



# Vision-Based Steering Control and Speed Assistance for Vehicles

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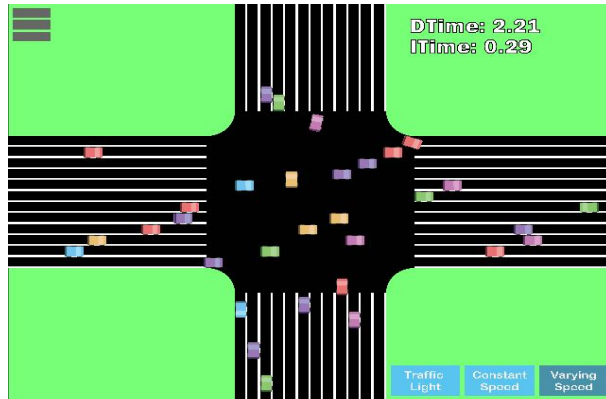
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# Introduction

- Lack of autonomy in Indian Transportation system
- Lots of accident happen due to human errors
- These accident can be avoid using autonomous car
- Stanley - The first Self Driving Car



Source: <http://imgur.com/gallery/h1RZT>

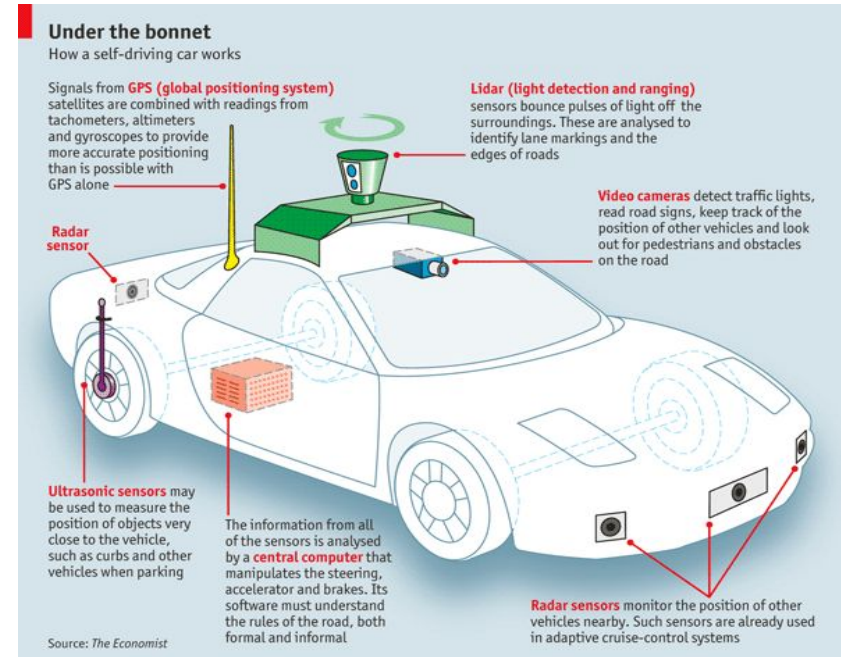


Source: [robots.stanford.edu/talks/stanley/](http://robots.stanford.edu/talks/stanley/)

# Autonomous Car

## Key components of Self Driving Car -

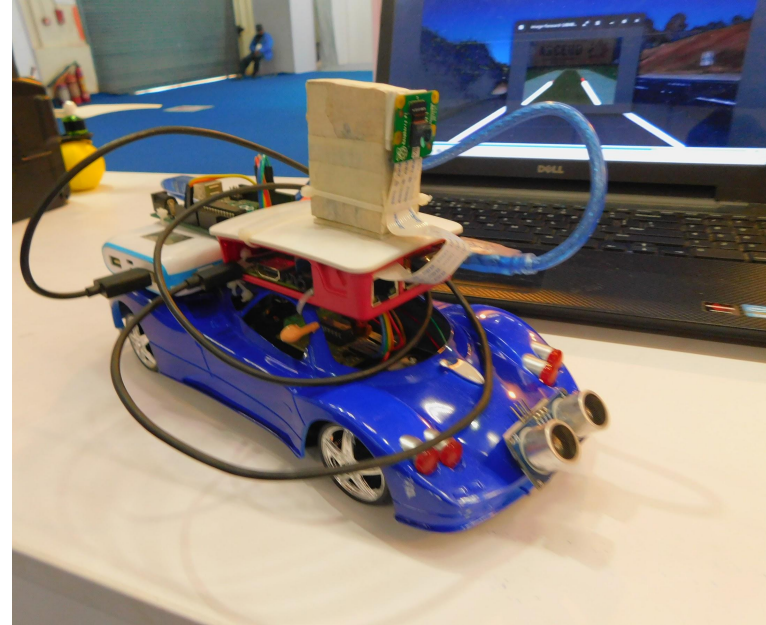
- Radio Detection And Ranging(RADAR)
- Ultrasonic Sensor
- Laser Detection And Ranging(LIDAR)
- GPS
- Video Camera
- Central Computer



