(Work Plan & Project Initiation (P.I.) Sheet)

PROJECT TITLE:	Implementation of ChemBERTa NLP Algorithm for Chemical Fingerprinting of Molecules and Property Prediction at BPC	PROJECT NUMBER:	
PROJECT CONSULTANT:	Emmanuel Oyekanlu (Deloitte(SFL)) & other Consultants from Deloitte SFL who possibly had worked with BPC in previous projects.	DATE INITIATED:	10/14/2024
KEY CUSTOMERS:	Big Pharma Company (BPC)		
PROJECT SPONSOR:	Big Pharma Company, SFL (Deloitte)		

OBJECTIVE: What is the purpose of the project?

Chemical fingerprinting is used to identify and characterize a substance by analyzing its unique chemical composition. This process essentially creates a "fingerprint" of the substance, thus allowing for comparison to known samples. Chemical fingerprinting enables tasks like materials quality control, source identification, and contaminant detection in various fields like environmental monitoring, food and drug analysis, forensics, molecules analysis and drug discovery to be easily accomplished.

ChemBERTa, an adaptation of the popular RoBERTa, machine learning and natural language processing (NLP) algorithm has been found to be quite useful for chemical fingerprinting and molecules properties prediction when trained on the Simplified Molecular Input Line Entry System (SMILES) data set. SMILES is the world largest open sources collections of chemical and molecules structures.

Big Pharma Company is seeking to leverage its extensive materials data sets and the ChemBERTa algorithm for chemical fingerprinting, molecules identification and representation, and molecules properties prediction.

Deloitte SFL engineers will be able to assist BPC by leveraging Deloitte's expertise in data engineering, machine learning, software development and natural language programming processing (NLP) to achieve an implementation of the ChemBERTa algorithm using BPC data.

Refine objectives by further discussion with project key customers

DELIVERABLES: What are the major deliverables or focus of the project?

- a. A ChemBERTa based NLP model that can do molecules property prediction using BPCs data.
- b. Integration of the developed model for downstream applications at BPC.
- c. Integration will be accomplished by deploying the model for BPC data property prediction through cloud and API (Application Programming Interface) end point deployments.
- d. Reliability of the deployed model will be accomplished by deploying through Docker containerization and Kubernetes deployments.

Refine deliverables by further discussions with key project customers

SCOPE:

Covered:

BPC data curation, data cleaning, and extraction, transform and loading processes.

ChemBERTa algorithm-based prediction model that can be used to achieve chemical fingerprinting and molecules property prediction of materials.

(Work Plan & Project Initiation (P.I.) Sheet)

Integration of the model for day to day use at BPC

Not Covered:

To be discussed with key customers

JUSTIFICATION / BENEFITS: Why initiate project? What are business reasons / paybacks for doing project? (Impact, Cost of Poor Quality [COPQ], etc.,

- (1) Enhanced molecules prediction, material identification and chemical fingerprinting
- (2) Better and faster drug discovery
- (3) Better and faster understanding of the biological, chemical and physical nature of materials through their chemical fingerprints
- (4) Possibilities of developing targeted drugs for specific BPC customers, leading to better customer engagements, customer retention and opportunities for new businesses and applications.
- (5) If BPC does not develop the model on time and at cost, there is possibility of their losing vital chunks of their customers to local and international rivals

RISKS: What potential problems are associated with the project?

- Limitations of existing dataset Small dataset of some material can lead to low prediction accuracy Possibly imbalanced data: special treatment required.
- It is expected that BPC will have a big structured and unstructured data set. Curating and data selection can take a lot of time.
- Pharmaceutical data set due to the possible connection with human data set through usage always need special care due to government regulations and security.

ASSUMPTIONS: What is required for the project's success? (i.e.: resources, budget, development time, etc.) Deloitte SFL resource time: Estimated 245-400 hours total. This can translate to ~ \$72,000 if Deloitte SFL engineers charges for their time at the rate of \$180/hr. Other miscellaneous charges from Deloite SFL can be added as appropriate.

