

**MACHINE LEARNING DAY 2**

# **DEEP LEARNING**

## **Session II: Linear regression**



**Isaac Ye, HPTC @ York University**

**[Isaac@sharcnet.ca](mailto:Isaac@sharcnet.ca)**

# Session II

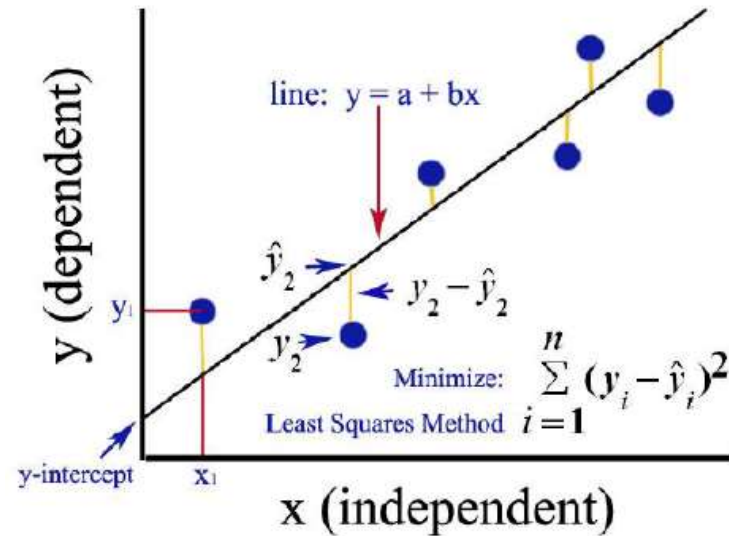
- Linear regression (multi-variables)
- PyTorch model/cost function/optimizer
- *Lab 2A: Multivariable linear regression with PyTorch*
- Running DL in Graham
- *Lab 2B - Working in Graham and running a simple code*

# Categories of ML problems

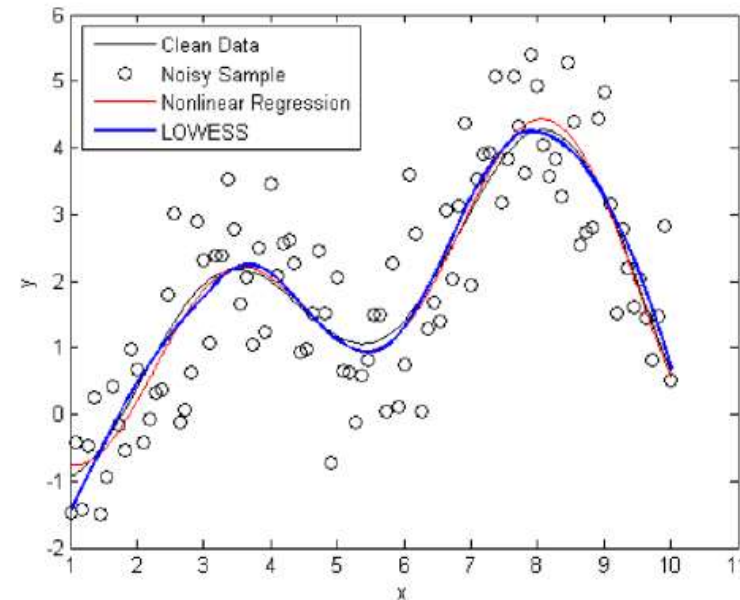
	Supervised	Unsupervised	Reinforcement
Discrete	<b>Classification</b>	<b>Clustering</b>	<b>Action space agent</b>
Continuous	<b>Regression</b>	<b>Dimensionality reduction</b>	<b>Action space agent</b>

# Regression problem

Fit the prediction function  $f(x)$  to the training data to predict continuous real value



