

# Investigating Quality Education from school surveys

Team Spark Champs

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# Introduction

Policy Improvement

Institutions

Decides Policy  
Allocate Resources

Feedbacks

Surveys indicate  
how effective are  
the policies

Educational Units

Deploy/Use  
resources

# Why Care? SDG 4- Quality Education

- **SDG 4** - ensure quality education accessible to all by 2030
- 262 million between age 6-17 were still out of school in 2017.
- Percentage of trained primary school teachers stagnant at 85 since 2015.
- Ground reality of government measures can be understood by student and teacher **feedbacks**.
- Study Hours **highest** for UAE but learning outcomes are **poor**
- Study Hours **lowest** for Finland but student performance **high**



Image Credits :

<https://leverageedu.com/blog/best-education-system-in-the-world/>

# Why Big Data?

- Process large data (~20 GB)
- Use techniques to analyse and obtain inferences
  - Similarity Analysis
  - Multi-Hyp Testing
  - Large Scale Machine Learning
- Need to use data pipeline to distribute the tasks on nodes and aggregate results
  - HDFS
  - PySpark
  - Tensorflow

# Keywords

- SDG Goal 4.1.1 ⇒ Achieving target proficiency level in Literacy and Numeracy
- Feedback features ⇒ Students' responses on survey questions [done by [PISA](#)]
- Subject Scores ⇒ Students' scores in different subjects [done by [PISA](#)]

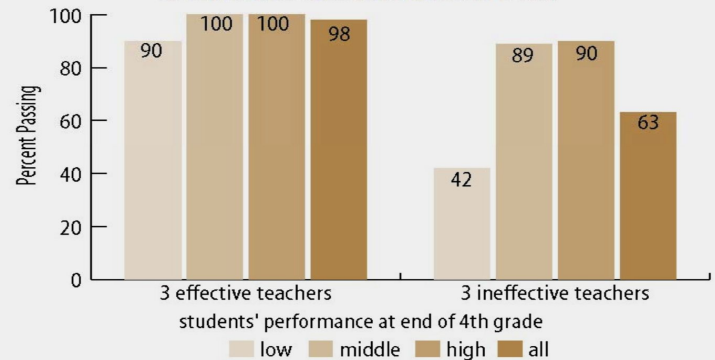
# Background

- Studies [[Ref 2.](#)] shows that school and family have only very little impact on academic achievement among pupils from disadvantaged backgrounds.

*But Teachers are the most influential factor in student learning.*

## The effect of teachers accumulates

4th graders of all abilities will pass 7th grade math test with 3 effective teachers in a row



Source: Sitha Babu and Robert Mendro, *Teacher Accountability: HLM-Based Teacher Effectiveness Indices in the Investigation of Teacher Effects on Student Achievement in a State Assessment Program*, AERA annual meeting, 2003.

Image Credits: [References 2](#)

# Past Work

- Multinomial regression analysis was conducted[[Ref1](#)] to identify characteristics of students
  - Students for scholarship
  - Extracurricular activities
  - Parents' education
  - University they study inwhich make perception about quality of higher education dissimilar.
- Studies [[Ref2](#)] shows that “Discovery”-based approaches have produced very positive outcomes in classes taught by exceptional and highly committed educators.
  - However,considerable time and energy is required on the part of the individuals, average teacher is undoubtedly not in a position to contribute.

# Data

- PISA [~20 GB] (<https://www.oecd.org/pisa/>)
  - Reading, mathematics and scientific literacy scores of more than 710,000 15 - year old students representing 31 million student from **79 countries**. Along with general survey from teachers, parents school principal and students every 3 year since 2000
  - Number of features > 1120, observations > 1 million
  - **Representative Data, Feedback Data (ST\*), Subject Scores (PV\*)**
- UNESCO SDG [1 GB] ([UIS Statistics](#))
  - Country wise quality indicators per year (1970-2019) (1GB)
  - Number of Features: >3000



