

MU SIGMA

DO THE MATH

Chicago, IL Bangalore, IN

www.mu-sigma.com

The Art of Networking

by Prabakaran Chandran

Thursday Learning Hour 23 June 2022

Proprietary Information

"This document and its attachments are confidential. Any unauthorized copying, disclosure or distribution of the material is strictly forbidden"

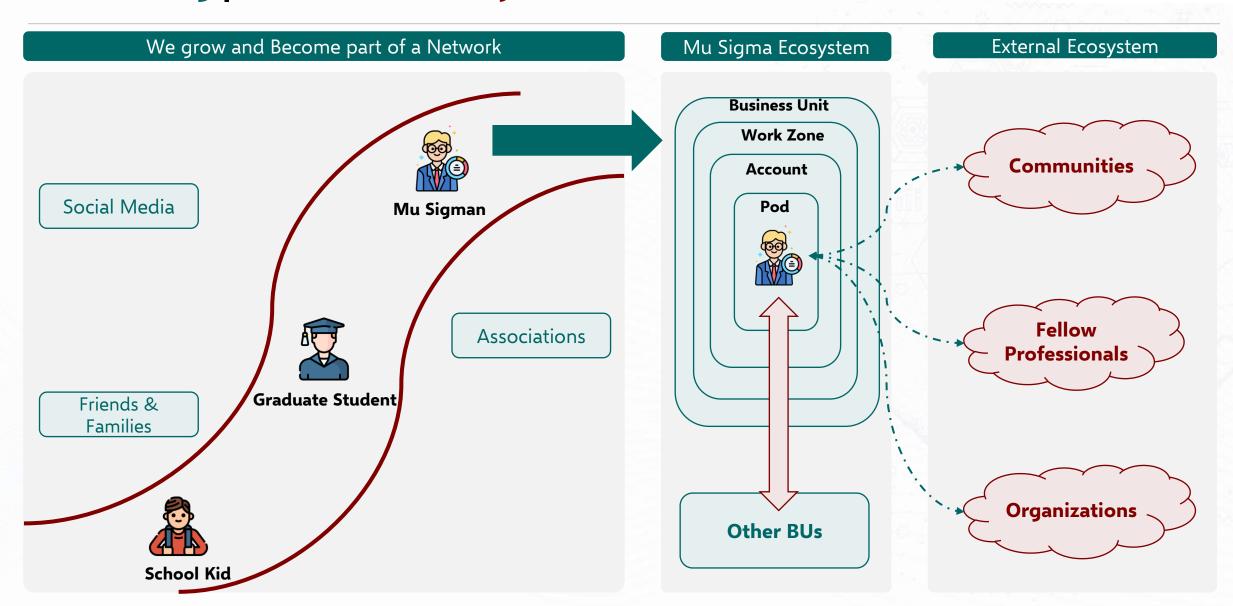


What are we going to learn today?

- 1. What is Networking?
- 2. Why is it Important?
- 3. Towards Efficient Networking?
- 4. Using the Channels Wisely
- 5. Networking for Techies and Business Folks

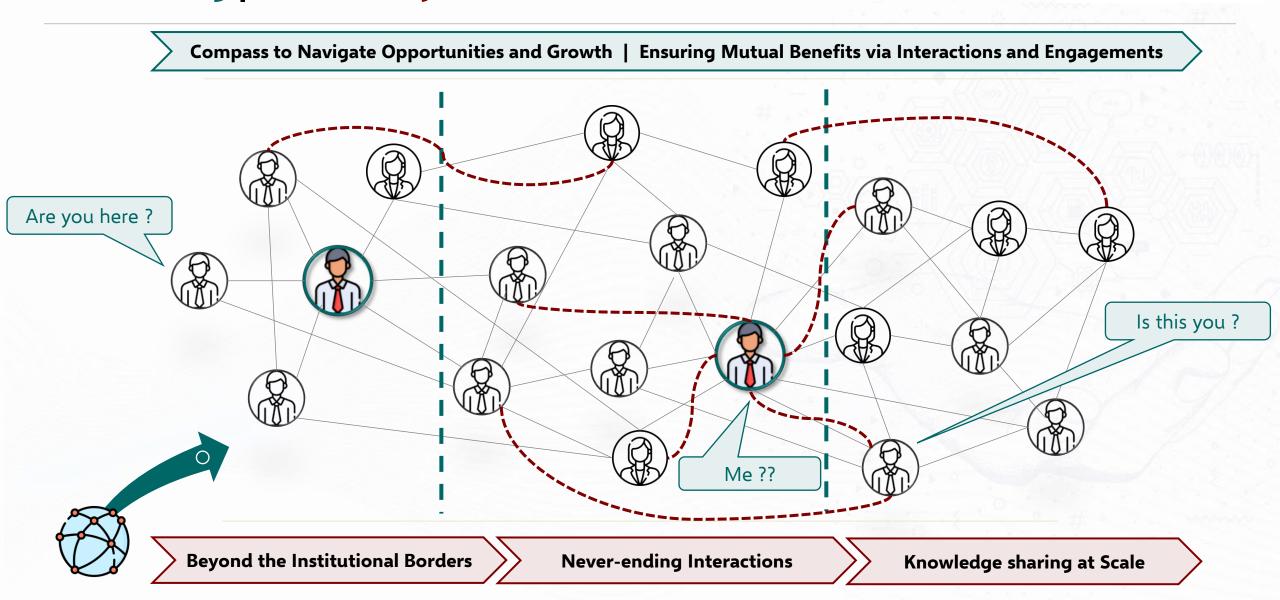


Networking | What is it Actually?



