

Deep Learning.

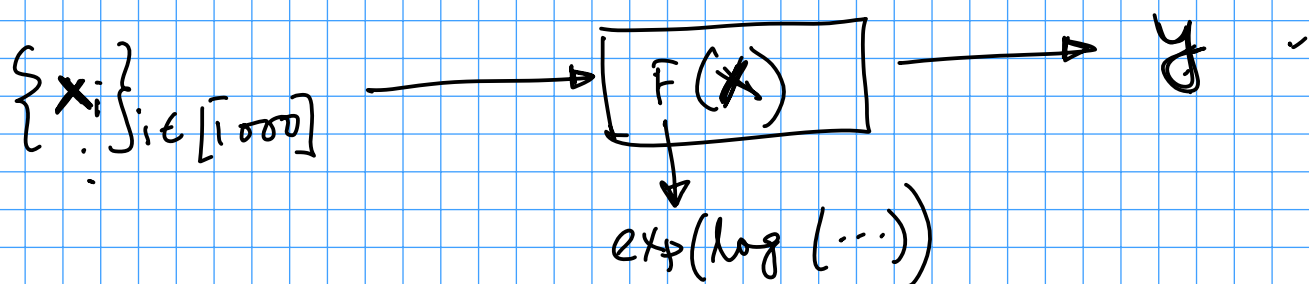
Today: 31/03 is overview of NN, DL.

→ 07/04 - 09/04 → Two guest lectures by Ashish Vaswani from Google.
 . 14/04 → DL.

Linear model for regression
 for classification

$$k(x_i, x_j) = \phi^T(x_i) \phi(x_j)$$

→ $(x_i, x_j) = k(x_i, x_j)$
 ↓
 +ve semi-definite



$$\min_w (y - w^T x)^2 \rightsquigarrow \text{poor error.}$$

$$\min_{\theta} (y - F_{\theta}(x))^2 \rightarrow \theta$$

2014 → TF becomes public.

$\frac{\partial f_{\theta}}{\partial \theta} \rightarrow \text{GD. / SGD / Adam.}$

Here will discuss about complex relationship between (y, x) , $\rightarrow y = F(x)$

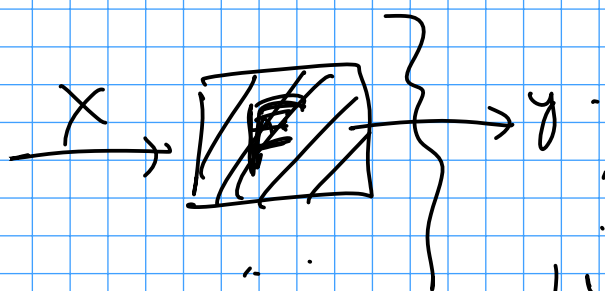
$$\min_{\theta} (y - F_{\theta}(x))^2 \rightarrow F_{\theta}(x) \approx F(x)$$

True F .