# Methods

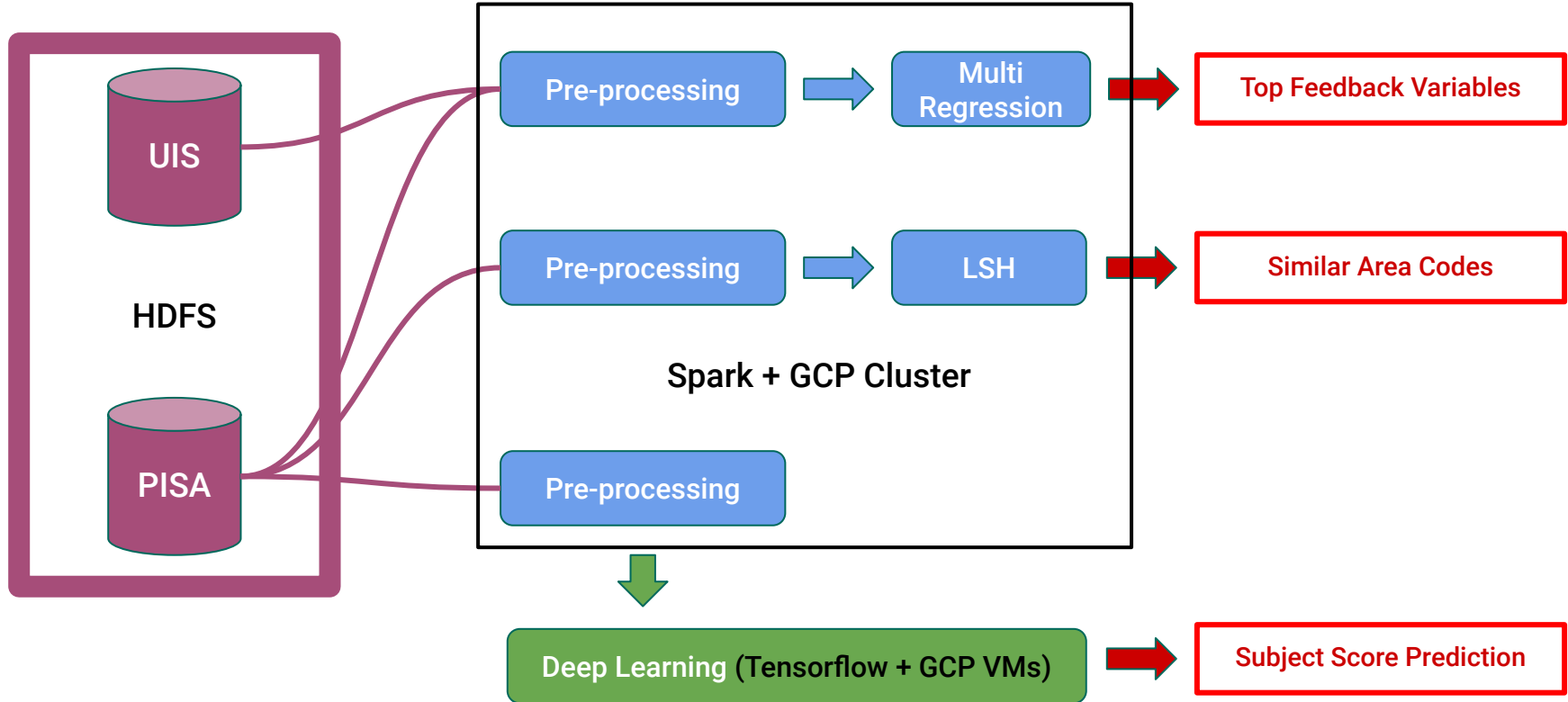


Fig: Code Analysis Pipeline

# Methods - Hypothesis Testing

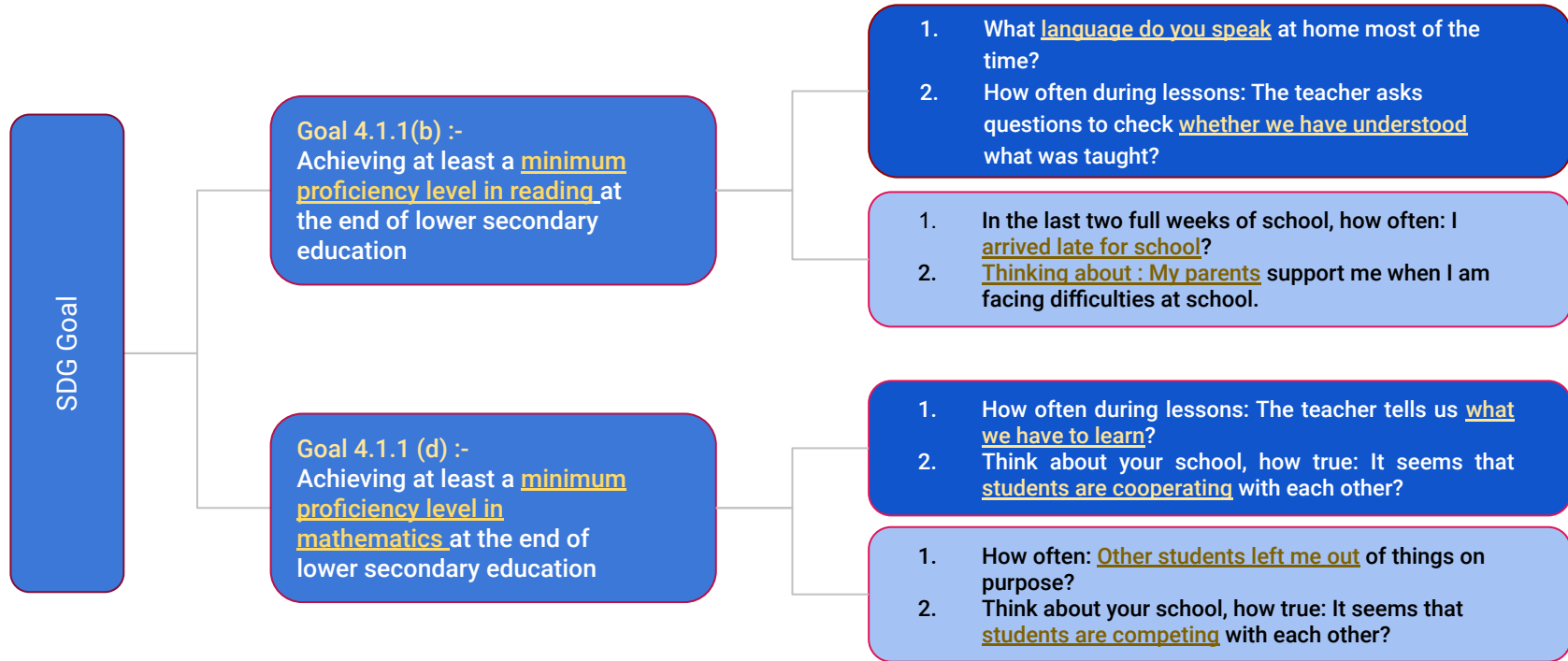
*Task* :- Find Correlated Feedbacks for the success of a SDG Goal

*Approach* :-

- **Spark RDD**, data  $\Rightarrow$  (Y = goal score, X = features) for each country
- 2 Target SDG goals
- **Multi-linear regression**  $\Rightarrow$  beta value and p value for each feature
- Top **positively** and **negatively** correlated features for each SDG goal

SDG 4.1.1 (b)	Feature	Beta Value	P value
1	ST153Q10HA	0.41500	0.0
2	ST207Q04HA	0.29350	0.0
-2	ST166Q03HA	-0.23453	1.24e-55
-1	ST205Q02HA	-0.36735	6.4e-114

