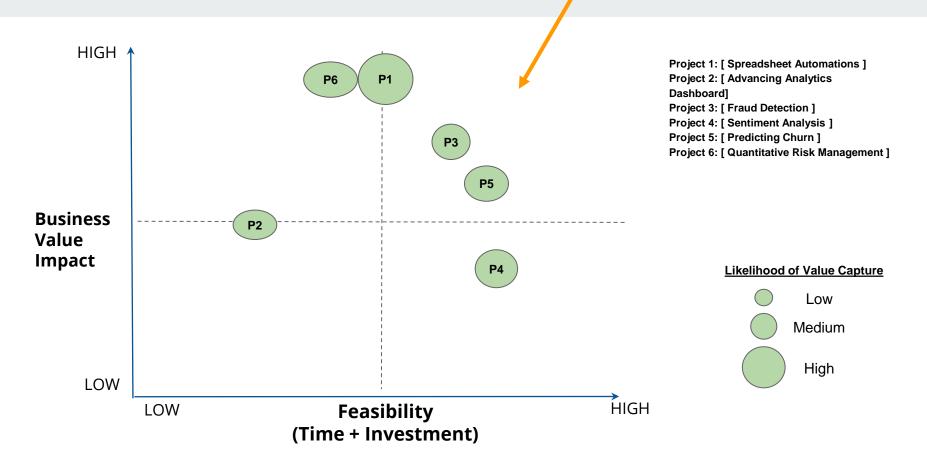
# 100 Day Data Science Plan: Building A Data Strategy

<u>Step 2, Part 2:</u> Complete the "Data Science Opportunity Matrix" below by modeling each of the six projects in terms of feasibility (time & investment), business value impact, and likelihood of value capture



### **Executive Summary**

#### Purpose of 100-day plan

- Establishing a data infrastructure that will pave the way for an enterprise 'Analytics
- Processing Automation' or APA. The over all analytics would utilize a cloud based
- autonomous database

#### **Approach**

- Utilizing a an entire team with multi skill sets. There will be a heavy emphasis on cloud
- infrastructure where all the built in AI and ML engineering would take place. An RPA infrastructure
- platform would be utilized in order to handle the automation tasks

#### Results

- A n established Analytics Processing Automation (APA) platform that would be utilized by the
- entire organization

### **Scope of Work for First 100 Days**

#### 1.PURPOSE of 100-DAY PLAN

- Establish a proper data infrasrtcuture that can be utilized for various projects
- Assiging the proper team members for the specific roles and projects
- Ensure that all stakeholders have a cleaer and solid picture of the 6 major projects

#### 2.APPROACH

- I have used a variety of approaches per project. This has always been a case by case bais
- All 3 Descriptive, Predictive & Prescriptive approaches have been utilized
- I view each of the projects not as separate entities but also have interrelations with each other. One project's success would eventually benefit the others.

# **Candidate Data Science Projects**

|   | Functional Area                                  | Project Description  |
|---|--|--|
| Project 1:<br>Spreadsheet Automations               | Financial Management                             | Majority of Excel & Google Sheets that involves basic to intermediate financial processing would eventually be automated                             |
| Project 2 Name: [Advancing the Analytics Dashboard] | Business Functional Area:<br>[Digital Marketing] | Using advanced ML in order to have deeper key insights embedded into the dashboard for digital marketing campaigns & over-all marketing performance. |

## **Candidate Data Science Projects**

| Project 3 Name: [FRAUD<br>DETECTION] | Business Functional Area:<br>[FINANCE – ANTI MONEY<br>LAUNDERING (AML)<br>DIVISION] | Heavy ML usage due to anomalies that occur within the transactions.   |
|--------------------------------------|---|---|
| Project 4 Name: [SENTIMENT ANALYSIS] | [LEGAL DEPARTMENT]  | Heavy analysis on a vast amoung of documents. Using AI & ML in order to see certain patterns and predict certain behaviours of clients. |

