



SDTM MAPPING AND DATA CONVERSION USING ML/NLP

THE PROBLEM

- Mapping of non-standard raw data to SDTM is manual and time consuming.
- Mapping is repetitive process.
- Endless mapping and multiple mapping decisions for same scenario.
- Require highly skilled resources to execute.
- No automated way to handle errors and multiple iterations.
- Identifying data types and code lists are highly time consuming.

WHY AUTOMATION

- 60%-80% of manual effort can be saved with more study data using the ML/NLP tool.
- Effective use and reuse of Study and Project Metadata.
- Mapping and reuse of Metadata between CDISC Metadata/Sponsor defined standards (including Metadata, IG Version, Control Terminologies, etc.).
- The ML engine is constantly trained for greater accuracy and prediction. Effective reuse of mapping decisions.
- The ML/NLP tool creates the SDTM Mapping Specs and standalone SAS® programs for each domain.
- Greatest advantage for choosing ML/NLP - automate all possible derivation logic, which is key time savings for programmers.

NEED OF MACHINE LEARNING

- Self-learning improves accuracy over time.
- Works well on unknown data.
- System acts as an expert by itself from the earlier learnings to fix issues.
- Percentage of fallouts are very minimum to handle manually.

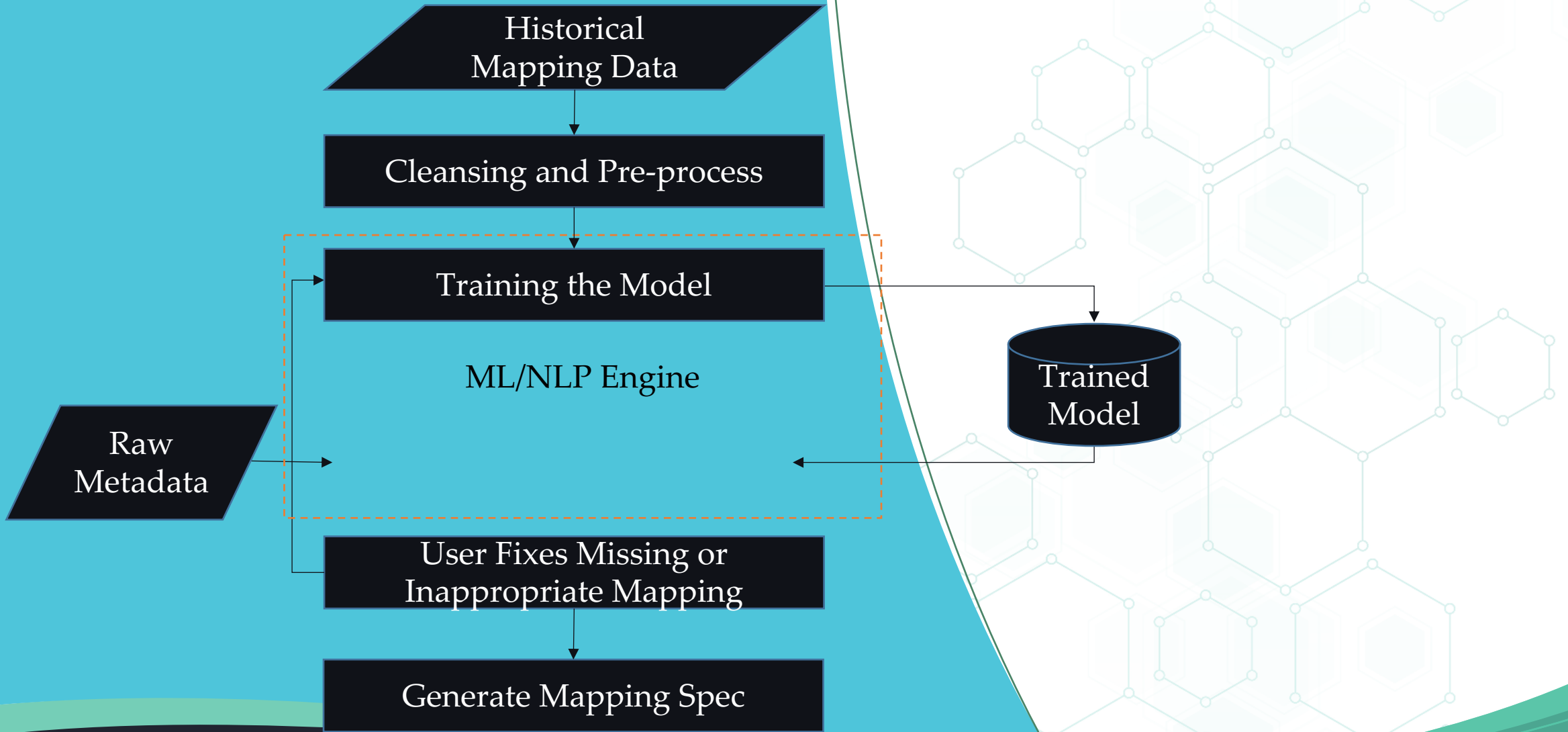
Semi
Automotive
Learning

Less Manual
Intervention

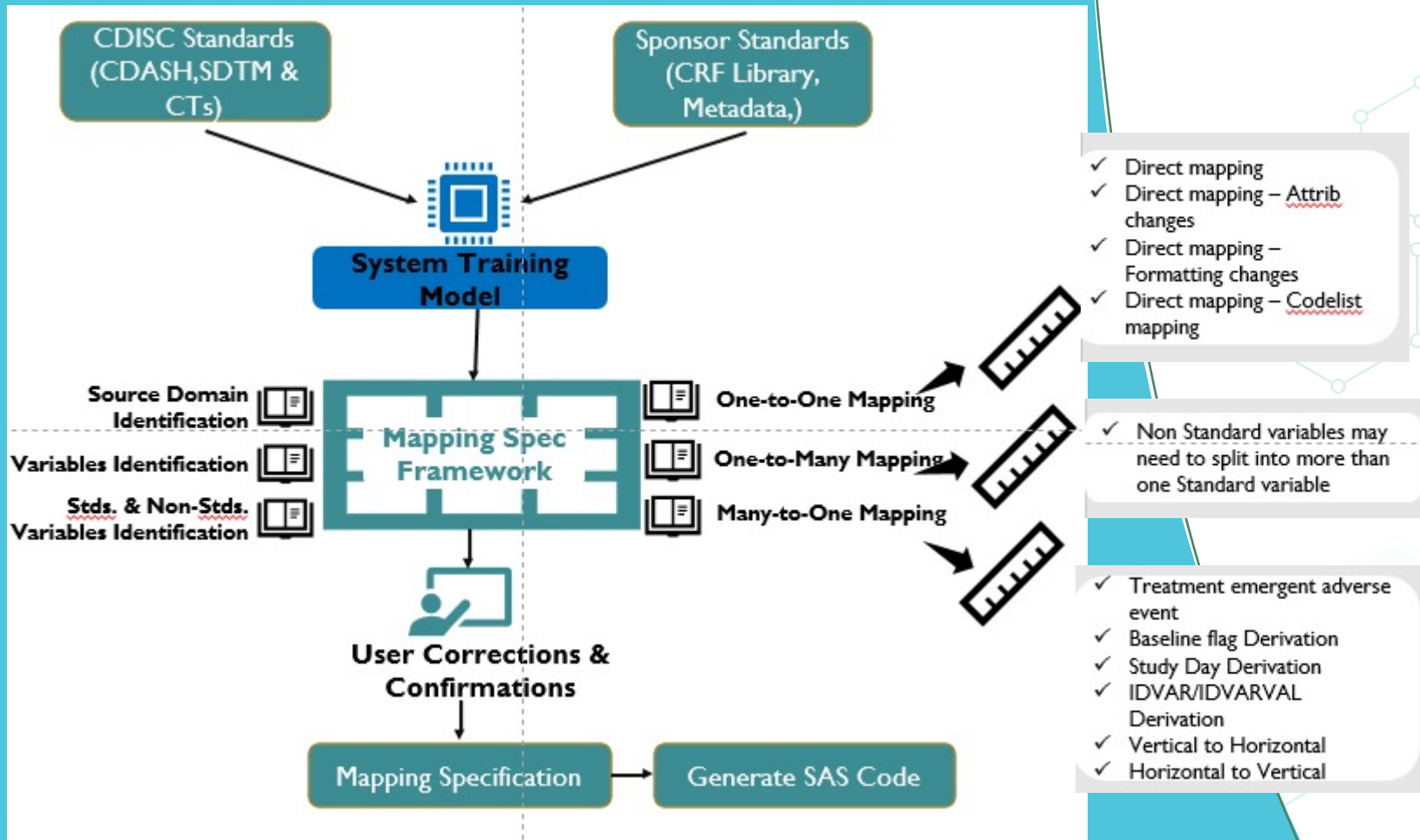
Auto
Processing

SDTM
Mapping
Model
Generation

HOW TO APPLY ML/NLP ?



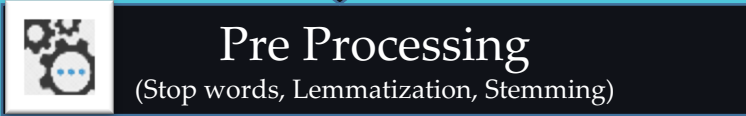
PROCESS



PROCESS TO GENERATE CLASSIFICATION MODEL



Training Data



-Continued

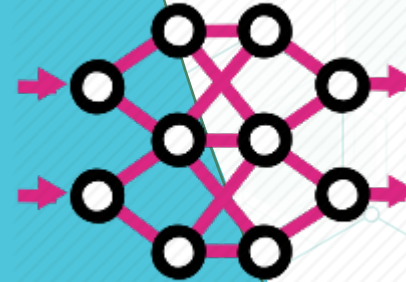
- Term Frequency – Inverse Document Frequency - TF-IDF
 - *Statistical representation of importance's of a word in a document*
 - *Used to weight the term in variable information in a vector*
 - *Vector will feed in to MLP Classifier*

$$w_{x,y} = tf_{x,y} \times \log \left(\frac{N}{df_x} \right)$$

TF-IDF
Term's within document y

$tf_{x,y}$ = frequency of x in y
 df_x = number of documents containing x
N = total number of documents

- Multi-layer Perceptron classifier (MLP Classifier)
 - *Simple and lightweight neural network*
 - *Works better with minimal training data*
 - *Less resource intensive*



STEPS



MINNAL - DOMAIN MAPPING

- System trained based on CDASH and CDISC standards. System can also be retrained based on the Sponsor standards.
- System will automatically predict the primary domains and secondary domains based on “Topic” variables collection.