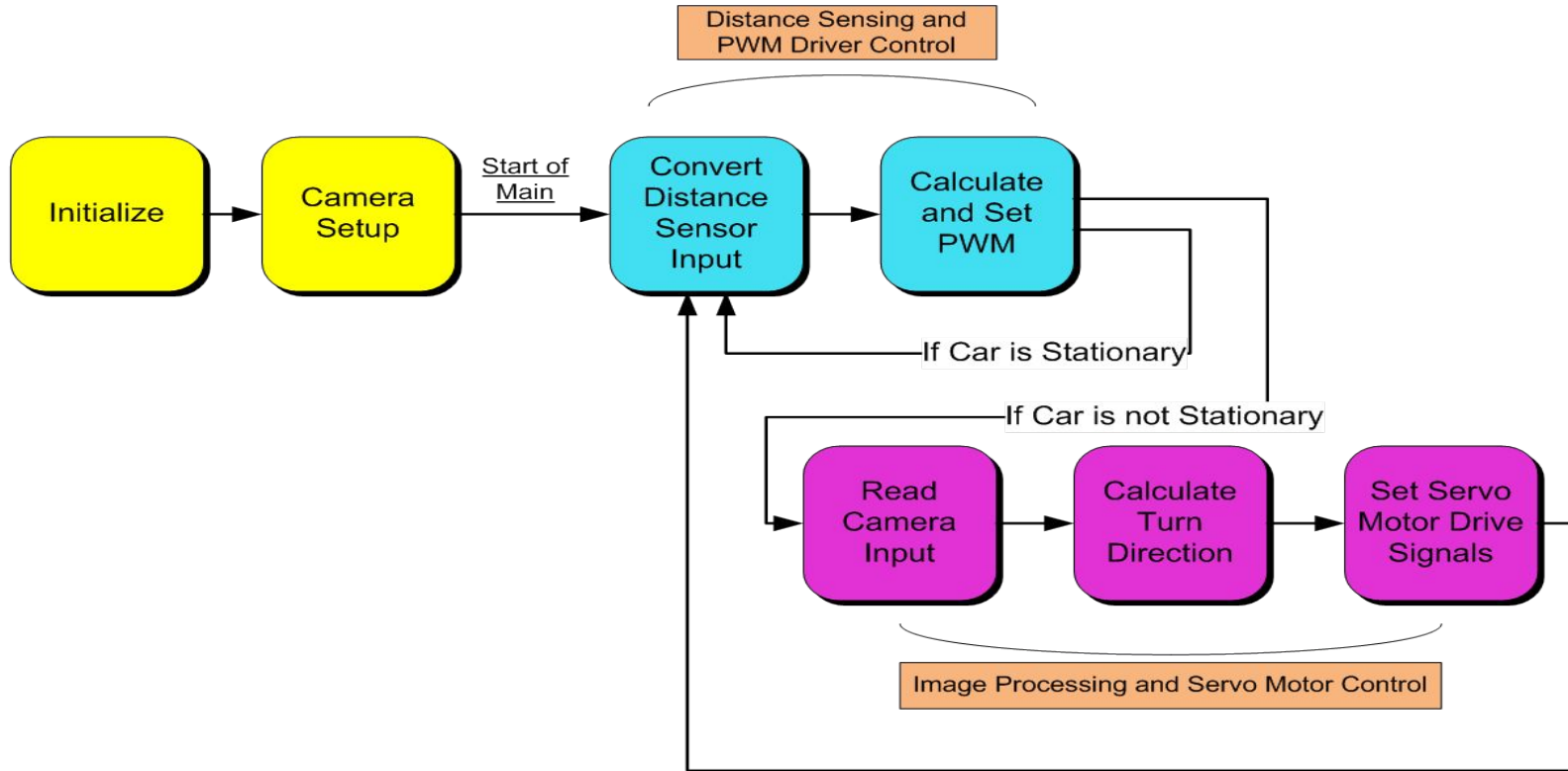
# Objective

## Modify a RC car to handle four tasks:

- Self-driving on a lane track
- Stop sign and traffic light detection
- Front collision avoidance
- Automatic speed control in the track