Linear regression



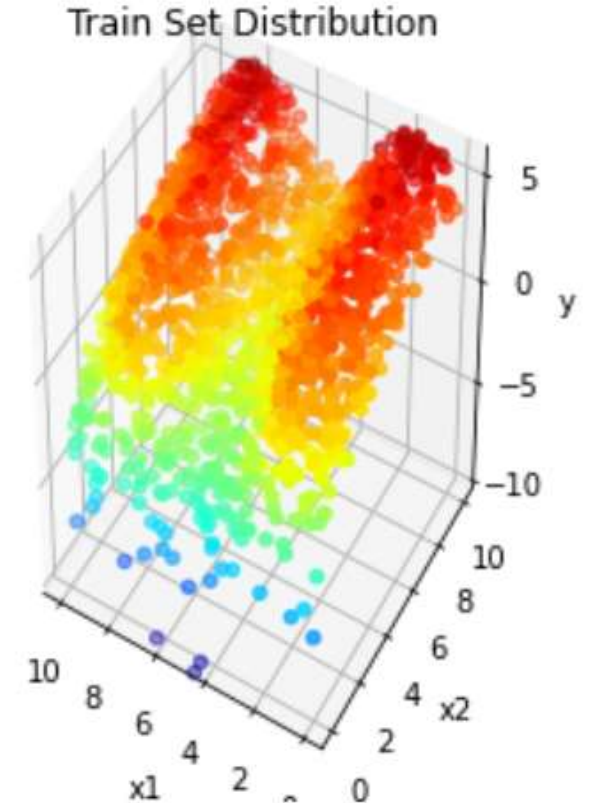
Nonlinear regression

# Linear regression: multivariable

## Data preparation: Input ( $x_1, x_2, y$ )

$x_1$	$x_2$	$y$
3.91870851	2.32626914	0.73817558
2.59194437	6.00656071	4.3940048
6.46991632	3.57514815	0.61488728
:	:	:
4.56486433	2.14296641	3.95964088
1.29483514	1.67730041	3.48018992

```
num_data = 2400
x1 = np.random.rand(num_data) * 10
x2 = np.random.rand(num_data) * 10
e = np.random.normal(0, 0.5, num_data)
X= np.array([x1,x2]).T # T for transpose from (2, 2400) to (2400, 2)
y=2*np.sin(x1) + np.log(0.5*x2**2)+e
```



# Model (Hypothesis)

$$H(x_1, x_2) = w_1x_1 + w_2x_2 + b$$

For the data with  $n$  number of features, it is can be written as

$$H(x_1, x_2, x_3, \dots, x_n) = w_1x_1 + w_2x_2 + w_3x_3 + \dots + w_nx_n + b$$

# Expression in matrix

$$H(x_{i1}, x_{i2}) = w_1 x_{i1} + w_2 x_{i2} + b$$

$$[x_1 \quad x_2] \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b = w_1 x_1 + w_2 x_2 + b$$

$$\begin{bmatrix} x_{11} & x_{12} \\ x_{21} & x_{22} \\ \vdots & \vdots \\ x_{n1} & x_{n2} \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} + b = \begin{bmatrix} w_1 x_{11} + w_2 x_{12} + b \\ w_1 x_{21} + w_2 x_{22} + b \\ \vdots \\ w_1 x_{n1} + w_2 x_{n2} + b \end{bmatrix}$$

$$H(X) = XW + b$$

# Layout

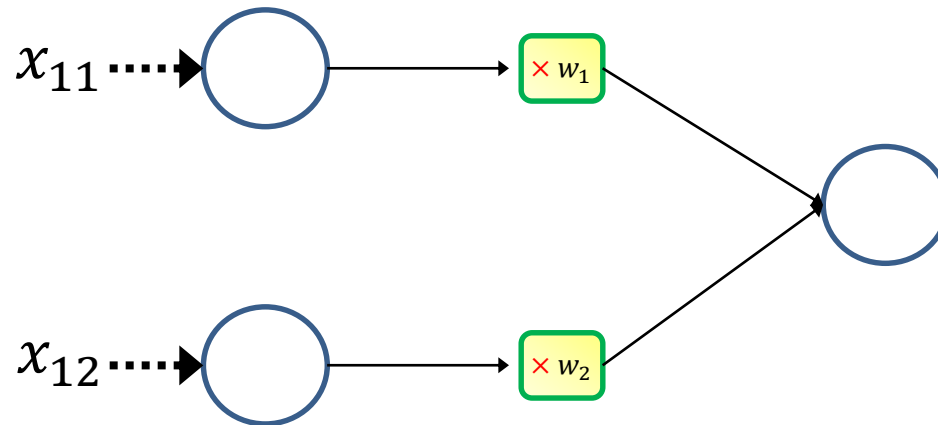
Input features = 2  
Output features = 1

Input layer

# of feature = 1

Output layer

$$\begin{bmatrix} x_{11} & x_{12} \\ x_{21} & x_{22} \\ \vdots & \vdots \\ x_{n1} & x_{n2} \end{bmatrix}$$



$$H(x_{i1}, x_{i2}) = w_1 x_{i1} + w_2 x_{i2}$$

Let's consider a simple  
case with  $W$  only.



# Cost function

$$H(X) = XW + b$$

$$cost = \frac{1}{m} \sum_{i=1}^m (H(x_{i1}, x_{i2}) - y_i)^2$$

We want to minimize the cost as well!