RESOURCES / ROLES: Include required resources / team members / responsibilities on the project

Main Resources:

- Project Manager (Data Engineering / Al Applications)
- 2 Data Engineer with knowledge of GPU, RAPIDS (cuDF, Dask), Python Vaex, and/or OpenCL libraries

Responsibilities:

Curating and transforming (tokenization, vectorization) BGC's data. These engineers must have adequate experience with traditional ETL applications and cloud-based ELT applications.

BGC already have cloud capabilities, hence Deloittes SFL engineers shall work along with BGC team to ensure that relevant data (original and transformed data) are available for the downstream NLP engineers.

The Data Engineers should be able to work with both relational and NoSQL databases since it is expected that clients such as BPC will have very big structured and unstructured data sets. Other responsibilities, including cleaning the data, and possibility separating the data from 3D molecular data to 1D and 2D data sets as needed.

(Work Plan & Project Initiation (P.I.) Sheet)

2 NLP Engineers with knowledge of HuggingFace suite of libraries.

Responsibilities:

Model training through transfer learning. Use BPCs data to fine-tune ChemBERTa. Based on availability of time and resources, it may be desirable to implement and train RoBERTa algorithm from ground up since the authors of ChemBERTa mentioned that prediction performance of ChemBERTa could be improved by training over more epochs.

Data Visualization Engineer with knowledge of 2D and 3D material visualizations.

Responsibilities:

Molecules exists with 3D geometries (viz: biological, chemical and physical properties). Hence, at least 1 data visualization engineer with extensive knowledge of open-source 3D graphics tools such as Plotly (with dash bio), ChemPlot, RDKit, Mayavi, PyVista etc. The visualization engineer is also expected to be comfortable with rendering results on dashboards.

• 2 molecules/chemical/material engineers and/or domain experts from BPC.

Responsibilities:

These domain experts shall serve as initial sounding board to advise the data and machine learning engineers regarding the suitability of their model to BPC's test data sets.

• 2 BPC Engineers:

Responsibilities:

For future model retraining and prevention of model drift, it is useful to have at least 2 engineers (data and NLP engineer) from BPC to work along with engineers from Deloitte/SFL. It will be great if these BPC engineers have working knowledge of Lambda architecture so that they can be able to effect retraining of the developed model using newly available batch and if required, to use streaming data in real time as well. These engineers can always reach out to Deloitte SFL consultants if they have issues that are not within their knowledge domain.

Software Engineer:

Responsibilities:

To transition the model into a deployable software. BPC may need the software with different types of access points for many of their molecule/chemical fingerprinting engineers. They may also need to a desktop version of the software. The software engineer will work along with the DevOps Engineer to package the software in the needed formats. For example, desktop deployment, deployment through Kubernetes cluster for reliability and software availability.

DevOp Engineer:

Responsibilities:

Work along with the Software Engineer to accomplish dockerization, Kubernetes deployment and provisioning for desktop version of the software.

Consultants:

Emmanuel Oyekanlu (Data/Al/Softwarization), other consultants from Deloitte SFL.

(Work Plan & Project Initiation (P.I.) Sheet)

Hardware Needs:

For this project, it could be reasonably assumed that BPC shall provide the Deloitte/SFL team access to their cloud platform. Hence the cost of using BPC's cloud platform for all phases of the project shall be paid by BPC.

Hardware instances for this type of project always have needs for parallel processing of data. If BPC cloud instances does not yet have support for GPU based parallel processing, request shall be made for cloud instances that support parallel processing capabilities. It was reported that an NVIDIA V100 GPU hardware was able to train 10million SMILES data within 48 hours.

Software Needs:

Many of the software tool that will be needed for this project may be readily available as part of the BPC cloud platform. Since BPC have a competent cloud team, it is reasonable to assume that BPC's cloud platform may have necessary parallel processing cloud software instances such as PySpark, and RAPIDS suite of libraries.