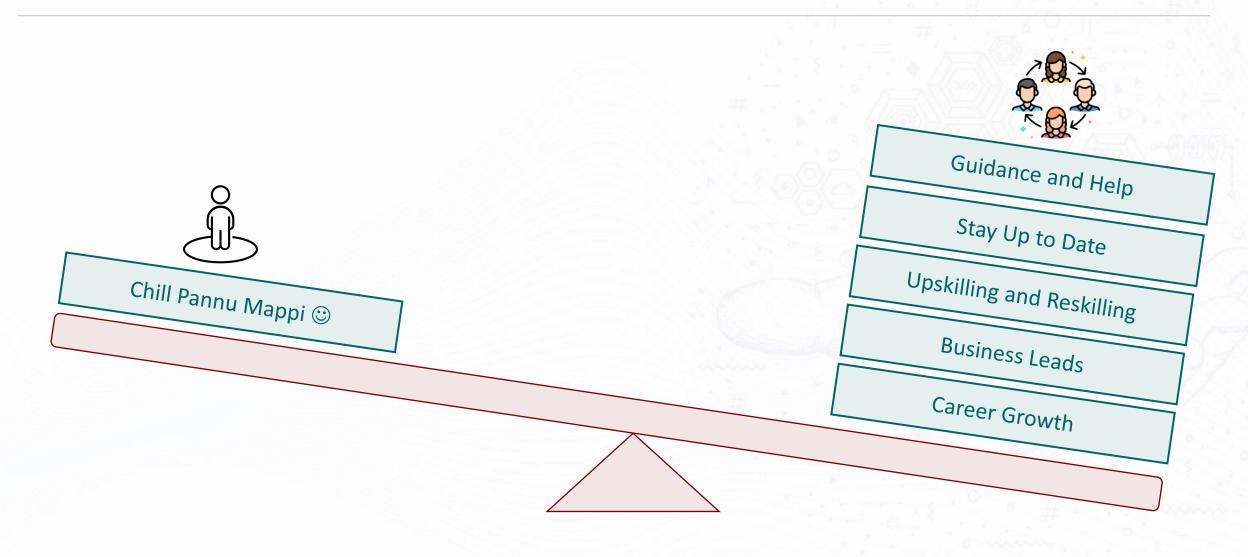


Networking | Where are you in the Network?



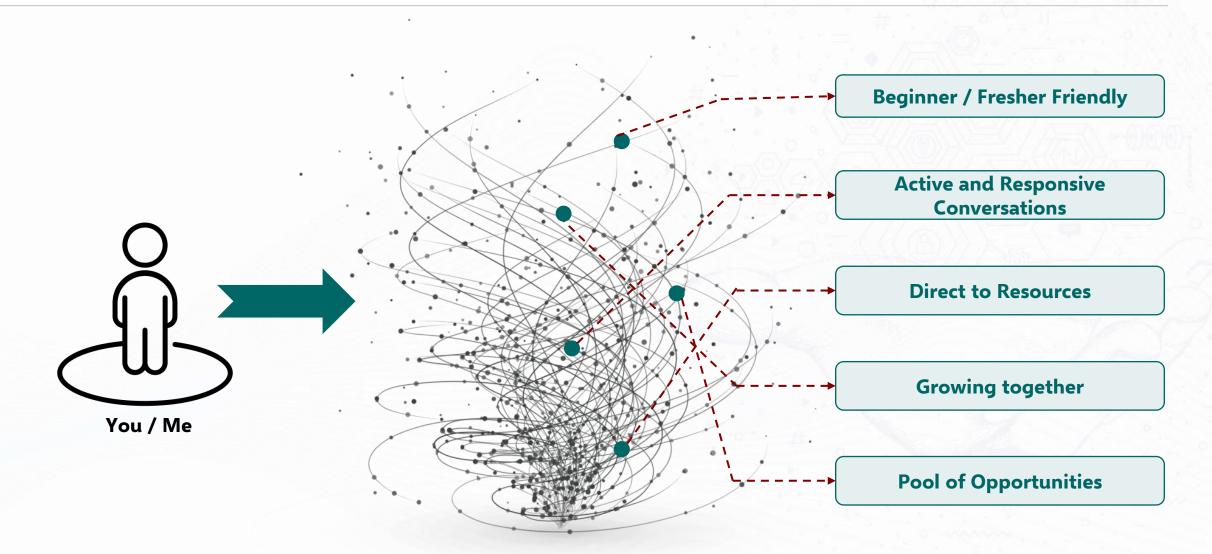


Networking | Why is it so Important?



