# Senthil Kumar

Applied NLP Data Scientist

Competent in the use of ML and DL for NLP tasks Proficient in building clean, modular and dockerized Python **applications** Of the total 11+ years of experience, 9.5 years of individual contribution and 1.5 years of people management Detailed Profile



senthilkumar.m1901@gmail.com

+91 984-186-9609/ +91 944-512-3824

Chennai, India

linkedin.com/in/senthilkumarm1901

github.com/senthilkumarm1901

# **WORK EXPERIENCE**

### **Deputy Manager**

Ford Analytics Division

05/2018 - Present

Roles

- Senior Analyst [1.5 Years] | Deputy Manager [2+ Years]
- A data science developer who employs state-of-the-art ML and DL techniques for NLP Applications
- Experienced in end-to-end ML application development
- -- from data acquisition, cleaning, labeling and preprocessing,
- -- to model development, deployment and maintenance
- Python/NLP Trainer | Technical Interviewer of NLP candidates across analytics teams

### **Assistant Manager**

LatentView Analytics

04/2014 - 04/2018 Chennai, India

Roles

- Individual contributor [2.5 years] | People Manager [1.5 years] | Social Media Analysis | NLP Projects
- LinkedIn Recommendation: "... extraordinary dedication contributed significantly to growing our analytic practice..." F100 Tech Client stakeholder
- LinkedIn Recommendation: "....Sincere, driven, articulate and utterly committed ... " Skip-level Reporting Manager at LatentView

#### Senior Consultant

Capgemini

01/2014 - 03/2014 Bangalore, India

#### Lead Analyst

Beroe Inc - A Procurement Intelligence Firm

07/2010 - 12/2013 Chennai, India

Achievements/Tasks

**Python Libraries** 

(extensive usage)

- LinkedIn Recommendation: " ... well organized, innovative ... and always ready to go the extra mile to support the client ... " - Client Engagement Manager

### **EDUCATION**

B.E - Electronics & Communication - 8.6 CGPA

Madras Institute of Technology, 2006 - 2010

Deep Learning Specialization

Deeplearning.ai-Coursera (5 courses), Dec 2018 - May 2019

Applied ML and Applied Text Mining 🗗

University of Michigan - Coursera, Dec 2017 - Jan 2018

12th Grade - 95% | 10th Grade - 92%

State Topper in Physical Science paper of 2006 TN Engineering Entrance Exam

GCP Big Data & ML Fundamentals 🗗

Google - Coursera, Apr 2021

Probability and Statistics Fundamentals 🗗

LinkedIn Learning, Dec 2021

# **TECHNICAL SKILLS**

Python (moderate to advanced), Markdown, Languages

Linux Shell (basics), SQL (basics)

Pandas, SpaCy, Re (Regular Expressions),

Transformers, Sklearn, PyTorch

Tools

(working

knowledge)

**Python Libraries** 

Git, WSL, Docker, Kubernetes, Poetry (Python

env), Conda, PyCharm

PySpark, FastAPI (REST API), Streamlit (UI), Altair (viz)

Page 1 of 2

### PROMOTIONS AND AWARDS

#### Ford: Promoted from Senior Analyst to Deputy Manager

In Nov'2019, after 1.5 years of joining Ford

#### LatentView: Promoted from Senior Analyst to Assistant Manager

In Oct'16, after 2.5 years of joining LatentView

#### Beroe: Promoted twice in my first comapny

During my 3.5 year stint in Beroe

#### Ford Asia-pacific Recognition Award

won in May 2019 for successful spearheading of a project

#### LatentView Analytics - Encore Award

won for company-wide best performance for the Jul-Sep 2016 quarter

#### Beroe - Knowledge Contributor Awards

Won twice for company-wide best performance in Q1 and Q2 calendar year 2013

# **SAMPLE KEY PROJECTS**

#### (1) BERT Fine-tuned Aspect-based Sentiment Analysis (more details in link -->)

- Goal: To Build a reusable Sequence Classification ML Pipeline
- -- which converts customer comments into trackable Aspect and Sentiment pairs
- The development environment, replicable via docker for model training and inference,
- -- (1) is used for building 30+ different text classification models
- -- (2) is used by analysts with limited knowledge in DL
- The pipeline helped in
- -- (1) less annotation for Training (compared to a traditional ML algorithm) by intelligent use of DL+ML models
- -- (2) achieving easily an F1 score of 85%+ for all models with tough 25+ classes and with just 2K-4K annotated data
- Libraries: Python, PySpark, SpaCy, PyTorch; Tools: GPUs, Shell scripting, Docker, PyCharm and GitHub

#### (2) Personally Identifiable Information (PII) Detection using NER (more details in link -->)

- Goal: To anonymize PII in text data
- -- (1) by building a Named Entity Recognition (NER) system which employed both RoBerta Transformer model and Rules-based logic
- -- (2) by replacing the PII words with appropriate generic text
- -- (3) that can result in less restricted use of the data
- Bootstrapped the training data using Spacy rules (thus easing the annotation process by not starting labeling from scratch)
- Achieved an F1 score of 89% for detecting the PII entities
- Deployed an asynchronous\* Inference REST API (using FastAPI and K8s) that can be plugged into multiple applications
- Libraries: Python, SpaCy, Transformers, PyTorch, Celery/Redis\*, FastAPI; Tools: GPUs, Docker, Kubernetes, PyCharm, and GitHub

#### (3) Unsupervised NLP Semantic Search Pipeline

- Goal: To connect two automotive domain specific text data sources,
- -- which technician comments about issues before the launch of a vehicle,
- -- by assigning semantically matching common part categories to every issue in both data sources
- Built a pipeline that ensembles results of 3 pairs of Retriever-Reader models wherein
- -- the Retriever narrows down the search space and
- -- the Reader zeroes in on the right results
- Built a simple Streamlit UI interface for a business user to try the approach and
- Created a CLI app for a domain analyst to experiment with different hyperparameters in parallel across different GPUs
- Libraries: Python, Sentence Transformers, PyTorch, SpaCy, Streamlit; Tools: GPUs, Docker, PyCharm, and GitHub

#### (4) Text Data Clustering Pipeline (more details in this link -->)

- Goal: To build reusable text data clustering pipeline
- = -- (1) with simpler Python APIs for non-NLP analysts.
- -- (2) for deriving actionable insights from unlabeled text corpus
- The clustering pipeline provided options for both Traditional Topic Modeling and DL-Embedding powered Hard Clustering
- Incorporated the models into an easy-to-use Streamlit UI deployed via K8s
- Libraries: Python, Sentence Transformers, Transformers, Sklearn, Seeded LDA, pyLDAvis, Streamlit; Tools: GPUs, Docker, Kubernetes PyCharm, and GitHub

#### (5) Miscellaneous Adhoc Efforts

- Social Media Analysis: Analyzed the latent preferences expressed by consumers owning different vehicle models in Reddit threads for aiding in targeted marketing
- = Built a multi-GPU inferencing pipeline for enabling parallel prediction of a Neural Machine Translation model (developed by a different team)
- Benchmarking of Speech2Text Models :
- -- Explored Speech2Text models such as DeepSpeech (open source), Microsoft Speech2Text API (paid) and Google Speech2Text API (paid)
- -- Compared their Word Error Rate performance for open source LibriSpeech and company-internal speech datasets
- Migrated docker environment and codebase of projects from 2019 to be compatible with Cuda 11 as company upgraded its on-premise GPU infrastructure
- Regularly aid fellow team members in the Kubernetes deployment of their applications