# Algorithm structure

Data  
Preparation

Model  
define

Cost  
function  
+ optimizer

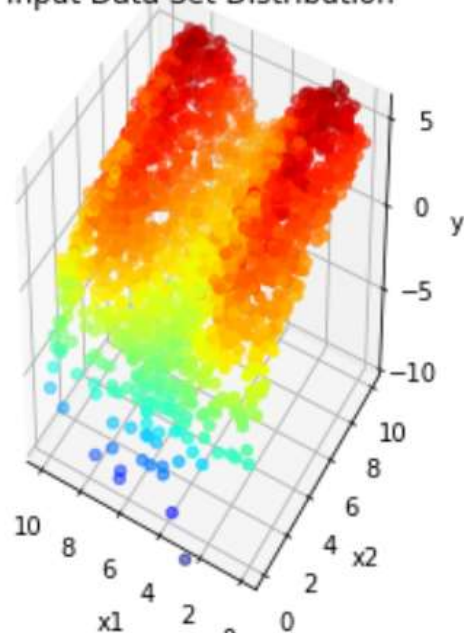
Model  
Test

# Data preparation

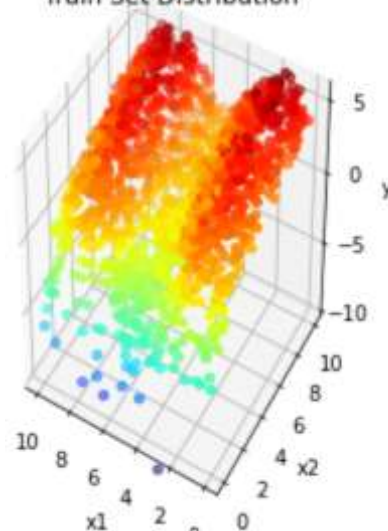
Data  
Preparation

$x_1$	$x_2$	$y$
3.91870851	2.32626914	0.73817558
2.59194437	6.00656071	4.3940048
6.46991632	3.57514815	0.61488728
:	:	:
4.56486433	2.14296641	3.95964088
1.29483514	1.67730041	3.48018992

Input Data Set Distribution

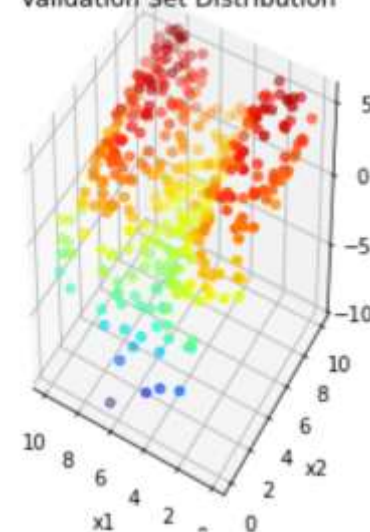


Train Set Distribution



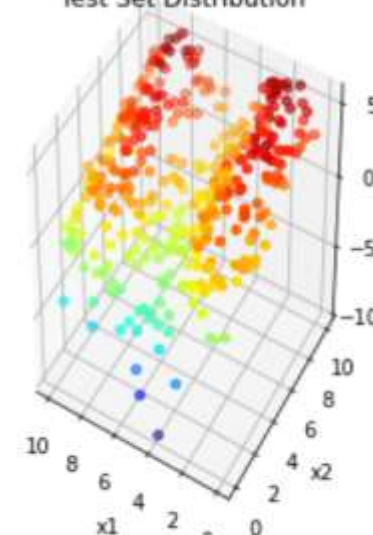
Train set

Validation Set Distribution



Validation set

Test Set Distribution



Testing set

In the code

```
train_X, train_y = X[:1600, :], y[:1600]  
val_X, val_y = X[1600:2000, :], y[1600:2000]  
test_X, test_y = X[2000:, :], y[2000:]
```

# Model define: linear regression

Model  
define

In the code

```
import torch
import torch.nn as nn

class LinearModel(nn.Module):
    def __init__(self):
        super(LinearModel, self).__init__()
        self.linear = nn.Linear(in_features=2, out_features=1, bias=True)

    def forward(self, x):
        return self.linear(x)
```

# Linear model in PyTorch

## Linear

```
CLASS torch.nn.Linear(in_features, out_features, bias=True)
```

[\[SOURCE\]](#)

Applies a linear transformation to the incoming data:  $y = xA^T + b$

### Parameters

- **in\_features** – size of each input sample
- **out\_features** – size of each output sample
- **bias** – If set to `False`, the layer will not learn an additive bias. Default: `True`

### Shape:

- Input:  $(N, *, H_{in})$  where  $*$  means any number of additional dimensions and  $H_{in} = \text{in\_features}$
- Output:  $(N, *, H_{out})$  where all but the last dimension are the same shape as the input and  $H_{out} = \text{out\_features}$ .

