Data Preparation for Data Science

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2016

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Hi, I'm Casey Stella!

Garbage In ⇒ Garbage Out

"80% of the work in any data project is in cleaning the data."

— D.J. Patel in Data Jujitsu

Data Cleansing \implies Data Understanding

There are two ways to understand your data

- Syntactic Understanding
- Semantic Understanding

If you hope to get anything out of your data, you have to have a handle on both.

Syntactic Understanding: True Types

A true type is a label applied to data points x_i such that x_i are mutually comparable.

- Schemas type != true data type
- A specific column can have many different types

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- Stripping whitespace
- Normalizing punctuation

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- Automation
- Outlier Alerting

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Semantic understanding does not imply SkyNet

DEMO

```
usage: SummarizerCLI
 -D -D cproperty=value>
                                                      Input properties
                                                      This screen
 -i,--input <SOURCE>
                                                      Input source
                                                      Load an existing
                                                      summary
 -m.--mode <MODE>
                                                      Type of mode. One of
                                                      SOL, CSV
 -nns,--non_numeric_sample_size <NUM>
                                                      Sample size for
                                                      non-numeric data.
 -ns,--numeric_sample_size <NUM>
                                                      Sample size for
                                                      numeric data.
 -o.--output <SOURCE>
 -pct,--percentiles <PCTILE1[,PCTILE2]*>
                                                      A comma separated
                                                      list of percentiles
                                                      in (0, 1007.
 -smo,--similarity_min_occurrance <NUM_OCCURANCES>
                                                      Min Occurrances to be
                                                      considered for
 -ssc,--similarity_score_cutoff <SCORE_CUTOFF>
                                                      Similarity score
                                                      cutoff. Scores are
                                                      range from [0.1].
```

Column Statistical Details HL7Text (0% Missing) ObservationYear (0% Missing) LabObservationGuid (0% Missing) UserGuid (0% Missing) HL7Identifier (0% Missing) ReferenceRange (5% Missing) Units (21% Missing) IsAbnormalValue (0% Missing)

ObservationValue (0% Missing)
LabPanelGuid (0% Missing)
HL7CodingSystem (0% Missing)
ResultStatus (0% Missing)
AbnormalFlags (92% Missing)

Summary for HL7Text

Count Statistics

Canonical Representation Count

Possible Value Synonymns

VALID STRING

Type Modifier Count Distinct Count STRING VALID

protein 1657 ■ hemoglobin 1585 % potassium 1474 *

Canonical Value

Albumin, Serum

Iron Saturation

Calcium, Serum

T4, Free(Direct)

bilirubin

chloride

word

eGFR

synonym

Bilirubin, Total

Prostate Specific Ag...■

Protein, Total, Seru... eGFR AfricanAmerican

eGFR AfricanAmerican

29014 348

Count

1679 ▲

1472 ▼

Summary for Units

Count Statistics

	STRING MISSING STRING VALID		
Canonical Representation Count	Canonical Value	Count	

Type

MISSING STRING 6122

Canonical Representation Count Canonical Value Count VALID STRING a/dl $x{d}{d}e{d}/ul$

4541 ▲ **4180** mg/dl 4032 *

Modifier Count Distinct Count

mmol/l 3443 🛎 % 2640 V

Possible Value Synonymns word synonym g/dL fL pg Ratio a/dL mL/min/1.73m2% M/uL mil/cmm

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- Your data science teams need software engineering chops.

Questions

Thanks for your attention! Questions?

- Code & scripts for this talk available on my github presentation page.¹
- Find me at http://caseystella.com
- Twitter handle: @casey_stella
- Email address: cstella@hortonworks.com

¹http://github.com/cestella/presentations/