# Block Diagram of Self Driving RC car



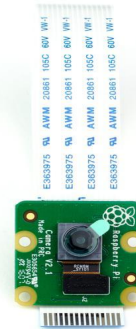
# System Design

The system consists of three subsystems:

- Input unit
  - Camera
  - Ultrasonic sensor
  - Motor driver circuit
  - Power supply
- Processing unit
  - Raspberry Pi
- RC car unit.

# Input Units

- Raspberry pi
- Raspberry pi camera
- Arduino
- Motor Driver circuit
- Ultrasonic sensor
- RC car
- Power bank





## **Raspberry Pi -**

The **Raspberry Pi** is a tiny and affordable micro computer.

- SoC: Broadcom BCM2837 (System on a chip)
- CPU: 4× ARM Cortex-A53, 1.2GHz
- RAM: 1GB
- Storage: microSD
- GPIO: 40-pin header



## **Power Supply -**

- Input Voltage - DC 5V
- Input Current - 500 / 1000 mA
- Output Voltage - 5V
- Output Current - 2A(Max)



## Raspberry Pi Camera -

- Resolution - 8 MP
- Video modes - 1080p30, 720p60 and 640 × 480p60/90
- Sensor - Sony IMX219
- Focal length - 3.04 mm



## Arduino -

- Microcontroller board based on the ATmega328P
- 14 digital input/output pins
- 16 MHz quartz crystal
- USB connection
- Power jack



## **Motor Driver Circuit -**

- Driver: L293D
- power supply: +5V ~ +46V
- I<sub>o</sub>: 2A
- Max power: 25W (Temperature 75 celsius)