# Cost function : mean squared error

## Optimizer: stochastic gradient descent

Cost  
function  
+ optimizer

Cost function

In the code

```
reg_loss = nn.MSELoss()
```

MSELoss

```
CLASS torch.nn.MSELoss(size_average=None, reduce=None, reduction='mean')
```

[SOURCE]

Creates a criterion that measures the mean squared error (squared L2 norm) between each element in the input  $x$  and target  $y$ .

Optimizer

In the code

```
lr = 0.005  
optimizer = optim.SGD(model.parameters(), lr = lr)
```

```
CLASS torch.optim.SGD(params, lr=<required parameter>, momentum=0, dampening=0, weight_decay=0,  
nesterov=False)
```

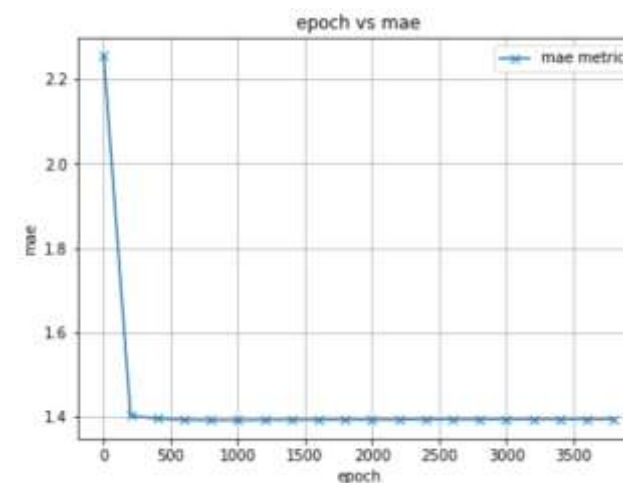
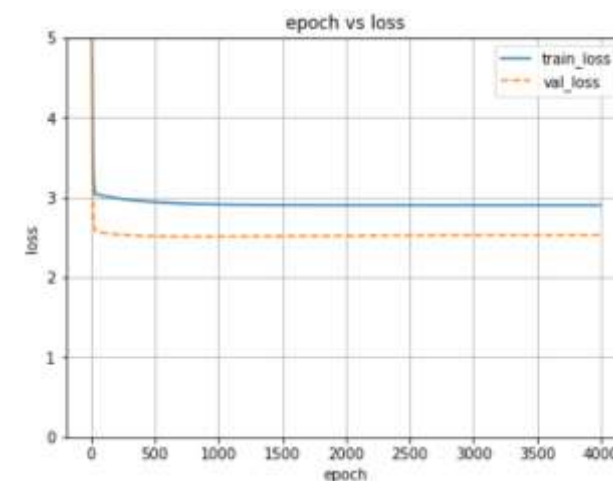
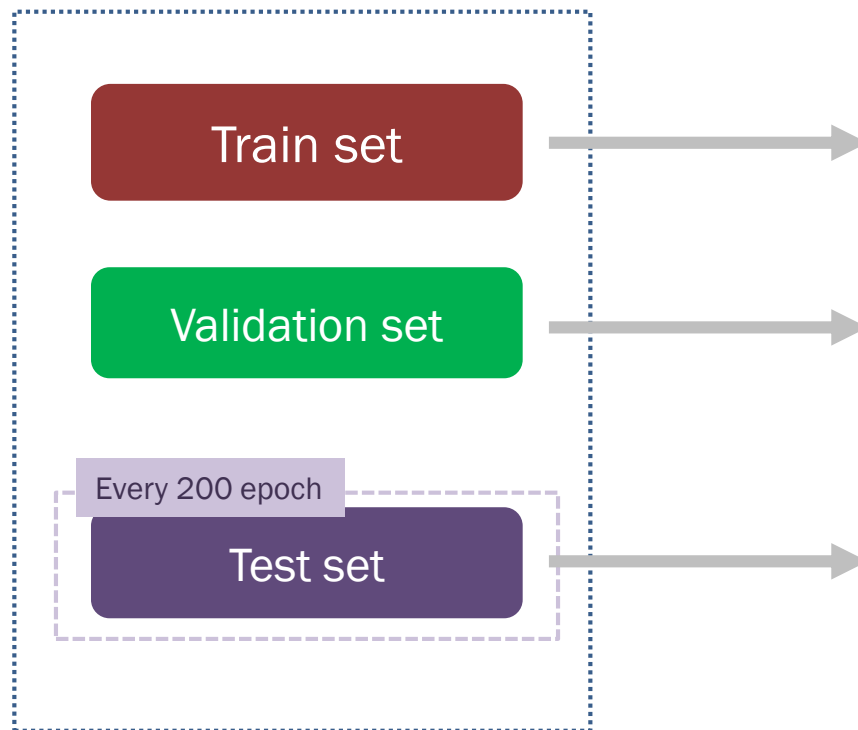
[SOURCE]

Implements stochastic gradient descent (optionally with momentum).