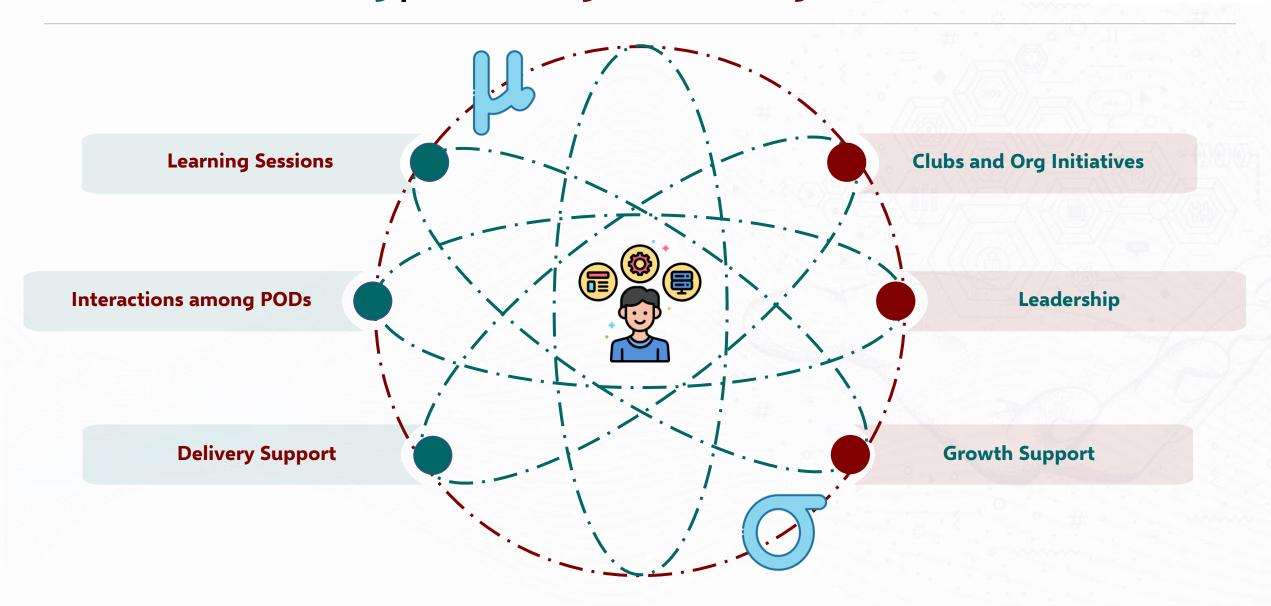


The Art of Networking | Do you want to be in a Better Network?