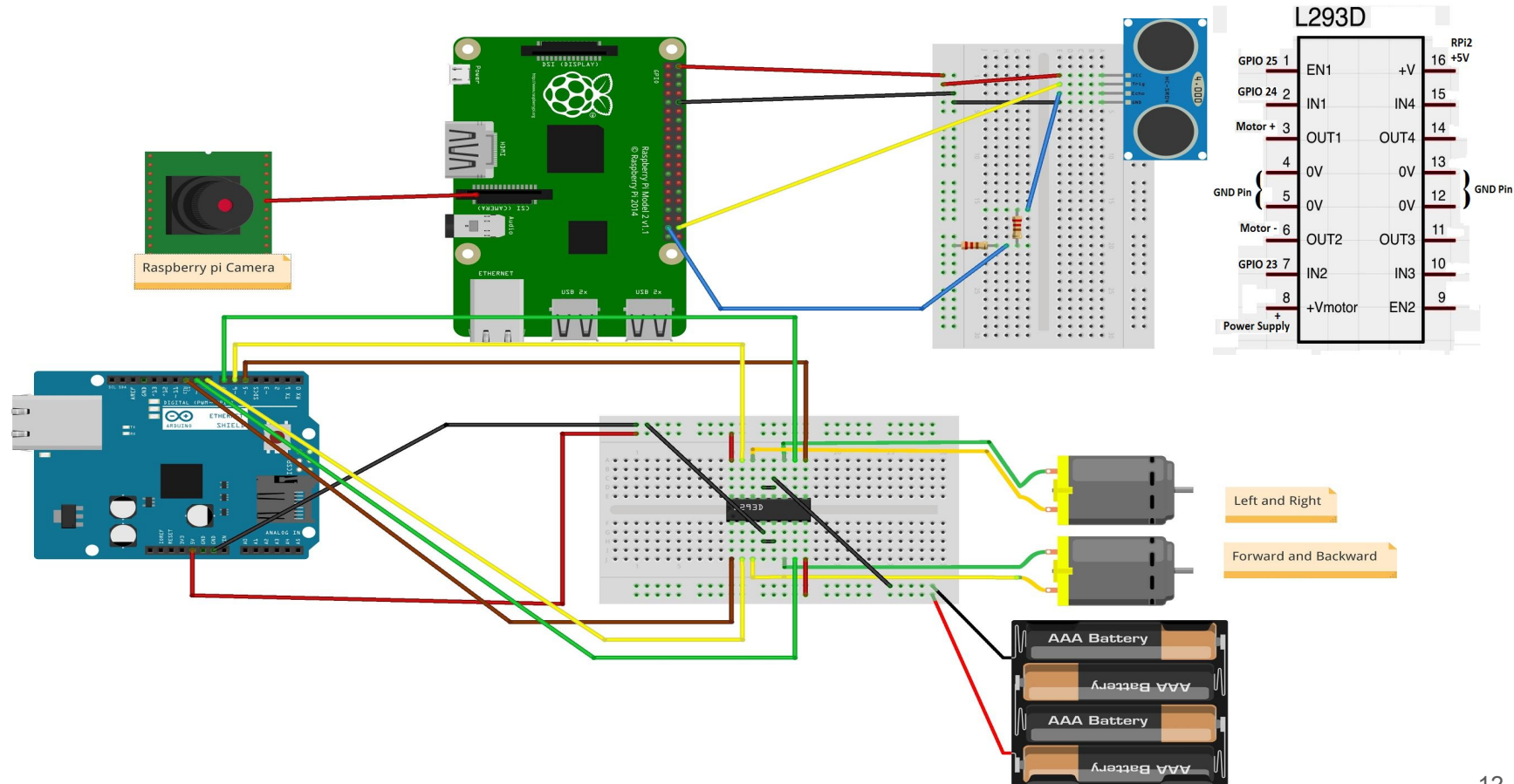


## **Ultrasonic Sensor (HC-SR04)-**

- Supply voltage 5V
- Ultrasonic Frequency 40 kHz
- Maximal Range 400 cm
- Minimal Range 2 cm
- Trigger Pulse Width 10  $\mu$ s
- Ultrasonic Sound velocity = 343 m/s



# Circuit Diagram



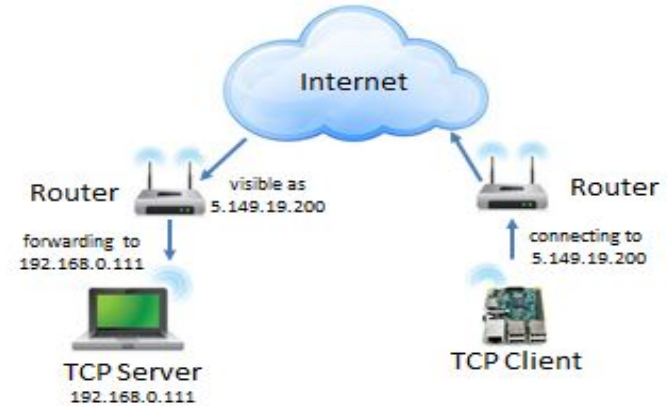
# Processing Unit

## The processing unit handles multiple tasks:

- Send data from to computer for training process.
- Neural network training and prediction(steering).
- Object detection(stop sign and traffic light).
- Distance measurement(collision avoidance).
- Sending instructions to Arduino through USB connection.

# TCP Server

- A server program that runs on the computer.
- Receive streamed image frames and ultrasonic data from the Raspberry Pi.
- Image frames are converted to grayscale and are decoded into matrix.
- Train the neural network and then send the model to Raspberry Pi.



# Object Detection

- Detect stop sign and traffic light (Red and Green)
- Used Haar cascade classifiers for object detection.
- Calibrate the camera for distance measurement.



# Driving on a lane track by Automatic steering control -

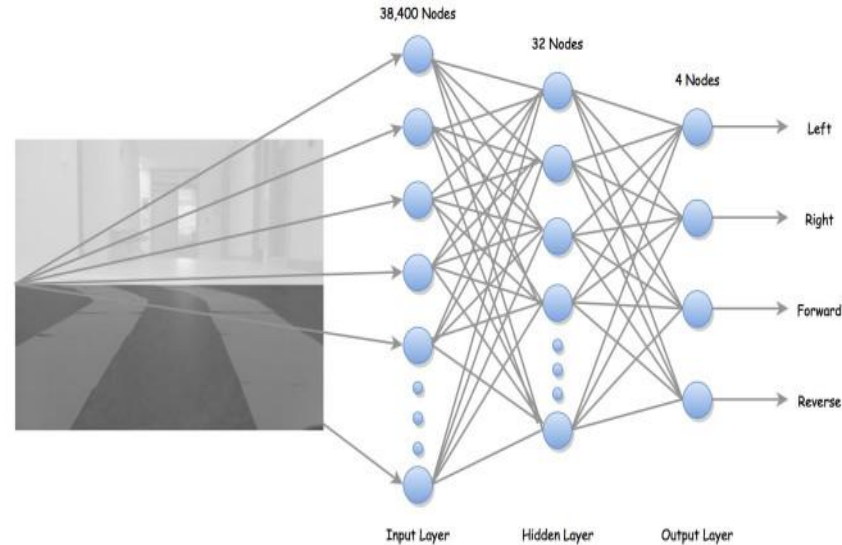
- Create path by A4 size paper.
- Define region of interest of the track by camera angle
- Capture the image in grayscale and store it while manually driving
- Label the image according to the steering action
- Define the parameter for neural network
- Feed the input parameters to the neural network algorithm
- Save the model
- Run the model in Raspberry Pi in Lane track