# Model test

Model  
Test

EPOCH=4000



# Lab 2A: Linear regression – multivariable

Exercise 1: different learning rate	Learning rate
Run 1	0.005
Run 2	0.05
Run 3	0.5

Exercise 2: w/ vs w/o bias	Bias
Run 1	Yes
Run 2	No

Exercise 3: different loss function	Learning rate
MSE	0.005
MAE (L1Loss)	0.005

<https://pytorch.org/docs/stable/nn.html#loss-functions>

## You may want to try

1. Increase size of data and re-run it
2. Use different optimizers

<https://pytorch.org/docs/stable/optim.html?highlight=optimizer#torch.optim.Optimizer>

**Break  
room**



# Running a DL code in Graham



*A consortium of 19 Ontario institutions providing advanced computing resources and support...*

**S**hared  
**H**ierarchical  
**A**cademic  
**R**esearch  
**C**omputing  
**NET**work



**compute**canada

- Member of Compute Canada and Compute Ontario
- 3,000+ Canadian and international users
- ~50,000 CPU cores
- 370+ GPUs
- 10 Gb/s network
- 100 Gb/s between national centres

# Virtual environment

Allows users to create virtual environments so that one can install Python modules easily

Many versions of same module are possible

```
[isaac@gra-login3 ~]$ module load python
[isaac@gra-login3 ~]$ module list

Currently Loaded Modules:
  1) nixpkgs/16.09      (S)    3) gcccore/.5.4.0    (H)    5) ifort/.2016.4.258 (H)    7) openmpi/2.1.1    (m)    9) python/3.7.4    (t)
  2) imkl/11.3.4.258  (math)  4) icc/.2016.4.258  (H)    6) intel/2016.4      (t)    8) StdEnv/2016.4    (S)

Where:
  S:  Module is Sticky, requires --force to unload or purge
  m:  MPI implementations / Implémentations MPI
  math: Mathematical libraries / Bibliothèques mathématiques
  t:  Tools for development / Outils de développement
  H:  Hidden Module

[isaac@gra-login3 ~]$ virtualenv --no-download ~/tf5
Using base prefix '/cvmfs/soft.computecanada.ca/easybuild/software/2017/Core/python/3.7.4'
New python executable in /home/isaac/tf5/bin/python
Installing setuptools, pip, wheel...
done.
[isaac@gra-login3 ~]$ source tf5/bin/activate
(tf5) [isaac@gra-login3 ~]$ deactivate
[isaac@gra-login3 ~]$
```

# Lab 2B – Working environment (Graham)

## Working environment in Graham

1. Log into graham.computecanada.ca with guest account and p/w : please see [\[this page\]](#) for further details.

(Use MobaXterm or Putty for Windows / Open terminal in Linux or Mac)

2. Load modules and make a virtual environment: please see [\[this page\]](#) for further details.

```
module load python
module load scipy-stack
virtualenv --no-download ~/ENV
```

3. Activate virtual environment and upgrade/install Pip and PyTorch: please see [\[this page\]](#) for further details.

```
source ~/ENV/bin/activate
pip install --no-index --upgrade pip
pip install --no-index torch
pip install --no-index torchvision torchtext torchaudio
pip install sklearn
```

4. (Optional) Deactivate virtual environment

```
deactivate
```

# Lab 2B – Running simple code



Break  
room

## Running a simple DL code in Graham

1. Clone the repository and change directory to Session\_2

```
cd /home/$USER/scratch/$USER  
git clone https://github.com/isaacye/SS2020V2_ML_Day2.git  
cd SS2020V2_ML_Day2/Session_2
```

2. Activate virtual environment (make sure you load python and scipy-stack module)

```
source ~/ENV/bin/activate
```

3. Run it

```
python SS20_lab2_LRm.py
```

Note that you may want to use a text editor (Nano/emacs/VI): Please see [\[Nano basic\]](#) for further details.

4. File transfer plotting files to your local computer using WinScp or MobaXterm (Windows) / sftp (Linux, Mac) and check it out

**Session break:**

**Please come back by 1:30 PM**