMPLEX, 0.5, (0, 0, 0), 1, cv2.LINE_AA)
cv2.putText(image, body_language_class.split(' ')[0]
           , (90,40), cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2, cv2.LINE_AA)
# Display Probability
cv2.putText(image, 'PROB'
           , (15,12), cv2.FONT_HERSHEY_SIMPLEX, 0.5, (0, 0, 0), 1, cv2.LINE_AA)
# probobility는 최댓값을 출력하도록 한다.
cv2.putText(image, str(round(body_language_prob[np.argmax(body_language_prob)],2))
           , (10,40), cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2, cv2.LINE_AA)
```

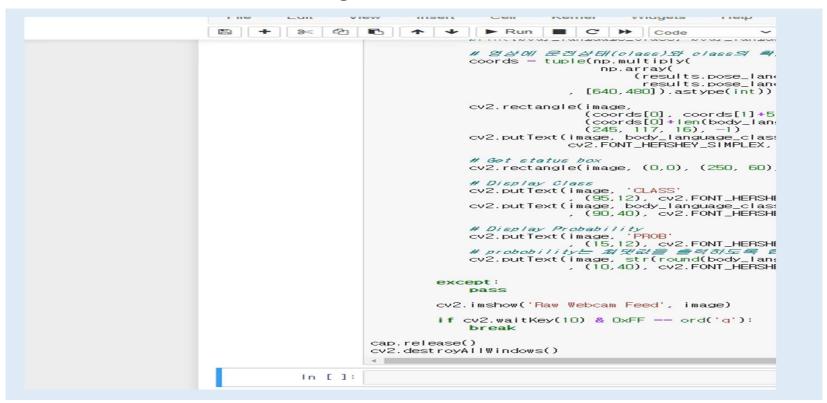
```
coords = tupre(np.murtipry)
                               (results.pose_landmarks.landmark[mp_holistic.PoseLandma
                                results.pose_landmarks.landmark[mp_holistic.PoseLandma
                        , [640,480]).astype(int)) #웹캠 frame에 크기를 맞춰준다.
            cv2.rectangle(image,
                          (coords[0], coords[1]+5),
                         (coords[0]+len(body_language_class)+20, coords[1]-30),
                         (245, 117, 16), -1)
            cv2.putText(image, body language class, coords,
                       cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2, cv2.LINE_AA)
           cv2.rectangle(image, (0,0), (250, 60), (245, 117, 16), -1)
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        except:
        cv2.imshow('Raw Webcam Feed', image)
        if cv2.waitKev(10) & 0xFF == ord('q'):
           break
cap.release()
cv2.destrovAllWindows()
no handling [0.55 0.45]
no handling [0.64 0.36]
no handling [0.75 0.25]
no handling [0.69 0.31]
no handling [0.73 0.27]
```

Expected Performance

- ✓ The automobile can notice that driver is not grabbing the steering wheel although driver place weighty objects on the handle.
- ✓ The steering wheel state warning system can be used with Tesla's technology.
- ✓ Develop systems which alarm or slow down the velocity of automobile when driver is not grabbing the steering wheel.
- ✓ The combined systems(steering wheel state warning system + Tesla's autopilot) can reduce the ratio of traffic accident largely.

Driver State Warning(Additional)

❖Video about driver state warning



This system percepts whether driver is on drowsy driving or safe driving

Thank You!