- System will also provide the functionality to edit the domain prediction. Once changed, System will retrain itself and be available for further studies within the same project.
- Retraining can be controlled by governance team/authorized role.

Upload	Domain Mapping	Intermediate Mapping	Trial Datasets	Final Mapping	Download Specification
Raw Dataset Name	Primary SDTM Domain	Secondary SDTM Domains	Action		
demo	Choose Domain DM	Choose Domains EX, SC	Edit		
comlog	Choose Domain CO	Choose Domains SC	Edit		
aecdisc	Choose Domain SC	Choose Domains AE	Edit		
aed	Choose Domain SC	Choose Domains QS	Edit		
ecgdet	Choose Domain SC	Choose Domains QS	Edit		
fnt301	Choose Domain QS	Choose Domains	Edit		
formats	Choose Domain QS	Choose Domains	Edit		
final	Choose Domain SC	Choose Domains	Edit		
epmhx	Choose Domain MH	Choose Domains QS, SC	Edit		
label	Choose Domain SC	Choose Domains	Edit		

MINNAL - INTERMEDIATE MAPPING

- Based on the keyword, System will also predict the target variable, and will also provide the possible matches.
- User can also manually map the target variable or create the custom domain. System will get retrained based on the user changes. The mapping will be available for subsequent studies.
- System will allow the user to provide a mapping and derivation comments that are used in the specification and eventually the program generator.