BPC can also be urged to make subscription for Databricks if not already available within their cloud domain. Databricks have natural support for MLFlow. This can be useful for timely transfer learning, model retraining and evaluation.

MAJOR MILESTONES / DECISION POINTS:	Est. Hours	Est. Complete	Est. Act. Complete
Initial phases of discussion by BPC and Deloitte	This could vary	This could vary	This could vary
SFLs Chief Officers, CTOs, Project Consultants,	depending on	depending on	depending on
etc.	relationship	relationship	relationship
	etween BPC and	between BPC and	between BPC
	Deloitte	Deloitte	and Deloitte
In-house discussion between Deloitte SFLs	Same as above	Same as above	Same as above
Chiefs, CTOs and the Project Consultants.			
Initial ETL/ELT activities. Data engineering, data	60	10/14/2024	10/25/2024
cleaning, data augmentation. Unity Catalogue			
arrangement if BPC cloud platform have support			
for unity catalogue.	00	40/05/0004	44/44/0004
ChemBERTa model fine-tuning through transfer	80	10/25/2024	11/11/2024
learning.			
Training with more BPC data set.			
Initial model evaluation.			
Discussion with BPC domain experts.			
Further retraining and model fine-tuning based on			
domain experts' suggestions.			
Software and DevOp Engineers dockerization,	55	11/11/2024	11/19/2024
cloud integration, Kubernetes deployment and			
possibly Desktop app design			

(Work Plan & Project Initiation (P.I.) Sheet)

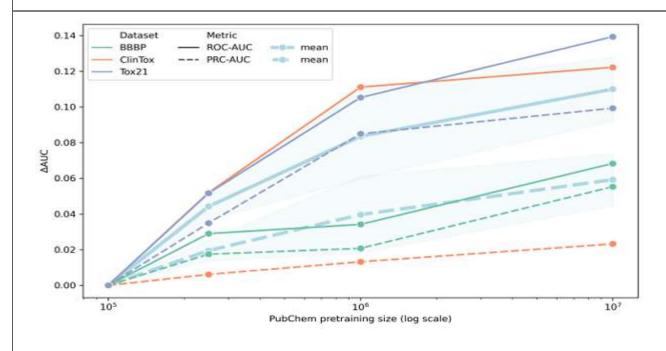
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Usage manual design by the software and NLP engineers	50	11/19/2024	11/26/2024
Training of BPC molecular engineers with regards to model/app usage.			
Feedback from customers. ML and app refinements.	TBD		

Project Time Line (Gantt Chart)				
	Approximately 7 weeks			
	Week 1 (Approximately 1 week based	A 2222	oximately 6 w	ranks
	on relationship between BPC and Deloitte)	Арри	oximately 6 w	CCRS
Project Initiation	Discussion s/Project specification (BPC/Deloitte SFL Team)			
Data ETL/ELT activities		Data Engineers activities		
ChemBERTa Model transfer learning with BPC data			NLP/ML engineers' activities	
Software & DevOps Engineers				Model Deployment

(Work Plan & Project Initiation (P.I.) Sheet)

SUCCESS CRITERIA:

Model Evaluation: ROC and AUC curves as suggested by initial authors of ChemBERTa will be useful
to measure developed models success. An example from the original ChemBERTa paper is shown
below:



Model: Testing further with known molecules and their signature to see if model generalizes well. Apart from ROC/AUC curves and model generalizations, success criteria can also be based on the percentage of accurate molecular property predictions.

Success can also be measured by field reports from model users as BPC.

PROJECT START UP APPROVAL SIGNATURES			
Agreement on PI she	et content		
Project Consultant:	Emmanuel Oyekanlu, Deloitte(SFL)	Date:	
Key Customers:	BPC	Date:	
They Guestermere:		Buto.	
Duele et Cureure eu	DDC	D-4	
Project Sponsor:	BPC	Date:	
		Date:	