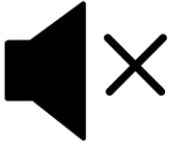


28 May, 2024

Match and Merge Customer Use Cases in MDM SaaS

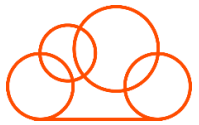
- Sourya Dass, Senior Solutions Architect
- Kamal Abrol, Senior Principal Customer Success Architect

Housekeeping Tips



- Today's Webinar is scheduled for **1 hour**
- The session will include a webcast and then your questions will be answered live at the end of the presentation
- All dial-in participants will be muted to enable the speakers to present without interruption
- Questions can be submitted to "All Panelists" via the **Q&A option** and we will respond at the end of the presentation
- The webinar is **being recorded** and will be available on our [Success Portal](#) - where you can download the **slide deck** for the presentation. The link to the recording will be emailed as well.
- Please take time to complete the **post-webinar survey** and provide your feedback and suggestions for upcoming topics.

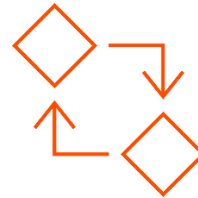
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**Bootstrap trial and
POC Customers**



**Enriched Customer
Onboarding
experience**



**Product
Learning Paths
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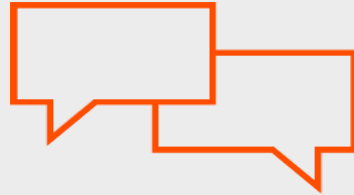
**Tailored training
and content
recommendations**

More Information



Success Portal

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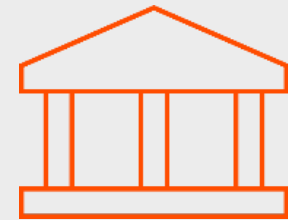
Communities & Support

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Documentation

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<https://www.informatica.com/in/services-and-training/informatica-university.html>

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Industry Centric Data Model & Match Strategy

Thousands of organizations across more than 35 industries have implemented **Informatica Intelligent MDM** and 360 applications, services of **Intelligent Data Management Cloud™** (IDMC). Based on this extensive experience, we came up with five steps to develop the right strategy to implement your MDM solution most effectively:

1. Define your business requirements.
2. Translate business requirements into technical requirements.
3. Select the right tools and vendors.
4. Execute the project.
5. Evaluate your results.

Identify the business opportunities

Get input from your business stakeholders and leaders to determine the key business objectives they wish to support, the challenges they face and the potential opportunities MDM can deliver.

Write down the answers to these questions:

- What are the goals for the business leaders?
- What challenges are they experiencing that impede success?

Examples include:

- Deliver relevant experiences to attract, retain and grow customer base
- Reduce procurement costs
- Accelerate financial close
- Increase cross- and upselling
- Increase ROI from marketing campaigns
- Reduce supplier spend and management cost
- Accelerate new product introduction
- Reduce risk associated with large modernization efforts such as ERP

Define functional requirements

- Functional requirements describe what the software should do. For example:
- The MDM solution should automatically treat two customer records as duplicates if the names, dates of birth, addresses and phone numbers match with a 90% confidence level.
- Data stewards should be able to update multiple records at once to process them quickly and in a controlled manner.
- Business users should be able to capture and visualize person-to-person, person-to-company and company-to-company relationships along with historical views.

Before you configure match and merge, ensure that you meet the following prerequisites:

- Understand the business objectives and requirements for matching and merging duplicate data.
- Analyze the attributes and quality of the data that you want to consider for the match and merge process.
- Determine the size of the data set, which could impact the performance of the match and merge process.
- Determine the match population, which improves match accuracy by accommodating variations and errors that are likely to appear in data for a particular population.

New...

Home

Explore

My Import/Exp...

Business Events

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My Jobs

Security

Global Settings

Person

Person

Click **Match** tab to configure match rules.

Model

Consumption

AttributesData QualityMatchSurvivorshipEvents

Basic Fields ?

T

Text

123

Integer

100

Double

1.23

T

First Name *

T

Middle Name

T

Last Name *

T

Title

T

Designation

T

Birthplace

Marital Status

Address

Phone

Education

Specialization

Financial

+ New...

Home

Explore

My Import/Exp...