# Results - Hypothesis Testing



Result: - **Most** or **least** correlated features for a given goal

# Methods - Similarity Search

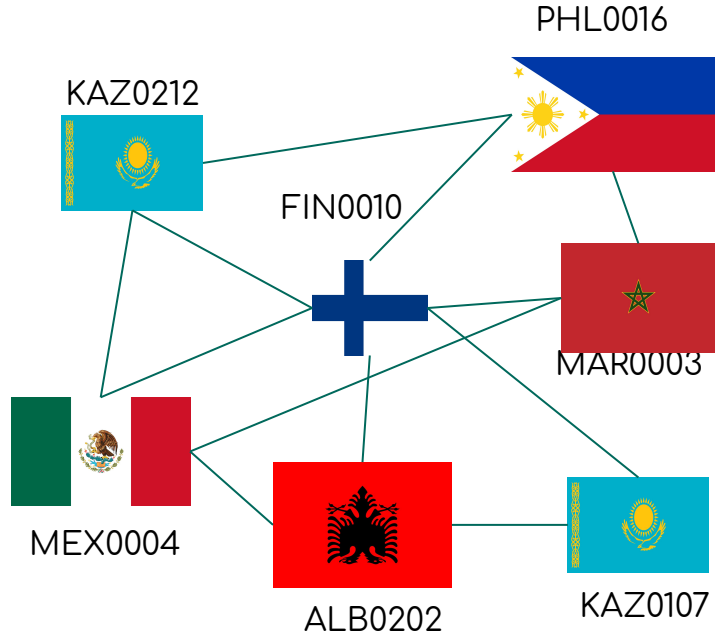
## Clustering and Similarity Search using LSH

*Task* :- Find similar schools and areas from survey feedback collected from students

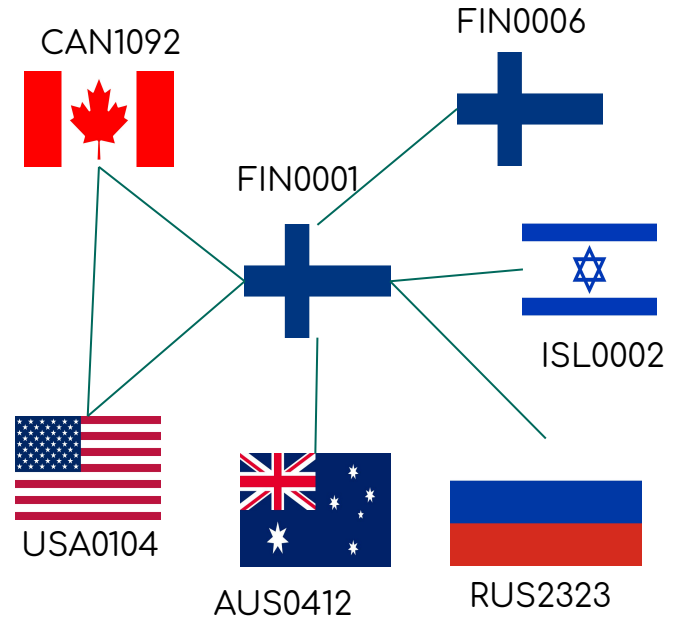
*Approach* :-

- **Characteristic matrix**  $\Leftarrow$  buckets [0-20%, 21-40%, 41-60%, 61-80%, 81-100%] for feedback responses
- **Signature matrix** (~1500 columns, ~60 rows)  $\Leftarrow$  **Characteristic matrix** (~1500 columns ~9000 rows).
- Analysis on band values- (b=5,r=12), (b=8,r=8), (b=12,r=5) [Sim > 0.8]
- **Locality Sensitive Hashing**  $\Rightarrow$  Similar area codes / schools (JSON)

# Insights - Similarity Search



Schools from some developing countries matched with some Finland area schools



Schools from some developed countries matched with some Finland area schools

# Methods - Deep Learning

*Task :-* Predict Performances in Subject based on students feed\_back data

*Approach :-*

- Distributed Preprocessing (using Spark Dataframes and HDFS, GCP Cluster **1 Master 2 Worker**)
- Distributed training [ TF2.0 Synchronous All Reduce ]
  - `tf.distribute.MirroredStrategy` - single node multi-gpu, each gpu have their parameter and update each other synchronously
    - GCP Compute Engine VMs with **2 Nvidia T4**
    - Training time ~ little more than a minute per epoch
  - `tf.distribute.experimental.MultiWorkerMirroredStrategy` - multi node multi-gpu, each node communicate with each other in round-robin using RPCs