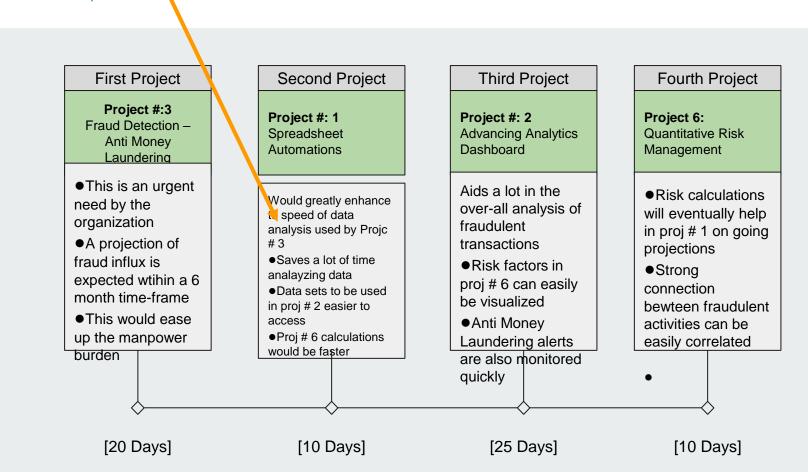
## **Candidate Data Science Projects**

| Project 5 Name: [PREDICTING<br>CHURN]                | Business Functional Area:<br>[DIGITAL MARKETING] | Reliance on Google Analytics & Google Ads/ Bing Ads / FB Ads are simply not enough. We will use certain ML models in order to maximize the gains. |
|--|--|---|
| Project 6 Name:<br>[QUANTITATIVE RISK<br>MANAGEMENT] | [FINANCE]  | Building models in order to understand the risks of financial portfolios.   |

**Step 2, Part 3:** Complete the "Data Science Road Map" below with the first four data science projects chosen for implementation.

| <u>Order</u> | <u>Project</u>          | Order Justification  |
|--------------|-------------------------|--|
| 1            | Fraud Detection         | There has been an urgent need from the organization due to the rising occurrences of financial fraud.  |
| 2            | Spreadsheet Automations | Since the main focus of the project is on an entire Automated Analytics Platform, automating most of the excel & google sheets is vital. There will be a lot of VBA, C# coding and a heavy reliance on an RPA platform (UIPath). |
| 3            | Analytics Dashboard     | The marketing team is highly in need of a more robust BI tool. Although the current platform functions well, the CMO wants to have a real time insight into each KPIs.   |
| 4            | Quant Risk Management   | Building ML models to analyze several portfolios is a complex task. Although this would take up manpower and would fully utilize the entire specializations of the team, the urgency is not that crucial.                        |

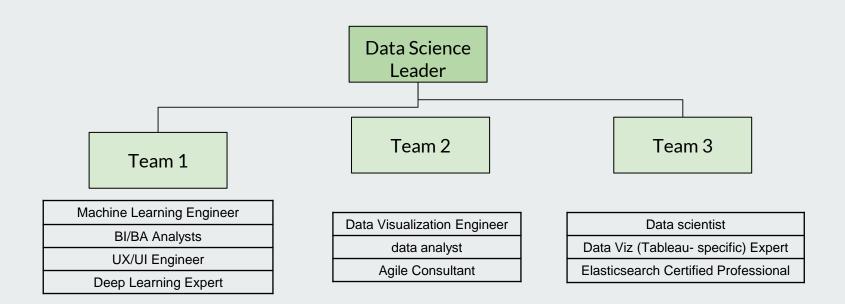
<u>Step 2, Part 3:</u> Complete the "Data Science Road Map" below with the first four data science projects chosen for implementation.