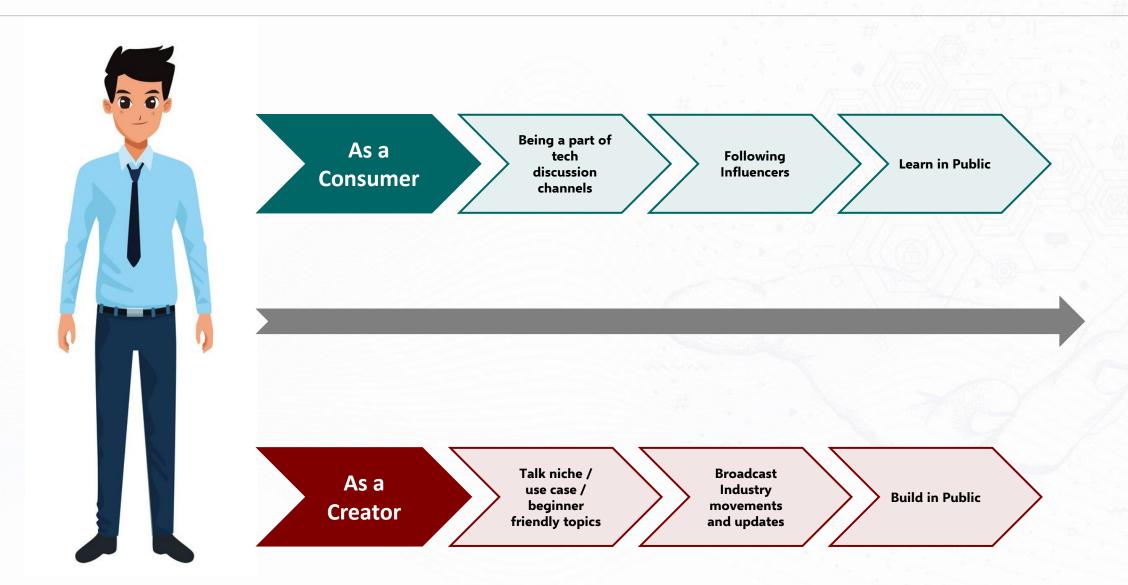


The Art of Networking | Networking inside Mu Sigma



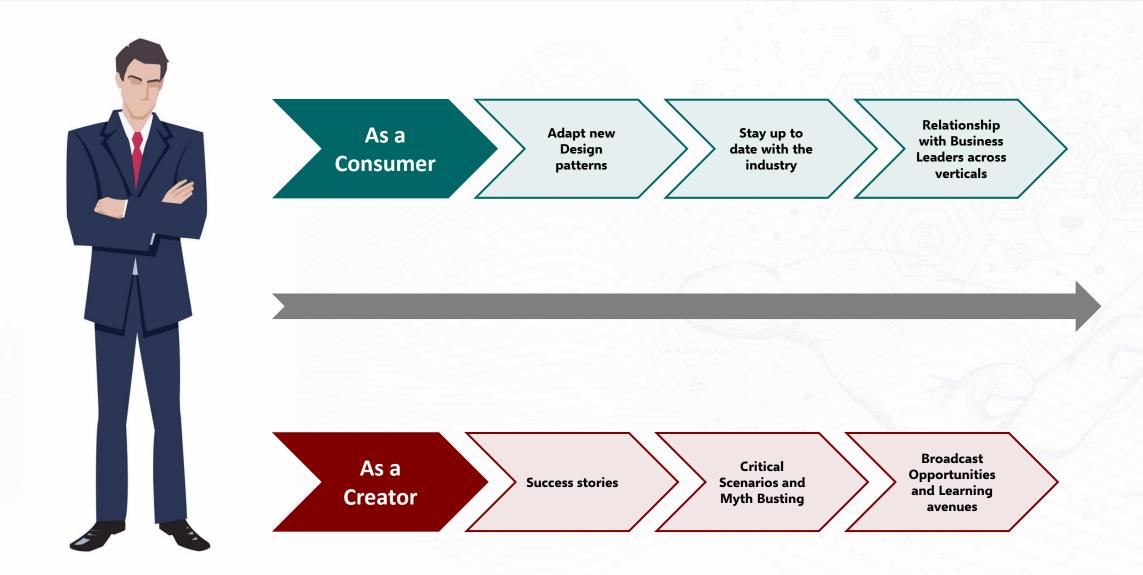


The Art of Networking | Networking as a Decision Scientist



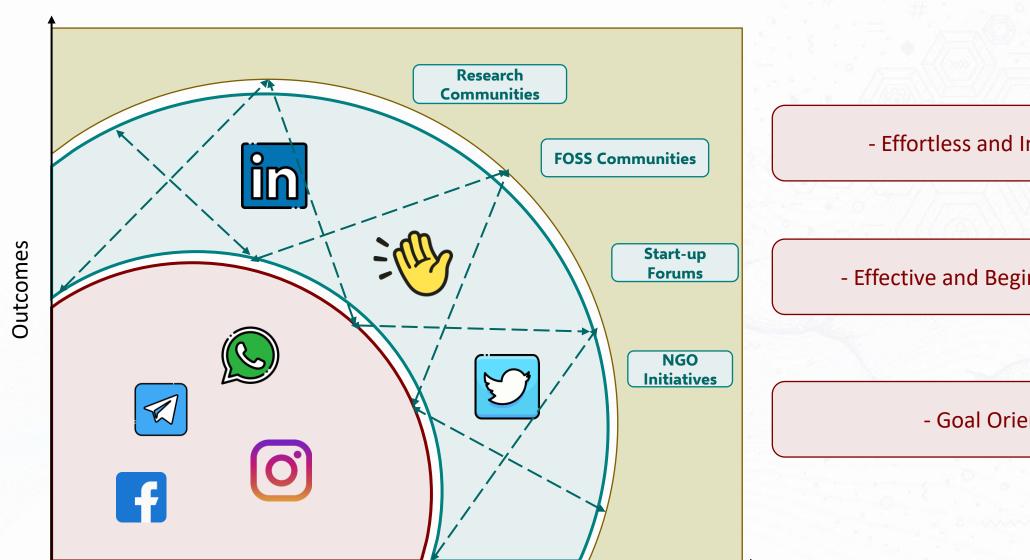


The Art of Networking | Networking as an Apprentice Leader





The Art of Networking | What are the different Channels?



