



**The network
structure,
visualization and
pictures of the
robot**



Modeling: Image classification



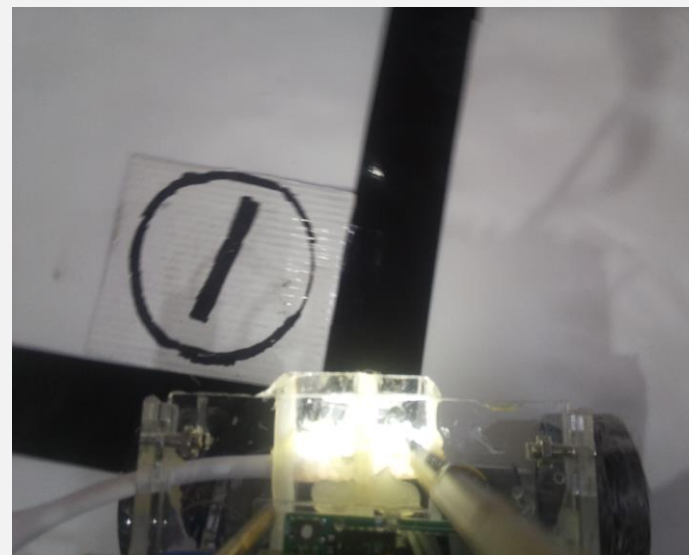
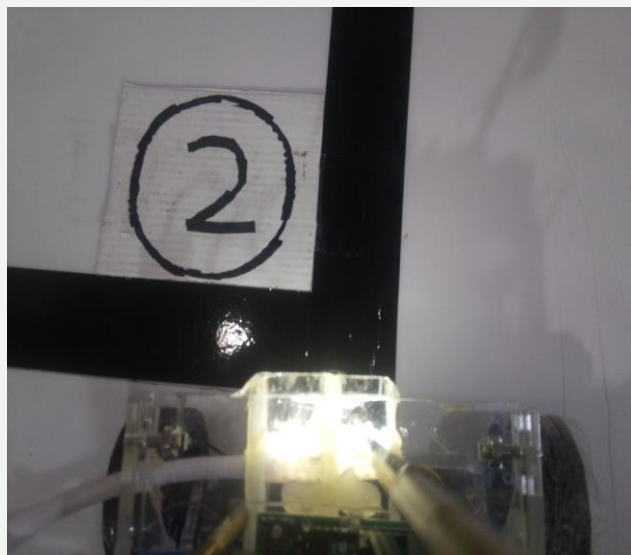


I utilized 3 models:

Model1: line tracking
and identify
intersections and parking
slots

Model2: self parking

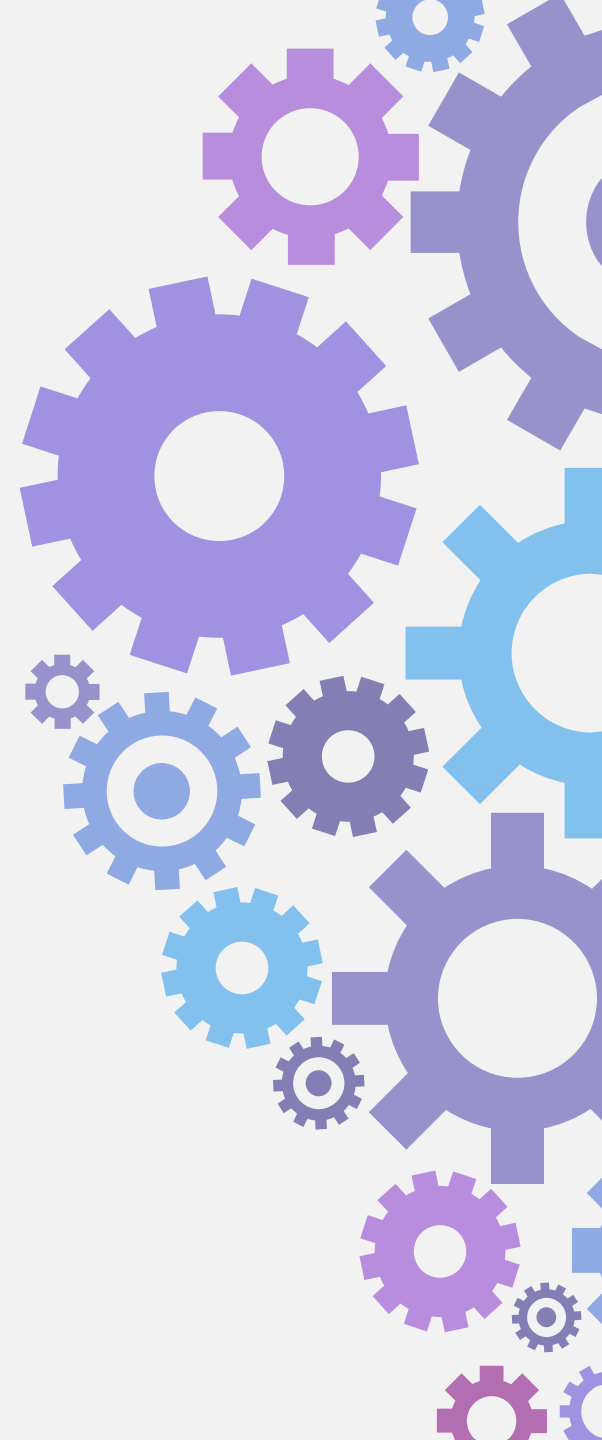
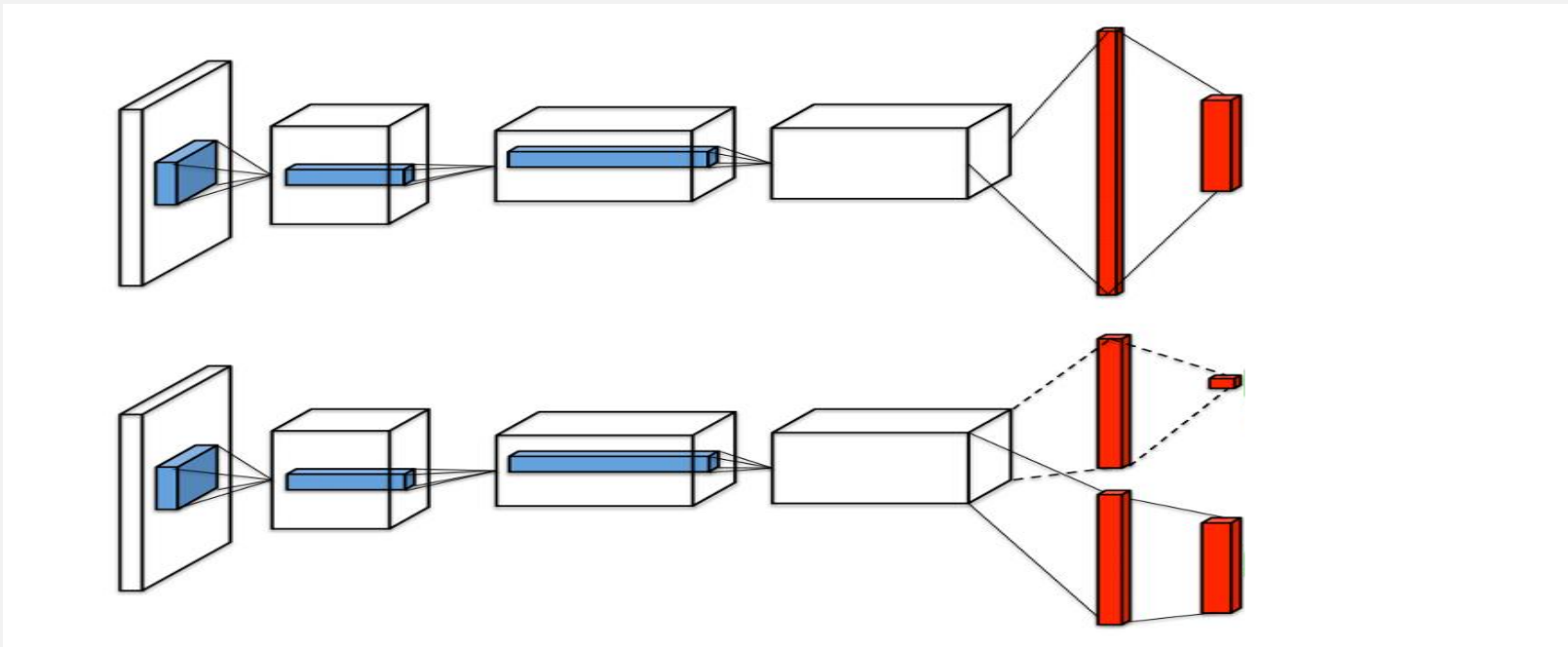
Model3: number
identification