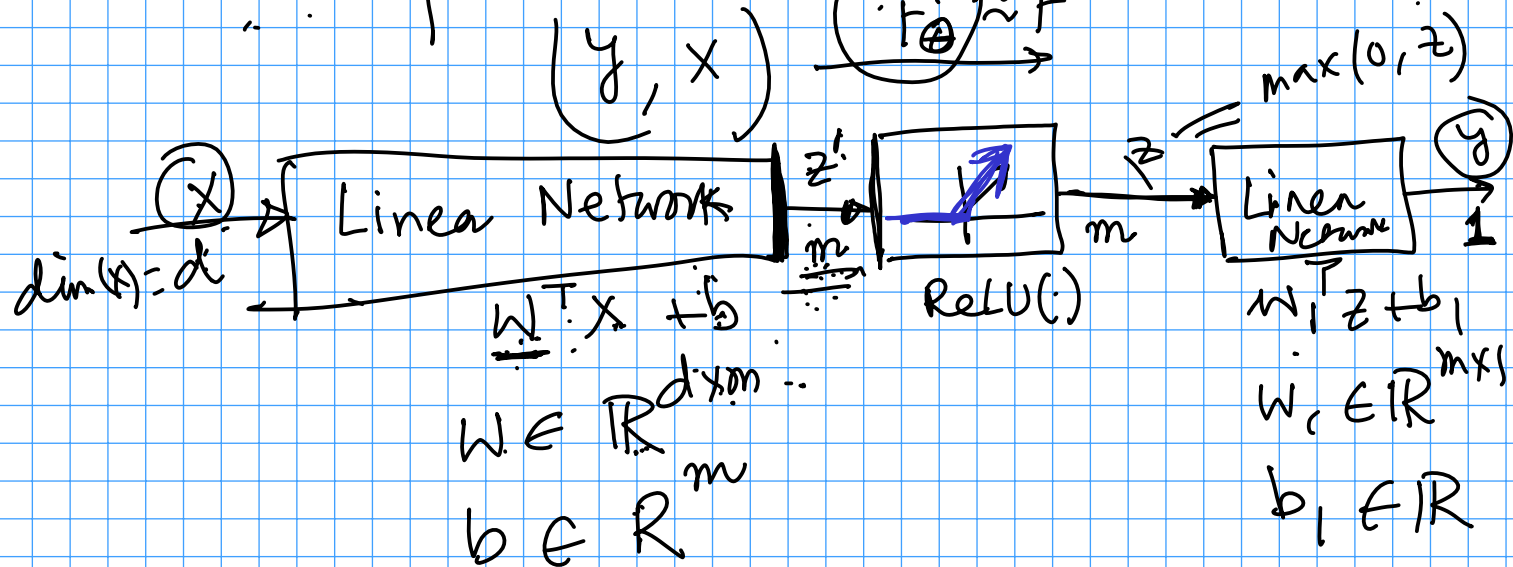
Challenge

① F can be complex \rightarrow ~~TF~~ PyTorch, GD, Adam.

② We don't know F .
we don't know if $F = \frac{\exp(\log(\dots))}{\dots}$
or $\sin(\exp(\dots))$
or $\text{sigmoid}(\dots)$