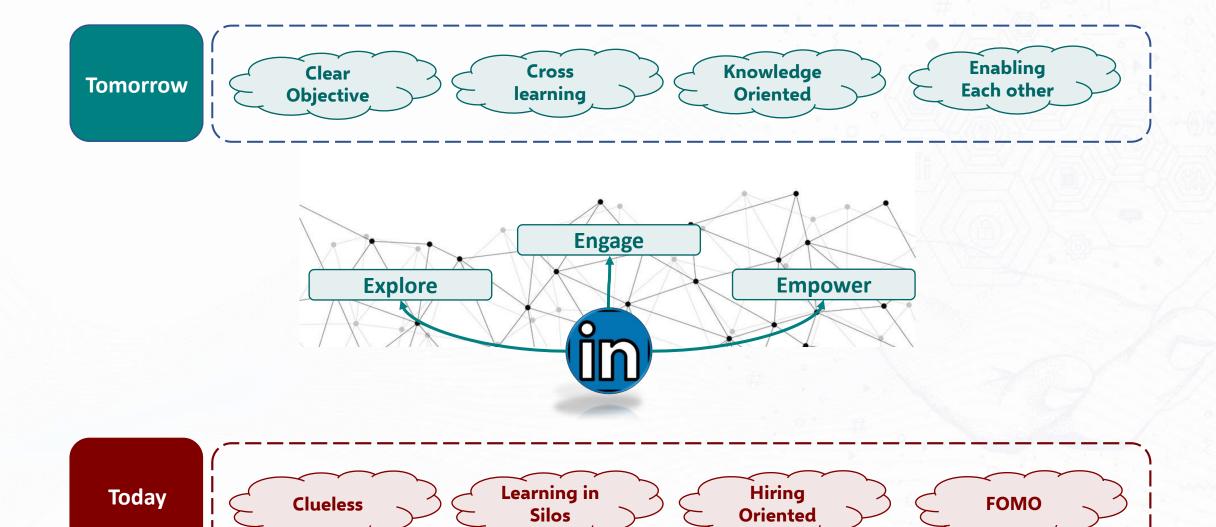
- Effortless and Impact less

- Effective and Beginner Friendly

- Goal Oriented

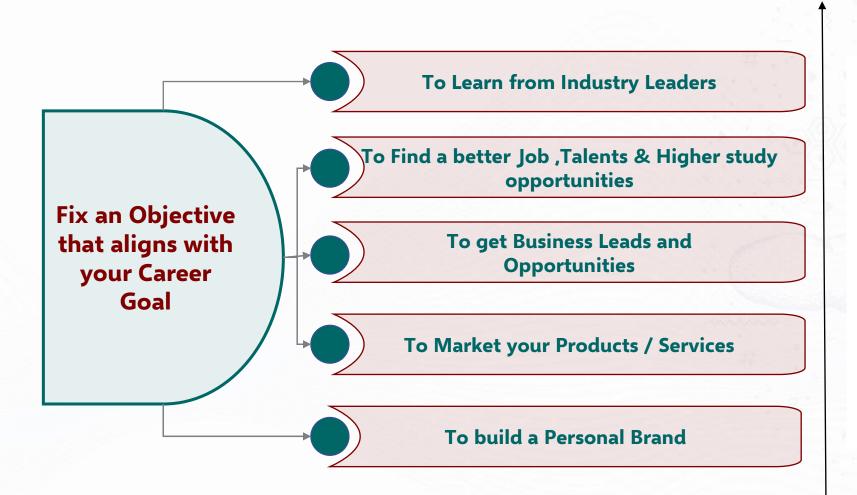
Mu Sigma

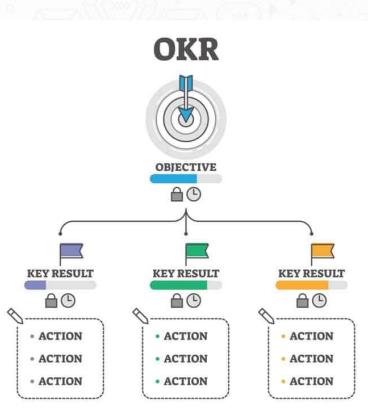
The Art of Networking | How can we use LinkedIn for Effective Networking?