The screenshot displays the Minnal.AI web application interface for the 'DEMO STUDY'. The 'Intermediate Mapping' tab is active, showing a table with columns for Source Domain, Source Variable, Source Label, Datatype, SDTM Domain, and SDTM Variable. The table lists mappings for various adverse event (AE) related variables. A search bar is visible on the right side of the table.

Source Domain	Source Variable	Source Label	Datatype	SDTM Domain	SDTM Variable
AE	studyid	Internal Id for the study	Numeric	AE	STUDYID
AE	Subject	Subject name or identifier	Text	AE	USUBJID
AE	AETERM	Adverse Event	Text	AE	AETERM
AE	AETERM_LL1	Adverse Event LL1	Text	AE	AELL1
AE	AETERM_LL1_CODE	Adverse Event LL1_CODE	Text	AE	AELL1CD
AE	AETERM_PT	Adverse Event PT	Text	AE	AEDECD
AE	AETERM_PT_CODE	Adverse Event PT_CODE	Text	AE	AEPTCD
AE	AETERM_HLT	Adverse Event HLT	Text	AE	AELHT

MINNAL - FINAL MAPPING SPEC

- System will provide the Key variables and sorting suggestion based on CDISC. User can modify based on the data collection, that will be used in generated program.
- System will have a standard derivation from global metadata and user provided derivations and automatically identify transpose finding type variables.

The screenshot shows the Minnal.AI interface with the 'Final Mapping' tab selected. The table lists RCTM Variables and their corresponding Code List Code, RCTM Label, and Core status. The right-hand panel shows the 'SDTM Domain' mapping for 'AE' (Adverse Event) with a search bar and a list of variables. The 'Mapping' section shows a source domain 'AE.STUDYID' and a source variable 'Source variable'. The 'Derivation' section shows a formula: `pub(study[id], hesh, )`.

