

✓ Pytorch with used cars

```
#####
### basic imports
import numpy as np
import pandas as pd
import math
import matplotlib.pyplot as plt

### sklearn
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split

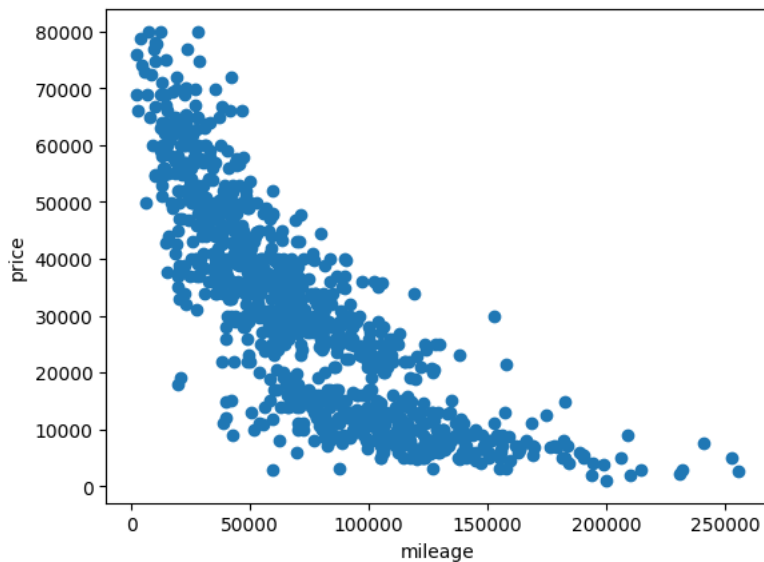
### pytorch
import torch
from torch.utils.data import Dataset
from torch.utils.data import DataLoader
from torch import nn

#####
### read in data and pull off x and y
dd = pd.read_csv("https://bitbucket.org/remcc/rob-data-sets/downloads/susedcars.csv")

y = dd['price'].to_numpy()
x = dd[['mileage','year']].to_numpy()

## look at data
plt.scatter(x[:,0],y)
plt.xlabel('mileage'); plt.ylabel('price')

↩ Text(0, 0.5, 'price')
```



```
#####
### train-test split

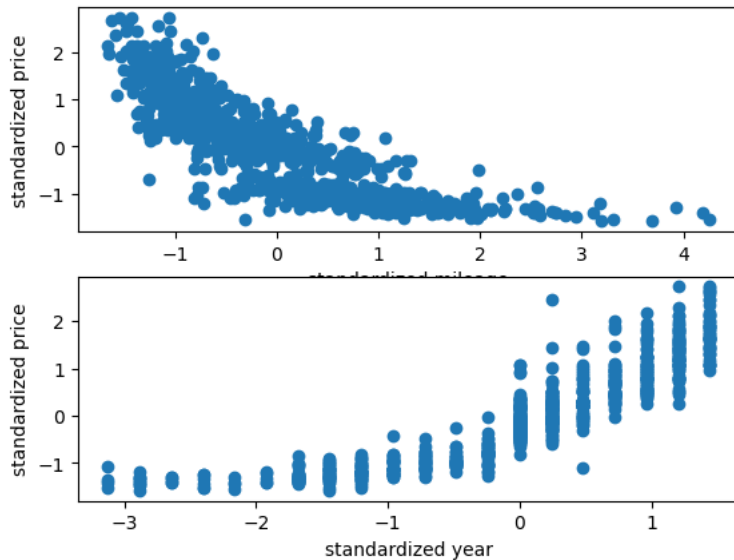
Xtr, Xte, Ytr, Yte = train_test_split(x,y,random_state=99, test_size=.2)

#####
## scale both x and y
sc = StandardScaler()
xtr = sc.fit_transform(Xtr)
xte = sc.transform(Xte)

scy = StandardScaler()
ytr = scy.fit_transform(Ytr.reshape(-1,1))
yte = scy.transform(Yte.reshape(-1,1))
```

```
fig, ax = plt.subplots(2)
ax[0].scatter(xtr[:,0],ytr)
ax[1].scatter(xtr[:,1],ytr)
ax[0].set_ylabel('standardized price')
ax[1].set_ylabel('standardized price')
ax[0].set_xlabel('standardized mileage')
ax[1].set_xlabel('standardized year')
```