LinkedIn Networking | Fix your Objective







LinkedIn Networking | Decide Your Audience



Students and Beginners



Practitioners & Working
Professionals



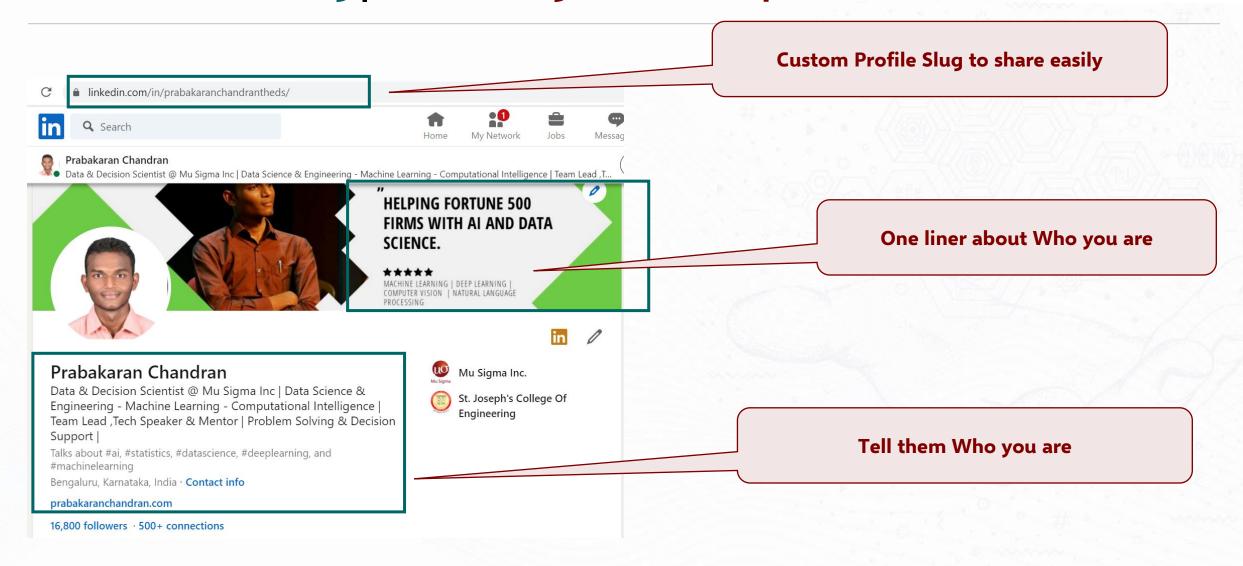
Academia & Research



Business Leaders & CXOs

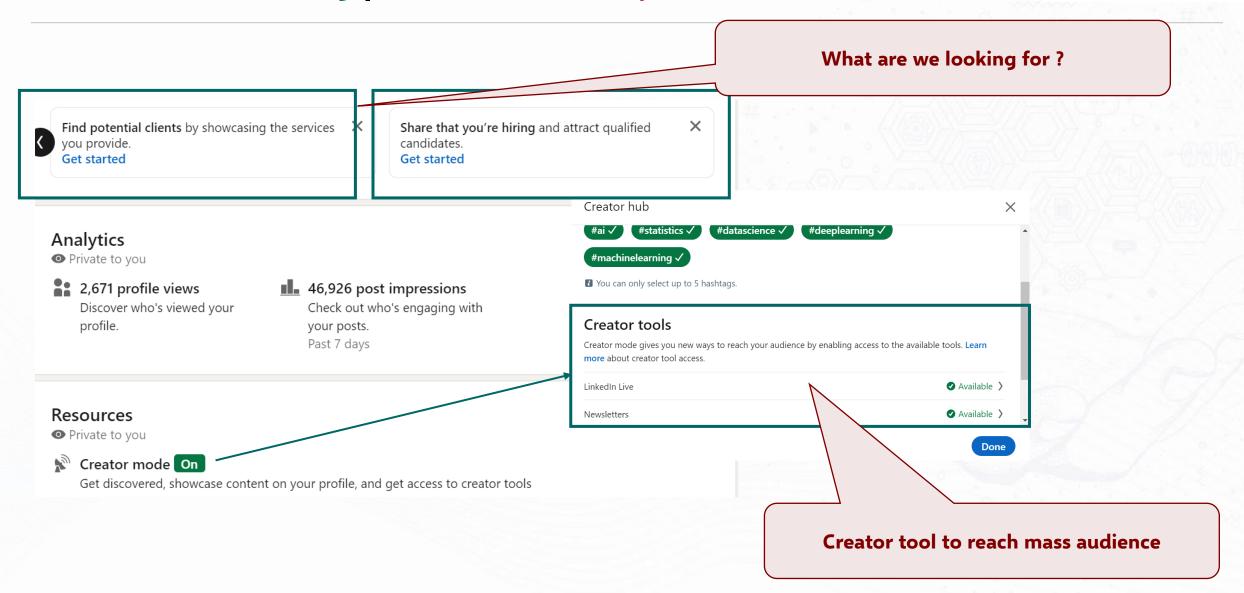


LinkedIn Networking | Profile brings the best Impression





LinkedIn Networking | Talk more about you





LinkedIn Networking | Talk more about you

Data Scientist (Trainee Decision Scientist - Level 2)

Full-time

Jun 2021 - Nov 2021 · 6 mos India

Al powered Product Development

1. Real time in-situ process monitoring and Simulation of Additive manufacturing process using Generative Neural Modeling

Trainee Decision Scientist - Level 1 - Data Science and Machine Learning Engineering

Full-time

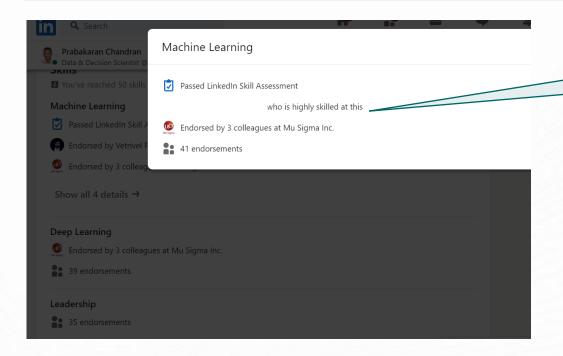
Jun 2019 - Jun 2021 · 2 yrs 1 mo Bengaluru, Karnataka, India

- 1. End to End Machine learning pipeline creation for Japan-based Construction Giant.
- Monthly payout prediction for an each Construction project based on Various Machine learning Algorithms, such as Random forest regression, Ensemble learning, Boosting and Stacking methods.
- We have implemented this pipeline start from Data manipulation, Feature Engineering, and Variable selection in order to decrease the RMSE and to get an optimized model
- The technology used: Python and its various DS and ML packages
- 2. The descriptive analytics web application for Japan-based IT firm
- I have built an R shiny based Web application for this firm to understand about their server failures, shipments, and Repairs

Talk about your experience in detail



LinkedIn Networking | Endorsements & Recommendations



Getting Recommendations from Managers / Clients

Getting Endorsements from Highly skilled Folks

Received

Given



Daniel Nehemiah C in · 1st

Data Scientist at Societe Generale Global Solution Centre

August 13, 2020, Daniel Nehemiah managed Prabakaran directly

Prabakaran's knack for learning new and advanced concepts is really something impressive! It was really nice having such a technically sound person in the team and people wouldn't even believe if they heard that a person with just 1 year of work experience is able to contribute so much to the team! All the best Prabakaran! Keep learning and keep teaching!