# **Our Highest-Priority Data Science Projects**

| Order  |   | Direct Alignment with Strategic Goals? | Cost          | Complexity of Implementation | Certainty of<br>Value Capture | Magnitude of Benefit |
|--------|---|--|---------------|------------------------------|-------------------------------|----------------------|
|        |   | 1=Low; 5=High                          | 1=High; 5=Low | 1=High; 5=Low                | 1=Low; 5=High                 | 1=Small;<br>5=Large  |
| First  | Project 3:<br>[Fraud Detection<br>Anti Money<br>Laundering] | 4                                      | 3             | 3                            | 5                             | 5                    |
| Second | Project 1:<br>[Spreadsheet<br>Automations]                  | 5                                      | 5             | 3                            | 5                             | 5                    |

### **Initial Structure of the Data Science Team**



# I have identified six strategies for promoting a data-driven culture in our business

### Strategies for promoting a data-driven culture

|   | Strategy  |
|---|---|
| 1 | Identifying more than specific data science methodology for each project, then deciphering if<br>1 or 2 methods are enough to arrive at a solution  |
| 2 | Viewing each project not just as a stand alone task but also as highly inter-connected to other projects. After all, a poor performing project would eventually affect the performance of others. |
| 3 | Focus and drive all projects to specific proof of concepts. Look at them as a whole unit with the common goal of making our organization much better.   |
| 4 | Maximizing the skills and talents of our employees and partner consultants, before looking externally for a potential consultant  |
| 5 | Maximizing the current infrastructure we have; both on premise & cloud based.   |
|   | Agile methodology is the main framework used all throughout the 6 projects. Since these projects would be unpredictable and involve certain iterations, Agile trumps waterfall in all 6           |
| 6 | of them.  |

# **Technical Infrastructure Needed to Support the Data Science Organization**

| Data<br>Requirements | What data should be included in the Data Strategy? | This is a huge scope. This would include a lot of financial data and a lot of marketing (both digital & traditional).  There will also be a certain amount of qualitative data that would be gathered from UX research.  |  |
|----------------------|--|--|--|
| Data<br>Governance   | Data Availability                                  | <ul> <li>All data that would be utilized is readily &amp; openly available to the team members involved ONLY. We consider the data being used as HIGHLY CONFIDENTIAL, specifically for Project 3 (Fraud Detection) and Project 6 (Risk Management). All other data that deals with legal matters and marketing data should be kept with utmost confidentiality.</li> </ul>   |  |
|                      | Usability  | <ul> <li>Data usability would be on a case by case basis. As every project moves forward, certain groups and departments would be able to access the data that is being analyzed and presented. A simple scneario would be: a. All the data from the 'fraud detection' project could only be accessed by the team that works in the anti fraud / AML unit.</li> <li>B. The dashboard analytics that contains the marketing KPIs and near real-time monitoring would have certain 'viewing privileges' depending on one's current role and tasks.</li> <li>. the Risk Management data could not be viewed by anyone outside of their department.</li> </ul> |  |
|                      | Integrity  | <ul> <li>Privacy and security workshops would be provided to all employees undergoing all these projects.</li> <li>There will be a constant weekly audit to make sure that privacy &amp; compliance are being observed.</li> </ul>   |  |
|                      | Security   | <ul> <li>Data access of given on a 'per role/ privilege' basis.</li> <li>Data is behind a firewall.</li> <li>All users would ALWAYS be required to enter the proper user name and password.</li> </ul>   |  |
| Technology           | Data Architecture<br>Components                    | also had a dedicated Neo4.I graph DB that was started 2 years ago. This is mainly for the Fraud Detection project  |  |

# Technical Infrastructure Needed to Support the Data Science Organization

| Support for<br>Machine<br>Learning | Machine learning architecture                          | We opted for building our data architecture in-house. We have a very robust and highly experienced team not just in data science and analytics, but also our automation specialists have the necessary skill sets.  |
|------------------------------------|--|---|
| Skills and<br>Capacity             | Data literacy skills<br>and organizational<br>capacity | <ul> <li>All of the employees involved these projects are already skilled in their respective roles. But this does not limit them from having additional on-line training in the areas of: machine learning, automation, A.I. fundamentals, IT security, risk and compliance.</li> <li>All employees would have access to on line platforms &amp; learn free. Such platforms are: Udacity, Coursera, DataCamp, TreeHouse, Udemy, EDX and Linux Academy</li> <li>There will be constant BVIRs (Big Visual Information Radiators), whether they opt to work (1x a week due to COVID) at the office or a virtual version when working remotely (new norm).</li> <li>These BVIRs would have constant reminders embedded within regarding data literacy skills.</li> </ul> |