

Perfect 👍 You're planning it the right way, Rayhan. Since you'll give me the dataset later (1990–2022 stock data), I'll first give you a **structured flowchart/roadmap** that shows **how all the equations fit together into an ML + AI pipeline for Deepstock**.

Deepstock ML & AI Roadmap

1. Data Collection

- Historical stock data (1990–2022)
- Features: Open, High, Low, Close, Volume, Sentiment/Fundamental data

2. Data Preprocessing

- **Normalization:**

$$a_{scaled} = \frac{a_i - a_{min}}{a_{max} - a_{min}}$$

- **Anti-normalization (for output):**

$$\hat{y}_t = y_{t0} \cdot (y_{max} + y_{min}) + y_{min}$$

- **Feature Selection (Correlation):**

$$Corr(i) = \frac{cov(a_i, b)}{\sqrt{var(a_i) \cdot var(b)}}$$

3. Feature Engineering (Stock Technical Indicators)

- **SMA:**

$$SMA(t, N) = \frac{1}{N} \sum_{k=1}^N CP(t-k)$$

- **EMA:**

$$EMA(t, \Delta) = (CP(t) - EMA(t-1)) \cdot \Gamma + EMA(t-1)$$

- **MACD:**

$$MACD = EMA(t, k) - EMA(t, d)$$

- **OBV:**

$$OBV = OBV_{pr} \pm volume$$

- **RSI:**

$$RSI = 100 \cdot \frac{1}{1 + RS(t)}$$

4. Model Building (ML + AI)

Deep Learning Models

- **Neural Network / Perceptron:**

$$Z = W^T X + b$$

- **Activation Functions:** Sigmoid, ReLU, tanh
- **RNN:**

$$M_t = \tanh(W[STM_{t-1}, E_t] + b)$$

- **Autoencoder (Feature Compression):**

$$E(X, X') = \|X - X'\|^2$$

- **RBM (Pre-training):**

$$G(X, Y) = -\alpha^T X - \beta^T Y - X^T W Y$$

AI Algorithms

- **Backpropagation Update:**

$$W(n) = W(n-1) - \Delta W(n)$$

- **Extreme Learning Machine (ELM):**

$$Y_j = \sum_{i=1}^d \eta_i f(W_i X_j + e_i)$$

- **Radial Basis Function Network (RBF):**

$$y(x) = \sum_{i=1}^N \mu_i v_i(\|x - x_i\|) \quad y(x) = \sum_{i=1}^N \mu_i \nu(\|x - x_i\|)$$

5. Training

- Input: Technical Indicators + Normalized Data
- Output: Predicted Stock Price / Trend
- Optimization: Backpropagation, Gradient Descent

6. Evaluation

- **MSE:**

$$MSE = \frac{1}{n} \sum_{k=1}^n (y_k - \hat{y}_k)^2 \quad MSE = \frac{1}{n} \sum_{k=1}^n (y_k - \hat{y}_k)^2$$

- **RMSE:**

$$RMSE = \sqrt{\frac{1}{N} \sum_{t=1}^N (\hat{y}_t - y_t)^2} \quad RMSE = \sqrt{\frac{1}{N} \sum_{t=1}^N (\hat{y}_t - y_t)^2}$$

- **MAPE:**

$$MAPE = \frac{1}{N} \sum_{t=1}^N \frac{|y_t - \hat{y}_t|}{|y_t|} \quad MAPE = \frac{1}{N} \sum_{t=1}^N \frac{|y_t - \hat{y}_t|}{|y_t|}$$

- **Directional Accuracy (DA):**

$$DA = \frac{1}{N} \sum_{t=1}^N D_t \quad DA = \frac{1}{N} \sum_{t=1}^N D_t$$

Flowchart (Step-by-Step)

Data Collection → Preprocessing (Normalization, Feature Selection) → Feature Engineering (SMA, EMA, MACD, OBV, RSI) → Model Building (NN, RNN, CNN, AE, RBM, ELM, RBF) → Training (Backprop, Optimization) → Evaluation (MSE, RMSE, MAPE, DA) → Prediction Output

👉 When you give me the dataset, I'll help you **write the ML + AI code in Python (TensorFlow/PyTorch + Scikit-learn)** and integrate these equations step by step.

Do you want me to **draw this roadmap as a visual flowchart (diagram style)** so you can use it in your project report/presentation?