Gautam Rajeev - 1st

Data Scientist

August 2, 2020, Gautam managed Prabakaran directly

I worked with Prabhakaran for nearly a year on a critical advanced analytics engagement with one of the biggest CPG firms in the world. Despite being the youngest in our team, his immense curiousity and passion for machine learning made him the person we all relied on to drive solution design and come up with innovative ideas that became the selling point for our work. He is extremely meticulous with an in: ...see more



Vinayak Menon · 1st

MBA Candidate at Columbia Business School July 27, 2020, Vinayak managed Prabakaran directly

Prabakaran very early in his career has amassed significant knowledge in the ML space which he is



LinkedIn Networking | Content is the Catalyst

When I started learning and practising Data Science, I took more time to study and understands the statistical concepts which eased me to interpret the ML, DL results.

I used to come up with different statistical methods and Hypotheses tests to solve the problem given.

My mentors always encouraged me to explore more statistical terms and discussed it with me.

I would prefer these Statistical Resources to kick start your learning:

- Head First Statistics it's like a comic book, explained with intuitive examples, covers very basics of Stats (I guess till ANOVA)
- Introduction to Statistical Learning with R (ISLR) it has all the methods and their R implementation - (from Linear regression to Dimensionality Reduction)
- 100 statistical tests consists of 100 important hypothesis tests (beyond Z,t,f
- if you still want an easier approach with better explanation- stat quest (youtube channel) by Joshua Starmer

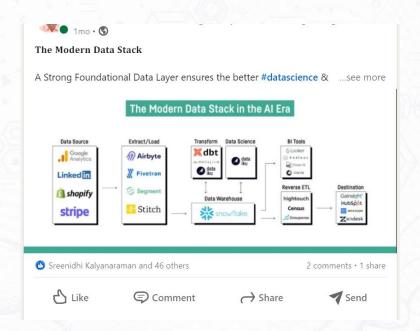
#statistics #statsticallearning #rprogramming #islr #learnwithKaran

€0 You and 130 others

6 comments · 1 share



Sharing Resources



Talking about Emerging / **Niche topics**

Learning Experience



LinkedIn Networking | Content is the Catalyst



Prabakaran Chandran • You

Data & Decision Scientist @ Mu Sigma Inc | Data Science & Engineering - M...

1mo • Edited • 🚱

3 important things while you deliver and deploy a **#machinelearning** product /pipeline

- 1. Model Explainability that can enable business to take decisions efficiently **#XAI #interpretation**
- We can use SHAPE, LIME, Explainable Machines
- 2. Prediction Intervals (check the concept of **#conformalprediction**) that ensure Precise levels confidence
- The field of Conformal prediction is being adapted into industry use cases these days.
- Post #Deployment Monitoring for the seamless business utilisation Which can mitigate pitfalls such as Model decay, data drift.
- Post Performance estimation, ML based Data drift Monitoring can be utilized.

The Deep learning space for Tabular Data is interesting and a lot of SOTA papers are being published.

TABNet paved a path for Transformers to be used in solving Tabular Data Problems

The Interesting fusion would be Boosted Trees and Deep Learning Architectures.

Manu Joseph's PyTorch Tabular is a go to Framework for easier Implementation of Deep learning models for Tabular Data.

I personally liked the structure of the framework and adapted the flow for few of my personal projects.

#Framework provides following Models:

FeedForward Network with Category Embedding

Neural Oblivious Decision Ensembles https://lnkd.in/gcSb4FH5

TabNet https://lnkd.in/gJbNdSgX

Mixture Density Networks

AutoInt https://lnkd.in/gJbNdSgX

TabTransformer https://lnkd.in/gkBMCRZv

FT Transformer https://lnkd.in/gaU2zwWR

Research Papers

Above all, Serving at scale is the important thing to be considered here. The model size is optimized with the help of the Microsoft Hummingbird framework.

Models need to serve millions of API calls. To make this happen FastAPI based framework is built to have optimized latency, throughput over MLFlow.

With the help of an improved model and a New ML serving System, the overall cost of serving millions of calls was less than \$10 (approx. INR 754.57)

To learn more things in detail check the case study here https://lnkd.in/gRgiRU3U

Image: Zomato's ETA estimation model serving framework

#machinelearning #customersatisfactionguaranteed #zomato
#machinelearningengineer #systemdesign #casestudy #lgboost #decisiontree
#datascience #analytics #artificialintelliegence #deeplearning

ML Serving Design

Processed Features

Professed Features

Professed Features

Professed Features

Professed Features

Professed Features

Fig. Date

Con Indicate

Fig. Date

Con Indicate

Fig. Date

Con Indicate

Fig. Date

Con Indicate

Fig. Date

Fig

Studies on Real time use cases

Sharing Design tips



LinkedIn Networking | Build Relationships

Follow / Connect with Influencers

- To Know more about specific topics such as NLP / UI-UX / GAI
- To Learn effective ways to build complex projects