↗ Text(0.5, 0, 'standardized year')



```
#####
## run a regression
```

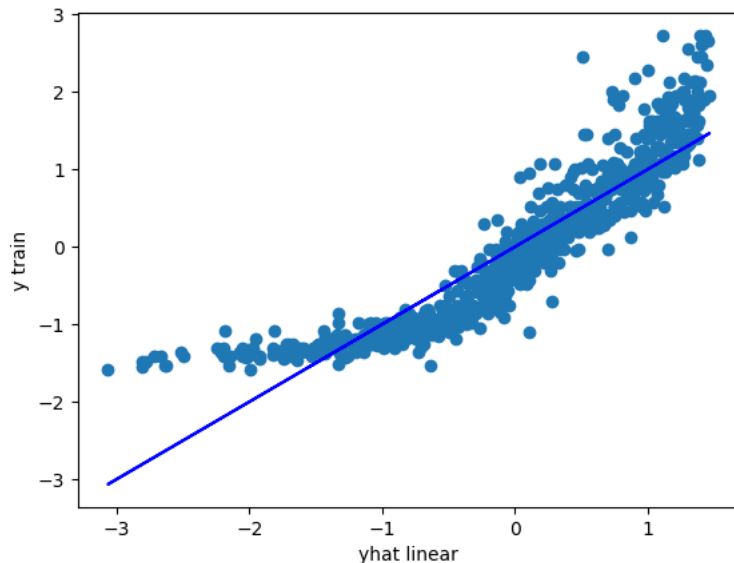
```
lmmod = LinearRegression(fit_intercept=True)
lmmod.fit(xtr,ytr)
print("Model Slopes:      ",lmmod.coef_)
print("Model Intercept:",lmmod.intercept_)
```

```
yhat = lmmod.predict(xtr)
```

↗ Model Slopes: [[-0.34300427 0.6241386]]
Model Intercept: [1.22554284e-14]

```
## train linear fit
plt.scatter(yhat,ytr)
plt.plot(yhat,yhat,c='blue')
plt.xlabel('yhat linear'); plt.ylabel('y train')
```

↗ Text(0, 0.5, 'y train')

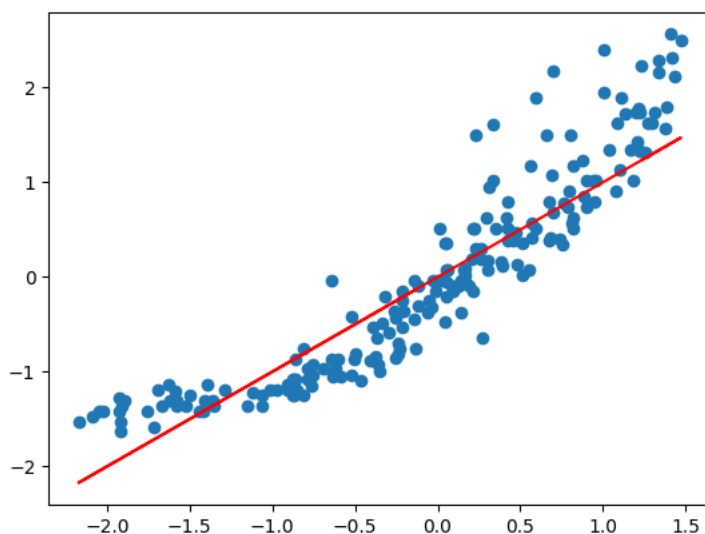


```
error = ytr-yhat
msee = (error*error).sum()/ytr.shape[0]
rmseLin = math.sqrt(msee)
```

```
## test linear fit
yhtr = lmmod.predict(xtr)
```

```
plt.scatter(yhtr,ytr)
plt.plot(yhtr,yhtr,c='red')
```

```
error = ytr-yhtr
msee = (error*error).sum()/ytr.shape[0]
rmseLinTe = math.sqrt(msee)
```



```
#####
```

```
## move to Tensors
```

```
xtr = xtr.astype('float32')
xxtr = torch.from_numpy(xtr)
ytr = ytr.astype('float32')
yytr = torch.from_numpy(ytr)
```

```
xte = xte.astype('float32')
xxte = torch.from_numpy(xte)
yte = yte.astype('float32')
yyte = torch.from_numpy(yte)
```