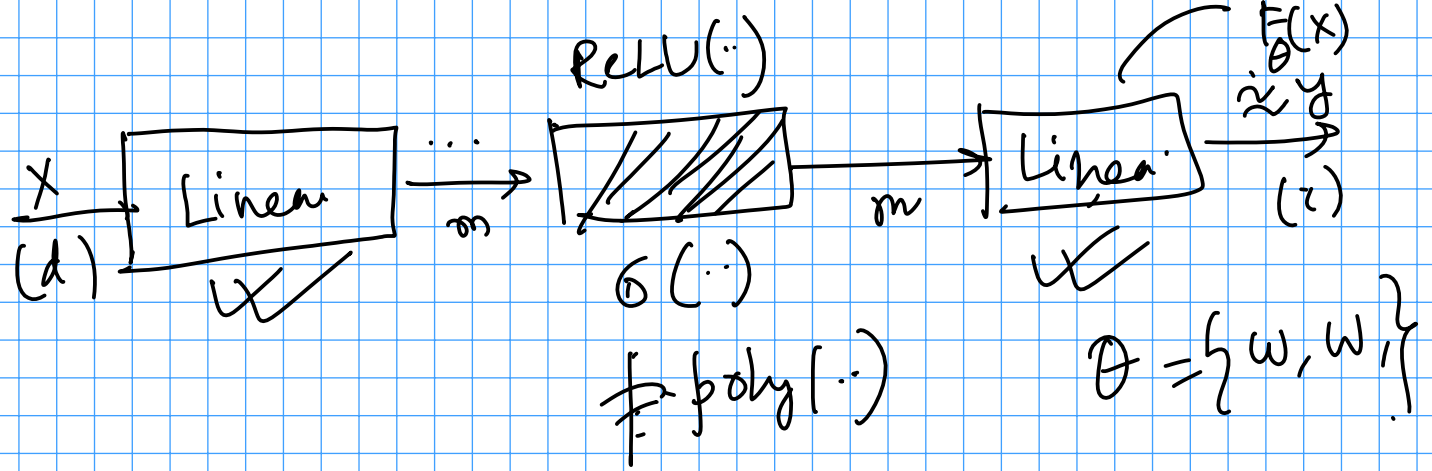


$$(y, x) \quad \text{where} \quad F_{\theta} \approx F$$

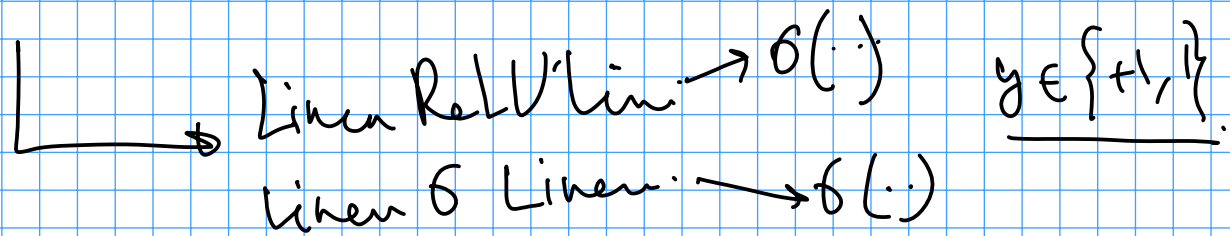
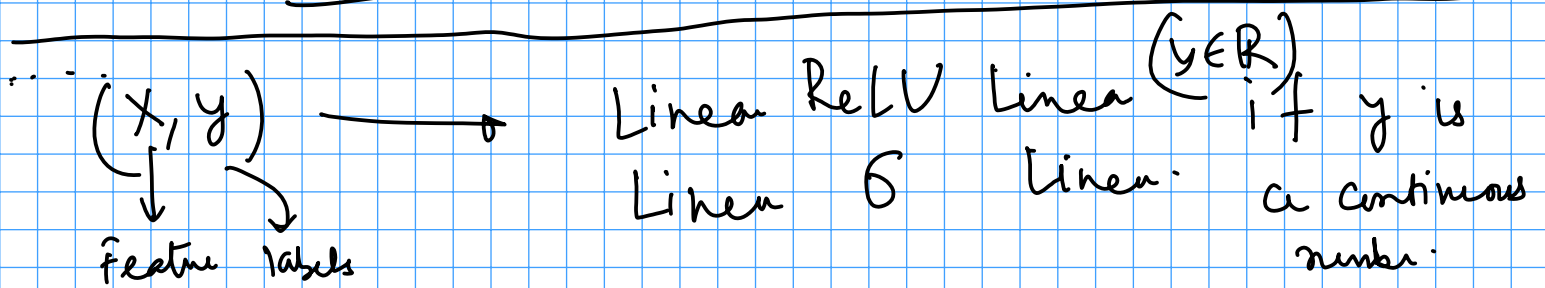


$$F(x) \approx F_{\theta}(x) ?$$

Linear + ReLU + Linear \approx any nonlinearity



$$F_{\theta}(x) \approx F(x)$$



$x_1, x_2, x_3, \dots, x_n$ sequence.

$y_1, y_2, y_3, \dots, y_n$ "