Choice of network:

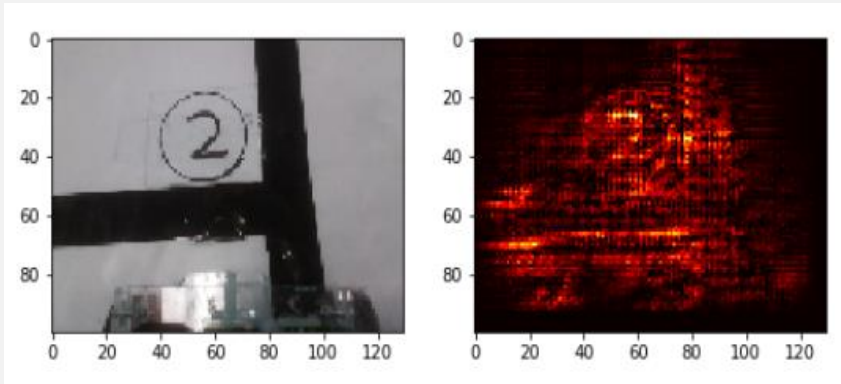
CNN + full connection layer

- ◆ *Synthesized network: concise codes, sample number need is low, but too much forward propagation time*
- ◆ *Seperated network: low forward propagation time.*



Sampling and visualization:

1. Different conditions, different lighting
2. Sampling with the lines: failed. Too much points of concentration from the visualized heat colormap.



3. Adding irrelevant samples

Improved generalization ability, concentrated interest points network



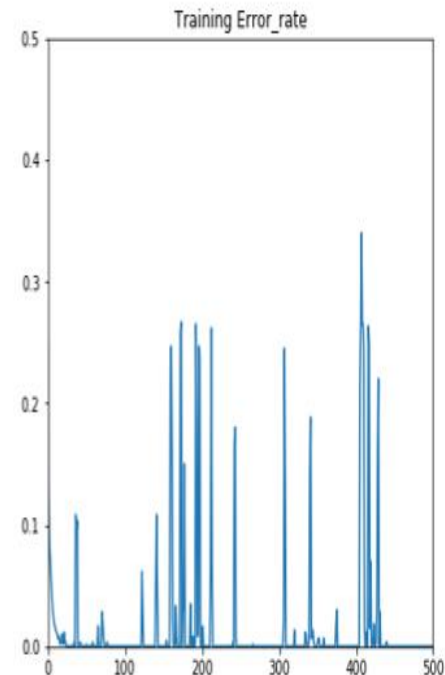
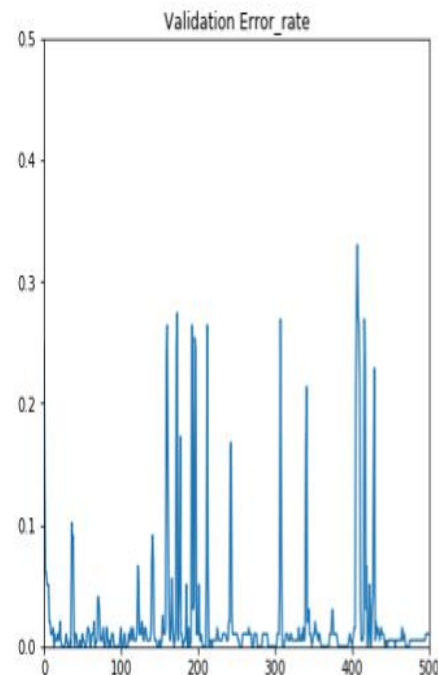
Training and testing the networks