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Global Settings

Person

PersonValid

Save

☑

⋮

✕

ModelData FlowConsumption

AttributesData QualityMatchSurvivorshipEvents

Match Models (6)

+

↕

↕

⌵

Find

Name	Version	Objective	Updated On	Status	Model Description
Baseline Match Model for person Deduplication	1	Resolve Duplicates	Apr 12, 2023	● Published	OOTB Match Rule set for person BE.
Person Demo	1	Resolve Duplicates	Jun 22, 2023	● Published	Match Rule set for person BE.
Custom_Match_Model	1	Resolve Duplicates	Jul 05, 2023	● Published	OOTB Match Rule set for person BE.
Person Demo		Duplicates	Dec 06, 2023	● Draft	Match Rule set for person BE.
Custom_Match_Model		Duplicates	Dec 06, 2023	● Draft	OOTB Match Rule set for person BE.
Custom_Match_Model		Duplicates	Dec 06, 2023	● Draft	OOTB Match Rule set for person BE.

You can see that some match rules are configured for you.
Click Custom_Match_Model match rule to check out the details.



- New...
- Home
- Explore
- My Import/Exp...
- Business Events
- Translations
- My Jobs
- Security
- Global Settings

Person

Person > Custom_Match_Model

Training of the machine learning model is in progress. You cannot edit the model.

Model Configuration Declarative Rules Machine Learning Model

Declarative Rules (5)

Rank	Merge Strategy	Match Strategy	Next	Match Level	Description
1	Manual	FUZZY		Loose	Strongly similar name and exactly matching identifier
2	Automated	FUZZY	Resident	Typical	Strongly similar name, address and exactly matching birth date
3	Manual	FUZZY	Person_Name	Loose	
-	Threshold-based	FUZZY	Person_Name	Typical	Similar name and same email
-	Threshold-based	FUZZY	Person_Name	Typical	Similar name and same phone number

Here you can see the declarative match rules that are a set of conditions and business entity attributes that are required to identify duplicate records. Click Machine Learning Model tab to see how intelligence can be added to the matching process.

Next

New...

Home

Explore

My Import/Exp...

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Person

Person

>

Custom_Match_Model

Draft

Publish

Training of the machine learning model is in progress. You cannot edit the model configuration, fields for match training, and declarative rules.

Model Configuration

Declarative Rules

Machine Learning Model

Training the Machine Learning Model

Labeling for batch 1 is in progress. Select conclusive labels for 24 record pairs to complete batch 1.

6

Record Pairs Reviewed

6

Record Pairs with Conclusive Labels

-

Precision

-

Recall

-

Accuracy

Error Matrix

Nothing to display. The error matrix appears after the first batch of record pairs is labeled.

With this approach you can train an ML model based on your business needs, where you can label record pairs . This reduces the manual effort required for defining declarative match rules. Click **Continue Labeling** to see how to label record pairs.