Attract Connections through content

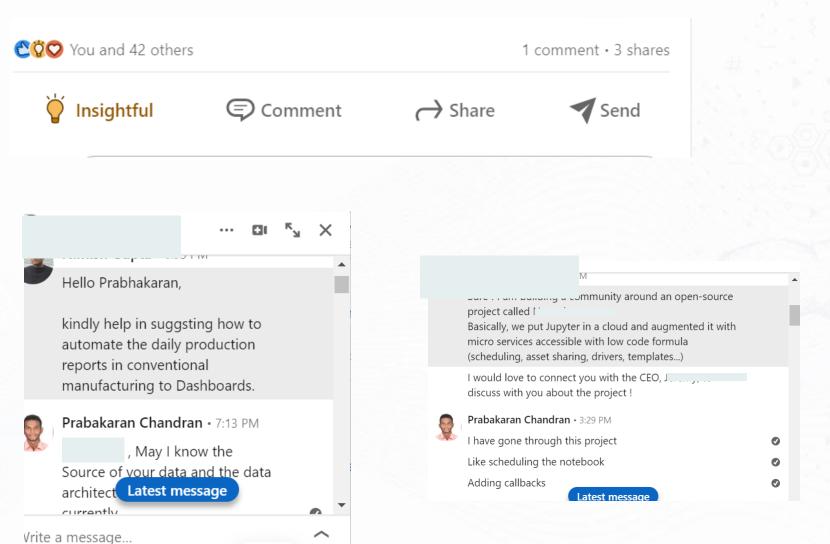
- Showcase capabilities to attract leads
- Beginner level contents improve the outreach
- Niche Content brings the collaboration Opportunities
- Learn in Public / Build in Public

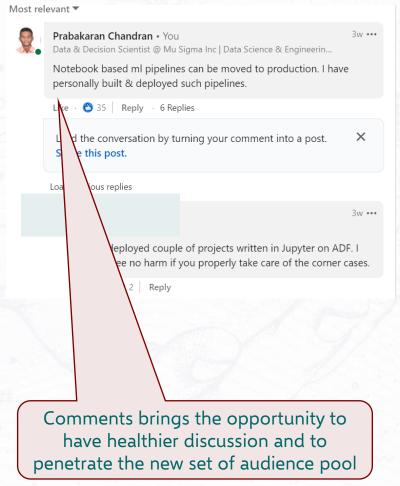
Collaborate with Connections / Institutes

- Co Authoring Papers
- Hackathons
- FOSS
- Community Projects



LinkedIn Networking | Engagement is the Node of Attraction







Enablers of Networking | Beyond LinkedIn



Communities and Meetups

- TensorFlow User Group
- Google Developers
 Group
- Al Coimbatore
- The Hack Weekly
- Kaggle Meetup
- FOSS Communities



Chat Rooms

- Club House
- Twitter Spaces
- Discord



Learn & Build in Public

- Showcases your progress
- Brings the visibility in to the Product
- Cost FreeMarketing andBranding



The Twitter Cult

- Networking beyond Race / Gender / Geography
- Better Conversion Rate



Enablers of Networking | Beyond LinkedIn



Blogs

- To Drive More Traffic
- To Validate your learnings
- To Enable others



Learning Sessions

- Mass Reach
- Networking with All sorts of students and Professionals



Bonanza: LinkedIn Profile Enhancement



Bonanza: Your First LinkedIn Post

$$Y_{i+1} = Y_i + b \cdot K_2$$
 $B = \begin{pmatrix} 2 & 1 & -1 & 0 \\ 3 & 0 & 1 & 2 \end{pmatrix}$

$$\sum_{i=1}^{n} (p_2(x_i) - y_i)^2 + y_2 = \frac{2t_9 x}{1 - t_0^2 x} + \frac{t_9 x}{cos x}$$

$$\sum_{i=0}^{\infty} (p_2(x_i) - y_i)^2 + 92x = \frac{24yx}{1 - 4y^2x}$$

$$\int_{1=0}^{2\pi} \left(\frac{1}{1-t} \right)^{-1/2} \int_{1-t}^{2\pi} \frac{1}{1-t} \frac{1}{2}$$

$$yolz = \int_{0}^{2\pi} \left(\int_{0}^{2} \left(\int_{1}^{1} r \, n \, dr \right) \, dr \right) \, dr$$

$$\int_{0}^{\infty} \left| \int_{0}^{\infty} \left| \int_{\frac{1}{2}}^{\infty} \left| \int_{0}^{\infty} \left| \int_{0}^{\infty}$$

$$\cos^2 k = 7$$
 $\delta(p_z) = \sqrt{0.16}$

$$f(x) = 2^{-x} + 1, \varepsilon = 0.005$$

$$\lim_{x \to 0} \frac{e^{2x} - 1}{5x} = \frac{2}{5}$$

$$\begin{array}{c} X = \begin{pmatrix} 2p \\ -p \\ 0 \end{pmatrix}$$

$$\int_{-\infty}^{\infty} \frac{(1+e^{-1})^{2}}{\cos 2x} = \cos^{2}x - \sin^{2}x$$

$$A+B+C=8$$

 $Sin^2x + COS^2x = 1$ $-18A+6B-3C=19$

$$n^2 \times + \cos^2 x = 1$$
 -18A+6B-3C=15

$$\frac{\sin x}{x} \leq \frac{x}{x} = 1$$

$$\gamma_1 = \lambda_1^2 - 3\lambda_4 + 1 \neq 0$$



Thanks for Joining!