# Insights - Deep Learning

Design :-

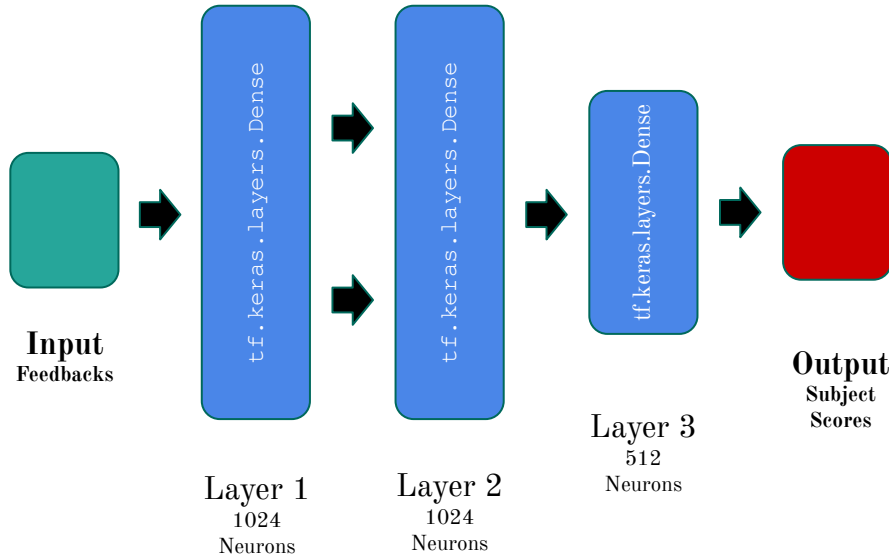


Fig :- Network architecture with 3 hidden layers

Inferences:

- MAE for prediction **0.609** [standardized]
- PV1MATH score for a sample
  - 463 (Actual)
  - 577 (Predicted)
- **Low Precision**

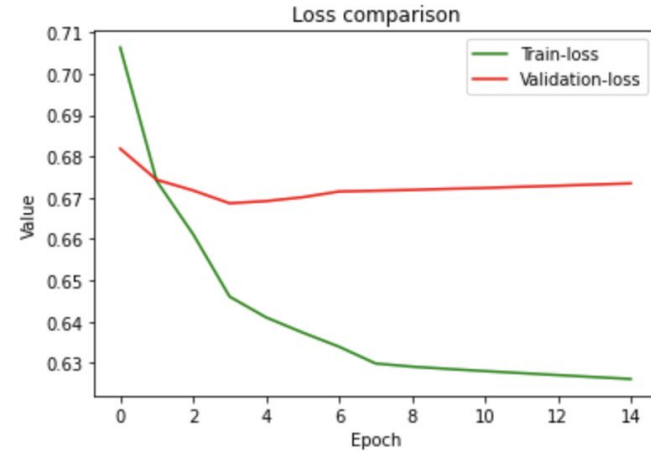


Fig :- Training History

# Conclusion

- With our results  $\Rightarrow$  Feedbacks show a considerable importance in education quality
- Cross-nation similar area codes found can help to decide similar policies, such as
  - Amount of teaching Hours
  - Degree of interaction between students
- Institutions can decide to spend depending on type of students in particular region.
- Inference on feedbacks can help achieve SDG goals
- Future Work - more semantic analysis on feedbacks can produce higher efficient inferences regarding policy improvement.



Image Credits:

<https://myventurepad.com/software-education-something-constantly-revised/>



# References

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2. <http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.498.6274&rep=rep1&type=pdf>
3. <http://www.oecd.org/pisa/aboutpisa/>
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