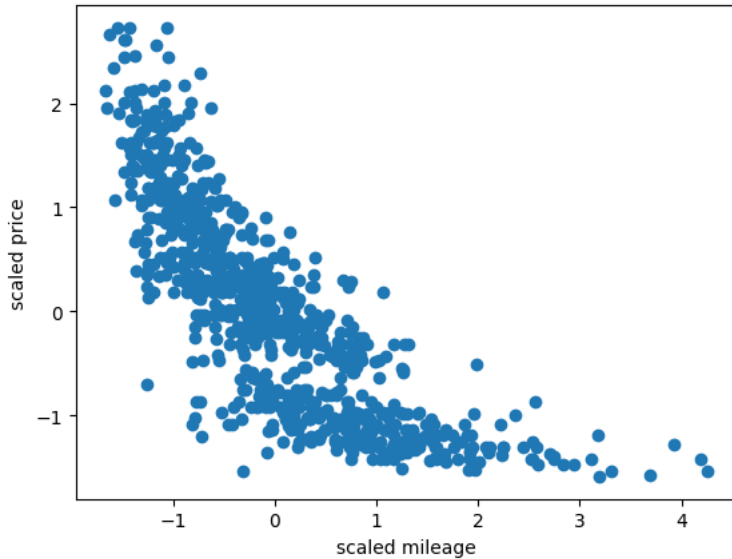
```
plt.scatter(xxtr[:,0],yytr)
plt.xlabel('scaled mileage'); plt.ylabel('scaled price')
```

```
print(xxtr.shape)
print(xxtr.dtype)
print(yytr.shape)
print(yytr.dtype)
print(yyte.dtype)
```

```

torch.Size([800, 2])
torch.float32
torch.Size([800, 1])
torch.float32
torch.float32

```



```

#####
## Dataset

```

```

class DF(Dataset):
    def __init__(self,x,y,transform=None, target_transform=None):
        self.x = x
        self.y = y
        self.transform = transform
        self.target_transform = target_transform

    def __len__(self):
        return len(self.y)

    def __getitem__(self,idx):
        return self.x[idx], self.y[idx]

```

```
tDF = DF(xxtr,yytr)
```

```

#####
## DataLoader
tdl = DataLoader(tDF,batch_size=50,shuffle=True)

```

```

#####
## model

```

```

class SLNN(nn.Module):
    def __init__(self,nunits=5):
        super(SLNN, self).__init__()
        self.SSM = nn.Sequential(
            nn.Linear(2,nunits),
            nn.Sigmoid(),
            nn.Linear(nunits,1))
    def forward(self,x):
        yhat = self.SSM(x)
        return yhat

```

```

nunits= 50
model = SLNN(nunits)

```

```

## see model
print(model)
## print out parameters
#for i, param in enumerate(model.parameters()):
#    print(f'***** {i}\n')
#    print(param)

```

```

SLNN(
  (SSM): Sequential(
    (0): Linear(in_features=2, out_features=50, bias=True)
    (1): Sigmoid()
    (2): Linear(in_features=50, out_features=1, bias=True)
  )
)

#####
### do it
def train_loop(dataloader, model, loss_fn, optimizer):
    size = len(dataloader.dataset)
    for batch, (X, y) in enumerate(dataloader):
        # Compute prediction and loss
        pred = model(X)
        loss = loss_fn(pred, y)

        # Backpropagation
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()
    return loss_fn(model(dataloader.dataset.x),dataloader.dataset.y).item()

learning_rate = 1e-3
l2par = .005

loss_fn = nn.MSELoss()
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate, weight_decay = l2par)

epochs = 5000
lv = np.zeros((epochs,))
for t in range(epochs):
    print(f"Epoch {t+1}\n-----")
    lv[t] = train_loop(tdl, model, loss_fn, optimizer)
print("Done!")