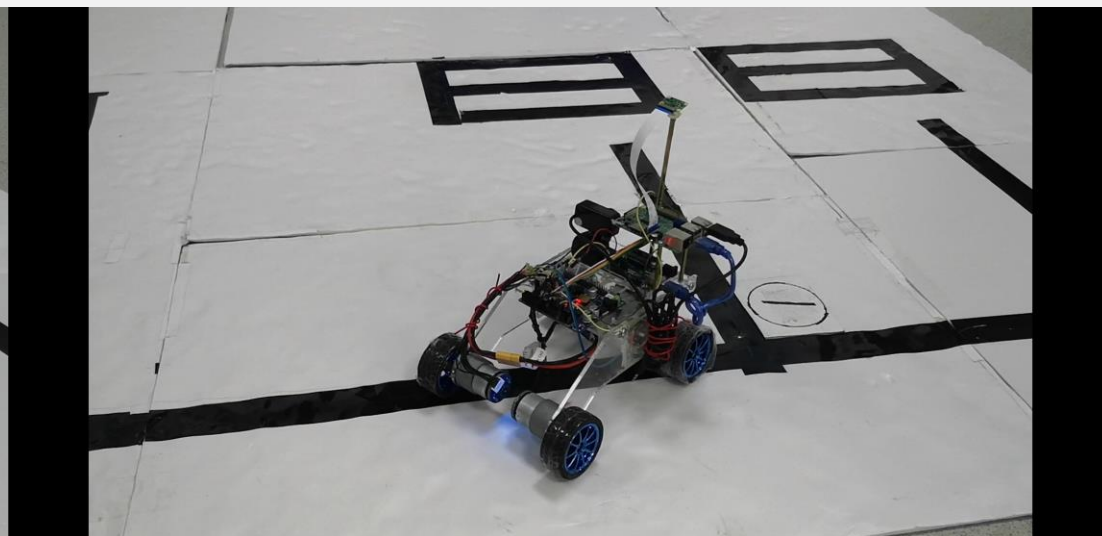
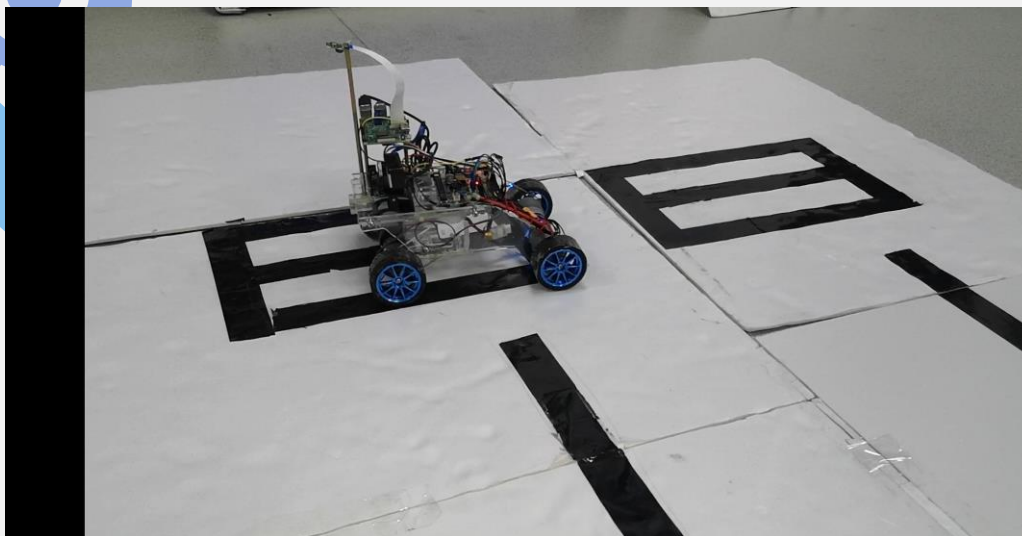
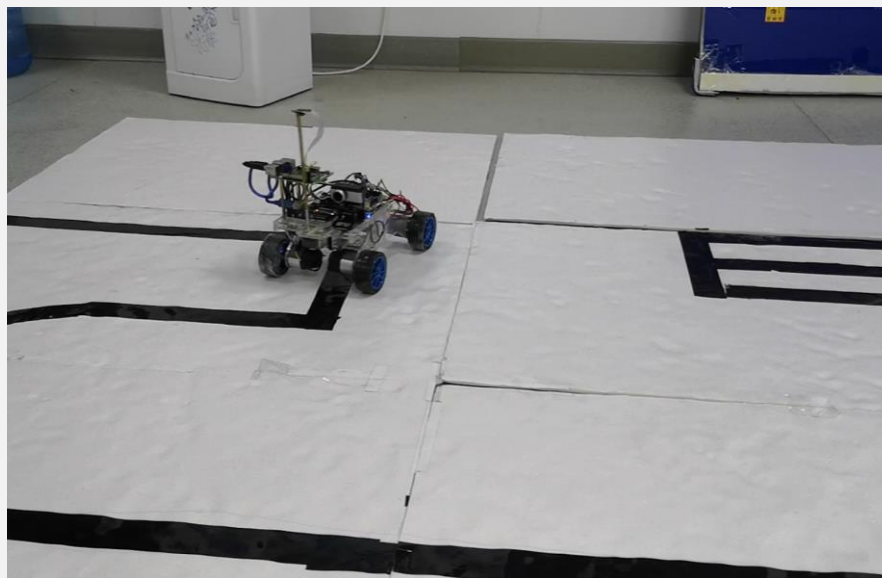
```
#hyperparameters
epochs=500
screen_height=100
screen_width=130##130
#batch_size=10
acc_v=torch.zeros((epochs))
acc_t=torch.zeros((epochs))
policy_net = DQN(screen_height, screen_width,4).to(device)
optimizer = optim.Adam(policy_net.parameters())#RMSprop
policy_net.train()
i=0
import time
tic=time.time()
num=1
for epoch in range(epochs):
    for step,(st,at) in enumerate(loader):
        policy_net.train()
        y_pred=policy_net(st)

        cri = nn.CrossEntropyLoss()
        loss=cri(y_pred,at.long())
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()
    policy_net.eval()
    y_predl=torch.argmax(policy_net(_st),1)

    acc=torch.sum(_at.long()==y_predl.long()).item()/len(_at)
    print(' |epoch', epoch, ' |loss:', loss, ' |accuracy:', acc)
    if epoch%100==99:
        torch.save(policy_net.state_dict(), 'F:/Linetrack_project/samples/fii/model%s.h5'%int(num))
        num+=1
    acc_v[epoch]=1-acc
    #if acc>=0.97:
    #break
    y_pred2=policy_net(S)
    y_pred2=torch.argmax(policy_net(S),1)
    acc=torch.sum(A.long()==y_pred2.long()).item()/len(A.long())
    #print(' |epoch', epoch, ' |loss:', loss, ' |accuracy:', acc)
    acc_t[epoch]=acc

toc=time.time()
print(' time:', toc-tic) #RMSprop
```







**Thanks for
reading!**