Continue Labeling

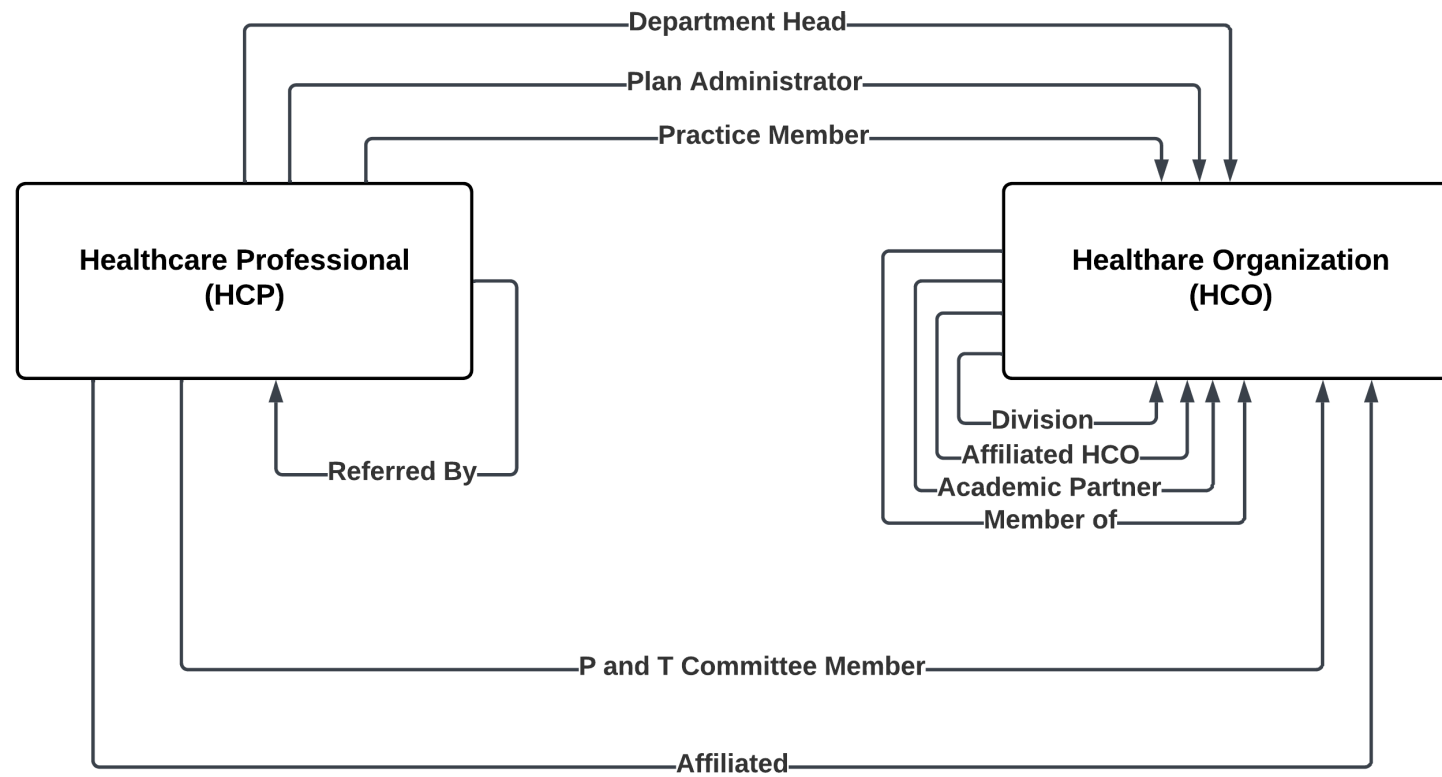
Rule Sequence	Rule Name
1	Ex-Name, Ex-Address, Ex-City, Ex-State, Ex-zip
2	Ex-Name, Ex-Address, Ex-zip
3	Ex-Name, Fuzzy Address, Ex-Zip (Conservative)
4	Ex-Name, Fuzzy Address, Ex-Zip (Typical)
5	Fuzzy-Name, Ex-Address, Ex-Zip (Conservative)
6	Fuzzy-Name, Ex-Address, Ex-Zip (Typical)
7	Fuzzy-Name, Fuzzy Address, Ex-Zip (Conservative)
8	Fuzzy-Name, Fuzzy Address, Ex-Zip (Typical)
9	Fuzzy-Name, Ex Address, Ex City, Ex State, Ex Zip
10	Match Quidel records with OneKey based on OneKey WKP ID

Rank	Merge Strategy	Match Strategy	Match Criterion	Match Level	Description
1	Automated	FUZZY	Person_Name	Typical	DrugEnforcementAgencyNumeric Exact, Full Name Fuzzy
2	Automated	FUZZY	Person_Name	Typical	DrugEnforcementAgencyNumeric Exact, StateLicenseNumeric Exact, FirstName Fuzzy
3	Automated	FUZZY	Person_Name	Typical	DrugEnforcementAgencyNumeric Exact, StateLicenseNumeric Exact, LastName Fuzzy
4	Automated	FUZZY	Person_Name	Typical	NationalProviderIdentifier Exact, Full Name Fuzzy
5	Automated	FUZZY	Person_Name	Typical	NationalProviderIdentifier Exact, StateLicenseNumeric Exact, FirstName Fuzzy
6	Automated	FUZZY	Person_Name	Typical	NationalProviderIdentifier Exact, StateLicenseNumeric Exact, LastName Fuzzy
7	Automated	FUZZY	Person_Name	Typical	StateLicenseNumber Exact, Full Name Fuzzy
8	Automated	FUZZY	Person_Name	Typical	DrugEnforcementAgencyNumeric Exact, SocialSecurityLastFour Exact, FirstName Fuzzy
9	Automated	FUZZY	Person_Name	Typical	DrugEnforcementAgencyNumeric Exact, SocialSecurityLastFour Exact, LastName Fuzzy
10	Automated	FUZZY	Person_Name	Typical	NationalProviderIdentifier Exact, SocialSecurityLastFour Exact, FirstName Fuzzy