```



```
Epoch 4993
```

```
Epoch 4994
```

```
Epoch 4995
```

```
Epoch 4996
```

```
Epoch 4997
```

```
Epoch 4998
```

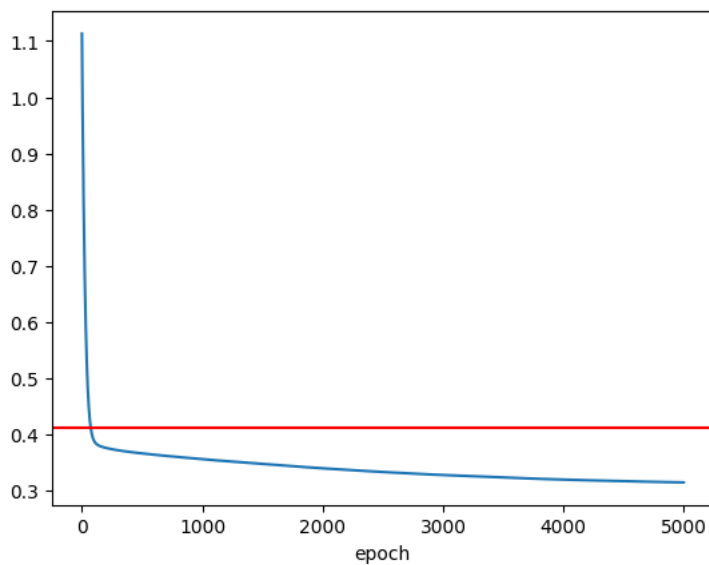
```
Epoch 4999
```

```
Epoch 5000
```

```
Done!
```

```
## plot loss over epochs
plt.plot(np.sqrt(lv))
plt.axhline(rmseLin,c='red')
plt.xlabel('epoch')
```

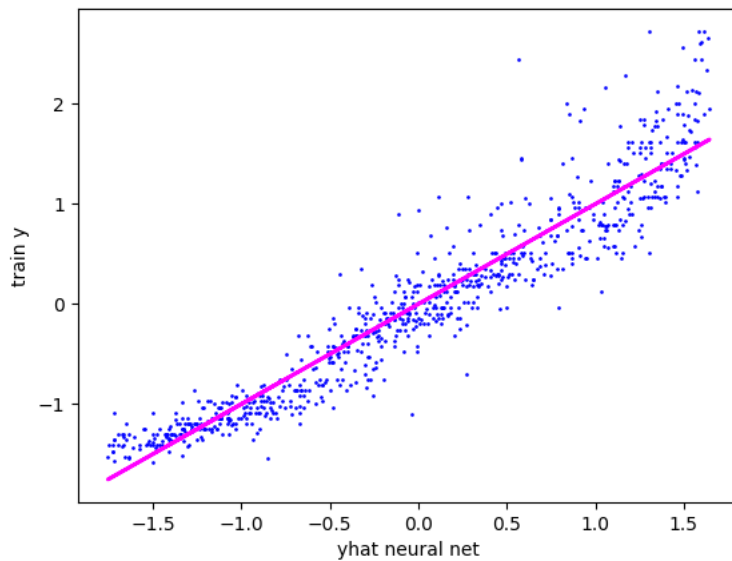
```
↗ Text(0.5, 0, 'epoch')
```



```
## plot fit
yyhat = model(xxtr).detach()
plt.scatter(yyhat,yytr,s=.85,c='blue')
#plt.scatter(yhat,yytr,s=.5,c='red')
plt.plot(yyhat,yyhat,c='magenta',linewidth=2)
plt.xlabel('yhat neural net'); plt.ylabel('train y')
tstr = 'fit from neural net with ' + str(nunits) + ' units'
plt.title(tstr)
#plt.legend(['y=x','neural net','linear'])
```

```
Text(0.5, 1.0, 'fit from neural net with 50 units')
```

