

Week-8

M T W T F S S

LI IOT (IIOT/Sensoriyaⁿ)

- IOT → A network of interconnected devices ranging from comp. & elec. gadgets to vehicles, all connected to the internet. The term came in 1990s.
Devices & prod. would eventually talk to each other & to network servers, in the interest of improving the performance.
- The IOT can be seen as connected to smart devices - watch, frdg., TV, bands, etc. It allows these devices to seamlessly exchange & consume data with minimal human intervenⁿ. Once data gets to the cloud, software processes it & then might decide to perform an acⁿ.
- 3 forms of internet - consumer, commercial & industrial. They have diff. target audience, TPs & strategies. Consumer market has market visibility.
- Commercial market has high marketability as they have services that compass financial & investment products. Enterprise IOT includes small, medium, & large businesses.
- By IIOT, we distinguish the assets that are connected, yet underlying idea is same.
- IIOT devices are wirelessly connected to the network.
- Obj. → ↑ efficiency & productivity thru automaⁿ.
- We want to collect comprehensive data from assets or processes. The data can be sent to

a cloud server for processing & decision making.

- IoT in Manufacturing - 11 IoT (use cases)
 1. Inventory → used with RFID, every inventory has its UID. RFID scans the tag & data extracted gets transmitted to cloud. Tracks locaⁿ, movement along supply chain, real-time status, better handling time
 2. Smart-Meter - monitors consumpⁿ of water, fuel, elec. → gauges for specific use & prevent over-usage of resources.
 3. Quality Control → (Pneumatics), use thermal & video sensors to collect prod. data, test at each manufacturing step to check for specificⁿ.
 4. Predictive Maintenance → monitor the operating equipment & perform analytics using related data to evaluate wear & tear.
 5. Industrial Safety → optimise safety of workers, equipment & opⁿ. Track KPIs. IoT measurable measure health metrics.
- 7 principles of IoT -
 1. Big Analog Data → convert digitalⁿ
 2. Constant Connecⁿ → monitor, maintenance, motioⁿ
 3. Real Time
 4. Spectrum of Insights
 5. Immediacy vs depth - quick analysis gives more insights, deeper results take time.

6. Shift left → Asset - Sensor → Server → Human
left

7. The next "V" → Visibility

L2 Time Series Forecasting

- 2 basic tasks → identify patterns, discover "varia^n" b/w variables of interest. The common method of identifying pattern is to decompose the data into components — trend, seasonal, cyclic & random.

- Exponential Smoothing

→ F_t forecast of t , D_t data of t .

$$F_{t+1} = \alpha D_t + (1-\alpha) F_t$$

→ F_1 → can be 1st value, last, or avg.

→ for large datasets, impact of F_1 is smaller.

→ value of α depends on demand "varia^n" (smaller α if "varia^n" is large)

- Time Series Model (+ Trends)

$$F_t = a + bt + \varepsilon$$
 use least SSE

→ linear reg. model — considers all data pts.
& gives equal wt. to the past history.

- Double exp. smoothing

→ ES + linear trend model

→ when exponential fit is applied to data with linear trend, we know that forecast lags behind the most recent value.

$$\text{Lag} = D_t - F_t = \frac{1-\alpha}{\alpha} \times b$$

- Holt's Method for trend

$$F_{t+1} = a_t + b_t \quad \begin{matrix} \xrightarrow{\text{level}} \\ \xrightarrow{\text{slope}} \end{matrix}$$

$$a_t = \alpha D_t + (1-\alpha)(a_{t-1} - b_{t-1})$$

$$b_t = \beta(a_t - a_{t-1}) + (1-\beta)b_{t-1}$$

$$a_1 = D_1, \quad b_1 = \frac{D_f - D_i}{n-1}$$

- Winter's model for seasonality

$$F_{t+1} = (a_t + b_t) \times C_{t+1}$$

fixed level. trend comp. seasonality index

$$a_{t+1} = \alpha \frac{D_{t+1} - (1-\alpha)(a_t + b_t)}{C_{t+1}}$$

$$b_{t+1} = \beta(a_{t+1} - a_t) + (1-\beta)b_t$$

$$C_{t+1} = \gamma \frac{D_{t+1}}{a_{t+1}} + (1-\gamma)C_{t+1}$$

- AR(1)

$$\hat{D}_t = D_t - \mu \Rightarrow \mu = E[D_t]$$

$$\text{AR}(1) \rightarrow \hat{D}_t = \phi_1 \hat{D}_{t-1} + \varepsilon_t$$

- MA(1)

$$F_t = \mu - \Phi(\varepsilon_{t-1})$$

$$\varepsilon_{t-1} = D_{t-1} - F_{t-1}$$

| M | T | W | T | F | S | S |
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- ARMA(1)

$$\hat{D}_t = \phi_1 \hat{D}_{t-1} - \theta (\varepsilon_{t-1}) + \varepsilon_t$$

L3 Excel Based Forecasting

- automatic paid solvers for excel.