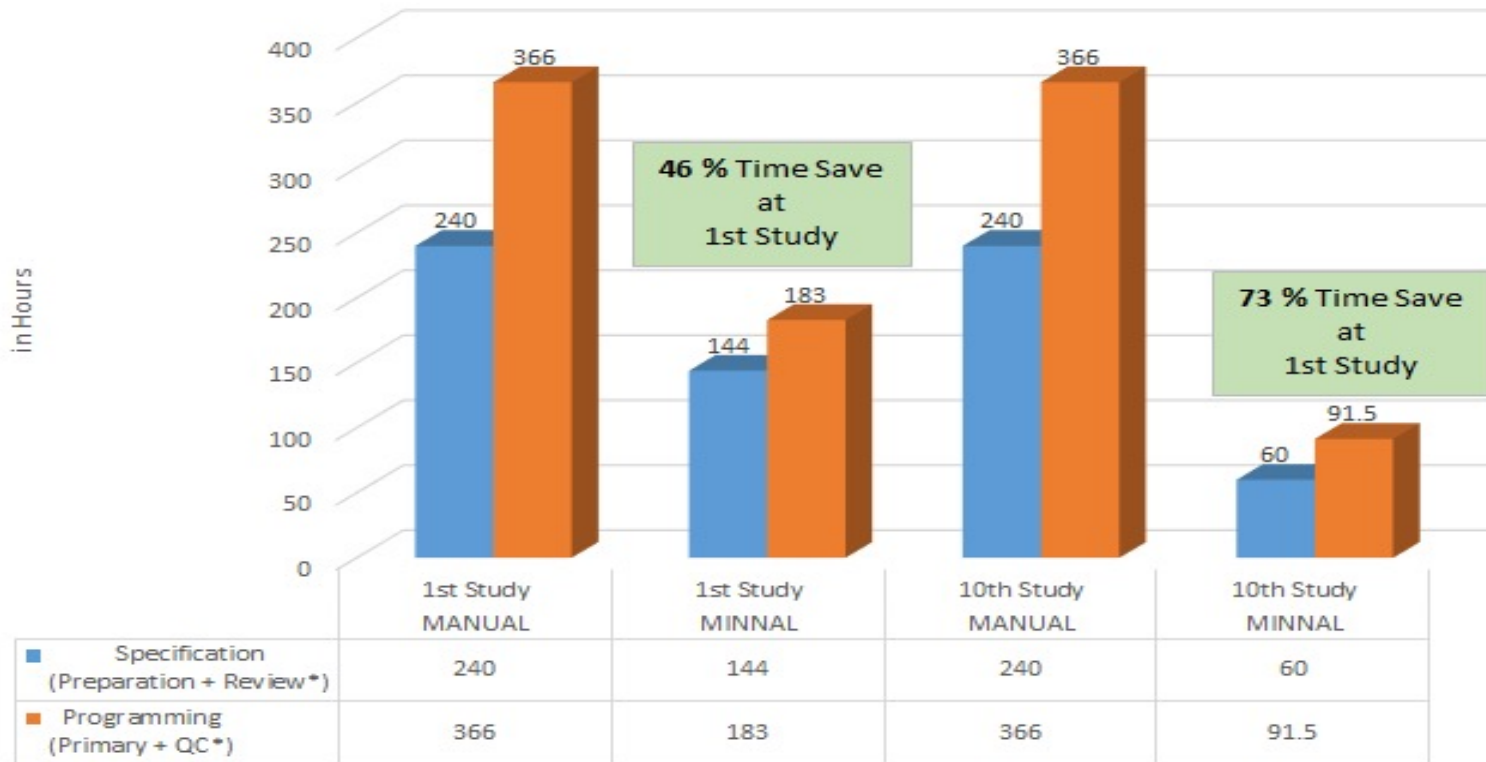
RCTM Variable	Code List Code	RCTM Label	Core
STUDYID		Study Identifier	Required
DOMAIN	DOMAIN	Domain Abbreviation	Required
USUBJID		Unique Subject Identifier	Required
PID		Patient Identifier	Permissible
SPDID		Sponsor Device Identifier	Permissible
ALSEQ		Sequence Number	Required
ARTRPID		Group ID	Permissible
AERPID		Reference ID	Permissible
ALSPID		Sponsor/Protocol Identifier	Permissible
APR NID		Link ID	Permissible

MINNAL - ADDITIONAL FEATURES

- System will generate separate spec for SUPP-- domains.
- System will perform default consistency check, like --DTC and --DY. It can be configurable by the user requirement.
- System will produce a SAS® program per domain, whereas the primary domain program will have the functionality to create its SUPP-- domain.
- System will take SDTM domain's variable order and its attributes to meet the submission guideline.

MANUAL VS MINNAL

SDTM MAPPER
MANUAL vs MINNAL



* By Manual

Input as based on High criticality study which containing 50 sources datasets

Derived on basis of 21 standard safety domain, 4 Custom domains with SE, SV, RELREC and five Trial domains

CONCLUSION

- ML/NLP helps to solve the problem with the historical data.
- It can predict new mapping scenarios.
- It can learn from the user action when System cannot determine a scenario.
- The accuracy will improve over time based on the volume of data.
- System was trained using 8 historical studies and tried on 3 new studies.
- Concluded that this approach could save significant time and resources.

QUESTIONS?

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