fit from neural net with 50 units

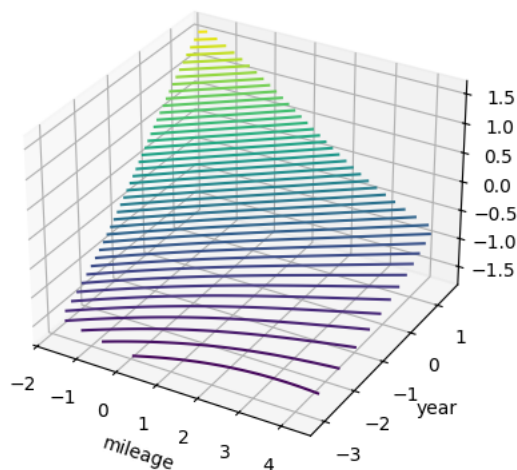


```
#####  
## get fit on 2-way grid
```

```
ng=100  
xg0 = np.linspace(xxtr[:,0].min(), xxtr[:,0].max(),ng)  
xg1 = np.linspace(xxtr[:,1].min(), xxtr[:,1].max(),ng)  
  
xx0, xx1 = np.meshgrid(xg0,xg1)  
  
xg = np.hstack([xx0.reshape((ng*ng,1)),xx1.reshape((ng*ng,1))])  
xg = xg.astype('float32')  
xgT = torch.from_numpy(xg)  
  
Yhatg = model(xgT).detach()  
Yhatg2 = Yhatg.reshape((ng,ng))
```

```
fig = plt.figure()  
ax = plt.axes(projection='3d')  
ax.contour3D(xx0,xx1,Yhatg2,50,cmap='viridis')  
ax.set_xlabel('mileage')  
ax.set_ylabel('year')  
ax.set_zlabel('price')
```

```
Text(0.5, 0, 'price')
```



```
#####  
### out of sample  
yhatte = model(xxte).detach().numpy()
```

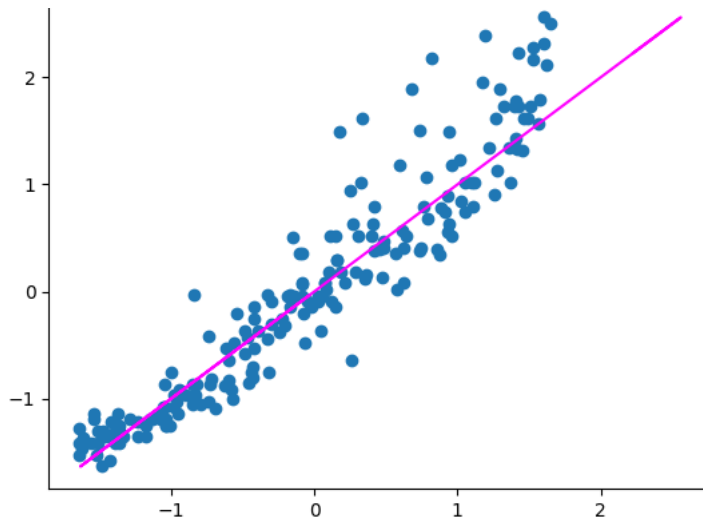
```
plt.scatter(yhatte,yyte)
plt.plot(yyte,yyte,c='magenta')
```

```
error = yyte.detach().numpy() - yhatte
msee = (error*error).sum()/yyte.shape[0]
rmseNnTe = math.sqrt(msee)
```

```
print(f'\n*****test rmse: Linear {np.round(rmseLinTe,3)}, NN {np.round(rmseNnTe,3)}\n')
```



```
*****test rmse: Linear 0.436, NN 0.361
```



```
print(type(error))
ee = yyte.detach().numpy() - yhatte
```



```
<class 'torch.Tensor'>
```

```
## original scale
yhatte0 = scy.inverse_transform(yhatte)
yhatteL = lmmod.predict(xte)
yhatteL0 = scy.inverse_transform(yhatteL)
```

```
plt.scatter(yhatte0,Yte,c='blue',s=1.5,label='nnet')
plt.scatter(yhatteL0,Yte,c='red',s=1.2,label='linear')
plt.title('out of sample predictions of test data')
plt.xlabel('yhat'); plt.ylabel('y')
plt.plot(Yte,Yte,c='magenta')
plt.legend()
```



```
<matplotlib.legend.Legend at 0x7c07618871d0>
```