# Neural Network architecture and it parameters -

- Input layer has **38,400** ( $320 \times 120$ ) nodes and **32** nodes (chosen arbitrarily) in the hidden layer, **4** nodes in the output layer (left, right, forward and reverse).
- Using Back Propagation algorithm to get the result.
- Once the network is trained, it only needs to load trained model afterwards, thus prediction can be very fast.
- It happens in real time with less latency.



# Workflow of Back Propagation Algorithm -

- Set the input parameters - 1.Input layer size      2.Hidden layer size      3.Numbers of label
- Set the training parameters -
  - Image array
  - Labels(Forward,Backward,Left,Right)
- Taking initial weight by random with extra bias 1 in each input layer.
- For each layer calculate Output of individual neuron =  $a_j^i = \sigma(\sum_k (w_{jk}^i \cdot a_k^{i-1}) + b_j^i)$
- Cost function =  $J(\Theta) = -\frac{1}{m} \left[ \sum_{i=1}^m \sum_{k=1}^K y_k^{(i)} \log h_{\theta}(x^{(i)})_k + (1 - y_k^{(i)}) \log(1 - h_{\theta}(x^{(i)})_k) \right]$
- For optimisation of cost function we need to derivative w.r.t weight.  $\frac{\partial J_{\Theta}}{\partial \Theta} = 0$
- Then save the model with optimize parameters for test.

# Notation used

- $x$  - Input
- $y$  - Output
- $\theta$  - weight
- $m$  - No of Training example
- $i$  - index of units in a layer
- $j$  - index of layers
- $k$  - index of labels
- $\sigma$  = Activation function (Sigmoid function)

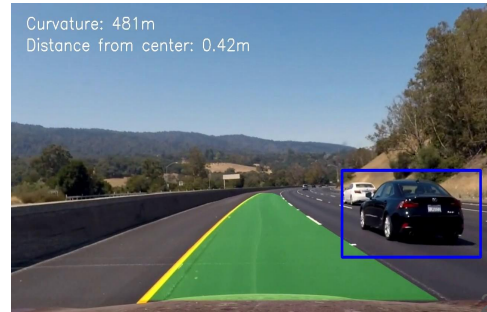
$w_{jk}^i$  is the weight from the  $k^{th}$  neuron in the  $(i - 1)^{th}$  layer to the  $j^{th}$  neuron in the  $i^{th}$  layer,

$b_j^i$  is the bias of the  $j^{th}$  neuron in the  $i^{th}$  layer, and

$a_j^i$  represents the activation value of the  $j^{th}$  neuron in the  $i^{th}$  layer.

# How to apply this in Real life

- Using image processing technique for better accuracy.
- Then apply the same algorithm with after processing the images or video.
- Then control the steering according to it.



# Result

- Prediction on the testing samples returns high accuracy if more amount of image collected.
- In actual driving situation, predictions are generated about 10 times a second (streaming rate roughly **10 frames/sec**).
- Haar cascaded classifier is rotation sensitive. In this project, however, rotation is not a concern as both the stop sign and the traffic light are fixed objects, which is also a general case in real world environment.
- Ultrasonic sensor is only used to determine the distance to an obstacle in front of the RC car and provides accurate results , On the other hand, Pi camera provides good enough measurement results after calibration of camera.

# Application

- Autonomous Transportation Systems  
inside industry
- Autonomy in Agricultural field
  - Automatic Weeder Machine
- Reduce human error
  - Prevent accident



# Future Work

- Implement the set up in the real automobile or on any robot.
- Integrate with powerful hardware like NVIDIA.
- Integrating with Robot Operating System (ROS) for control the hardware.
- Integrating GPS and Compass module to be added for long distance travel.

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# Q&A

Thank you