x_{i+1} can depend x_1, \dots, x_i $y_i \rightarrow x_i$

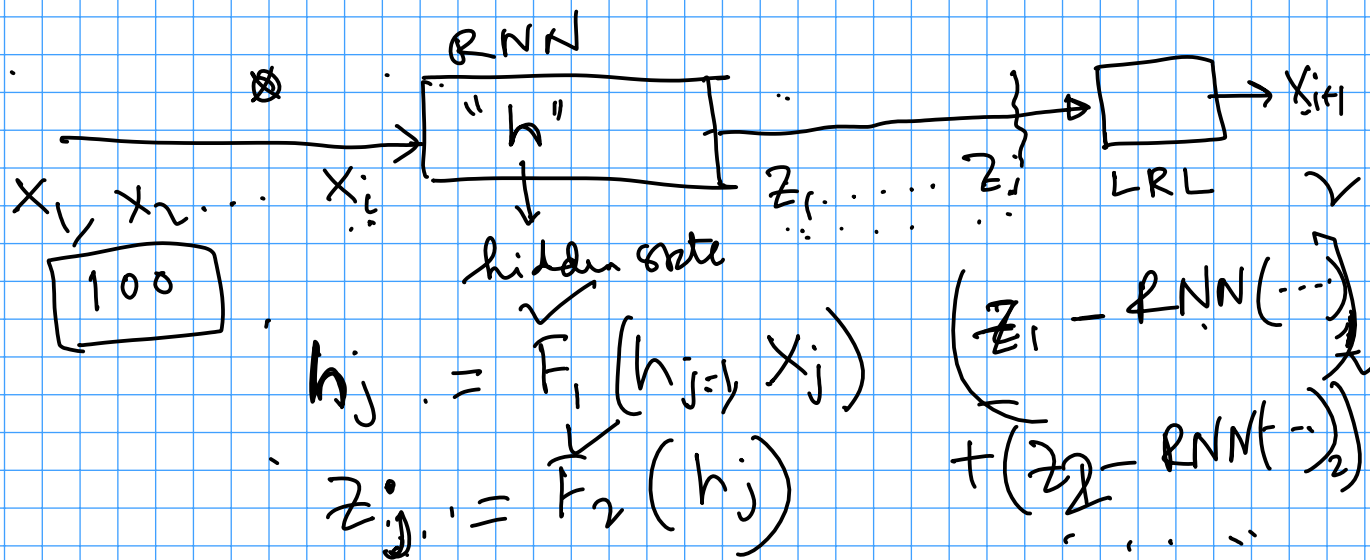
(i) stock market prediction

(ii) ~~stock~~ traffic forecasting

(iii) ~~stock~~ Dynamical system identification

$$\begin{matrix} x_1, x_2 \dots x_i & \xrightarrow{?} & (x_{i+1}, y_{i+1}) \\ y_1, y_2 \dots y_i & & \end{matrix}$$

Recurrent Neural Network.



* RNNs can simulate any algorithm
(Turing Machine)

How to optimize the training objective.

→ Back propagation:

$$\min_{\theta} (y - F_{\theta}(x))^2 \rightarrow L_{\theta}(x, y)$$

$$\begin{aligned} \theta_{k+1} &= \theta_k - \gamma \frac{\partial L_{\theta}(x, y)}{\partial \theta} \bigg|_{\theta = \theta_k} \\ &= \theta_k - \gamma \frac{\partial L_{\theta}}{\partial F_{\theta}} \cdot \frac{\partial F_{\theta}}{\partial \theta} \bigg|_{\theta = \theta_k} \end{aligned}$$

$(y - F_{\theta}(x, y))^2$

$$= \theta_k - \gamma \frac{\partial L_\theta}{\partial F_\theta} \frac{\partial F_\theta}{\partial \theta} \frac{\partial F_\theta(x)}{\partial \theta}$$

$$= \theta_k - \gamma \frac{\partial L_\theta}{\partial F_\theta} \frac{\partial (L_W R L_W)}{\partial L_W} \frac{\partial L_W}{\partial \theta} L_W R L_W$$

~~θ_k~~

$$\nabla_1 \boxed{L_1 R L_2 R L_3 R L_4}$$

$$\nabla_2 \boxed{L_1 R L_2 R L_3}$$

$$L_1 R L_2 R$$

$$\boxed{L(R(L))}$$

$$\theta_{k+1} = \theta_k - \gamma \frac{\partial L_\theta}{\partial \theta} \frac{\partial F}{\partial \theta}$$

$$\frac{\partial L_\theta}{\partial F_\theta} \frac{\partial F_1(F_2(F_3(F_4(x))))}{\partial \theta}$$

$$L(R(L))$$

$$\left(\frac{\partial L_\theta}{\partial F_\theta} \frac{\partial F}{\partial F_1(F_2(F_3))} \cdot \frac{\partial F_1(F_2(F_3))}{\partial F_1(F_2)} \right)$$