11	Automated	FUZZY	Person_Name	Typical	NationalProviderIdentifier Exact,SocialSecurityLastFour Exact, LastName Fuzzy
12	Automated	FUZZY	Person_Name	Typical	StateLicenseNumeric Exact, SocialSecurityLastFour Exact, FirstName Fuzzy
13	Automated	FUZZY	Person_Name	Typical	StateLicenseNumeric Exact, SocialSecurityLastFour Exact, LastName Fuzzy
14	Automated	FUZZY	Person_Name	Conservative	Full Name Fuzzy, SSN Last Four Exact
15	Automated	FUZZY	Address	Typical	DrugEnforcementAgencyNumeric Exact, Geocoded Street Fuzzy, Geocoded City Fuzz, Geocoded ZIP Exact,
16	Automated	FUZZY	Address	Typical	NationalProviderIdentifier Exact, Geocoded Street Fuzzy, Geocoded City Fuzzy, Geocoded ZIP Exact
17	Automated	FUZZY	Address	Typical	StateLicenseNumeric Exact, Geocoded Street Fuzzy, Geocoded City Fuzzy, Geocoded ZIP Exact
18	Automated	FUZZY	Resident	Typical	Full Name Fuzzy, Geocoded Street Fuzzy, Geocoded City, Geocoded ZIP Exact,
19	Automated	FUZZY	Person_Name	Typical	Full Name Fuzzy , TaxID Exact

Industry Centric Data Model & Match Practices

Life Sciences Data Model –HCO & HCP



Comprehensive Life Science Data Model -Key Entities

Health Care Organization

- Subtype
- Location
- Classification
- Segmentation
- Taxonomy
- Taxonomy
- Compliance
- Affiliations

Integrated Delivery Networks/ Accountable Care Organization

- Subtype
- Key people
 - Medical/pharmacy directors, physicians, VPs/Directors of Patient Safety/Innovation, CEOs of affiliated hospitals
- Segmentation
 - Patient demography, geography, organizational structures, payment model

Health Care Professional

- Subtype
- Designation
- Address
- Phone
- Specialty
- Taxonomy
- License
- Degree/Certificate
- Therapeutic areas
- Segmentation
- Indicators (KOL)
- Preferences (Channel, Content)
- Affiliations


Group Purchasing Organization

- Subtype
- Relationships

Contact

- Designation
- Address
- Phone
- Communication

Match Model for HCP Deduplication

 HCP > **Match Model for HCP Deduplication** ● Published ✕

Model Configuration Declarative Rules Machine Learning Model


Declarative Rules (7) Edit

Rank	Merge Strategy	Match Strategy	Match Criterion	Match Level	Description
1	Automated	EXACT	-	-	Exactly similar name and address
2	Automated	FUZZY	Person_Name	Typical	Strongly similar name and license no.
3	Automated	FUZZY	Person_Name	Typical	Strongly similar HCP name and alternate id
4	Automated	FUZZY	Person_Name	Typical	Strictly similar HCP name, address and specialty
5	Manual	FUZZY	Person_Name	Typical	Similar HCP name and address
6	Manual	FUZZY	Address	Typical	Similar HCP last name and address
7	Manual	FUZZY	Person_Name	Typical	Similar HCP alternate name, address and specialty


Candidate Selection Criteria

Field Name	Field Type	Filter Candidates By	Key Generation Level	Candidate Search Level
fullName	Person_Name		Standard	Typical

360 View

 **Alexander Taylor, MD**
Prescriber, Physician
8147242981, MEADVILLE, Pennsylvania

Summary Details Source Records Relationships History

 #ACT for Asthma

Nothing to display.

▼ Communication Preference +

Events

Email

Promotion

In-Person

Preferred Call Time: Afternoon

▼ Personal Information

Prefix Name:

▼ Specialty

▼ Otolaryngic Allergy Physician

Specialty Rank:

Secondary

Specialty:

Otolaryngic Allergy Physician

Taxonomy Code:

207YX0602X

▼ Otolaryngology Physician

Specialty Rank:

Primary

Specialty:

Otolaryngology Physician

Taxonomy Code:

207Y00000X

▼ Plastic Surgery (Otolaryngology) Physician

Specialty Rank:

Secondary

Specialty:

Plastic Surgery (Otolaryngology) Physician

Taxonomy Code:

207Y00000X

▼ License

▼ MD046736L

License Board State Code:

PA

▼ MD046736L

License Board State Code:

PA

▼ MD046736L

License Board State Code:

PA

► DEA

▼ Employment

▼ Sutter Health

▼ Sutter Bay Hospitals

California Pacific Medical Center

California Pacific Medical Ctr - St. Luke'S Campus

California Pacific Medical Ctr-Pacific Campus Hosp

California Pacific Medical center - Davies Campus

Eden Medical Center

Menlo Park Surgical Hospital

Mills-Peninsula Medical Center

Novato Community Hospital

Sutter Lakeside Hospital

Sutter Maternity & Surgery Center of Santa Cruz

Sutter Santa Rosa Regional Hospital

▼ Sutter Valley Hospitals

SUTTER AMADOR HOSPITAL

SUTTER AUBURN FAITH HOSPITAL

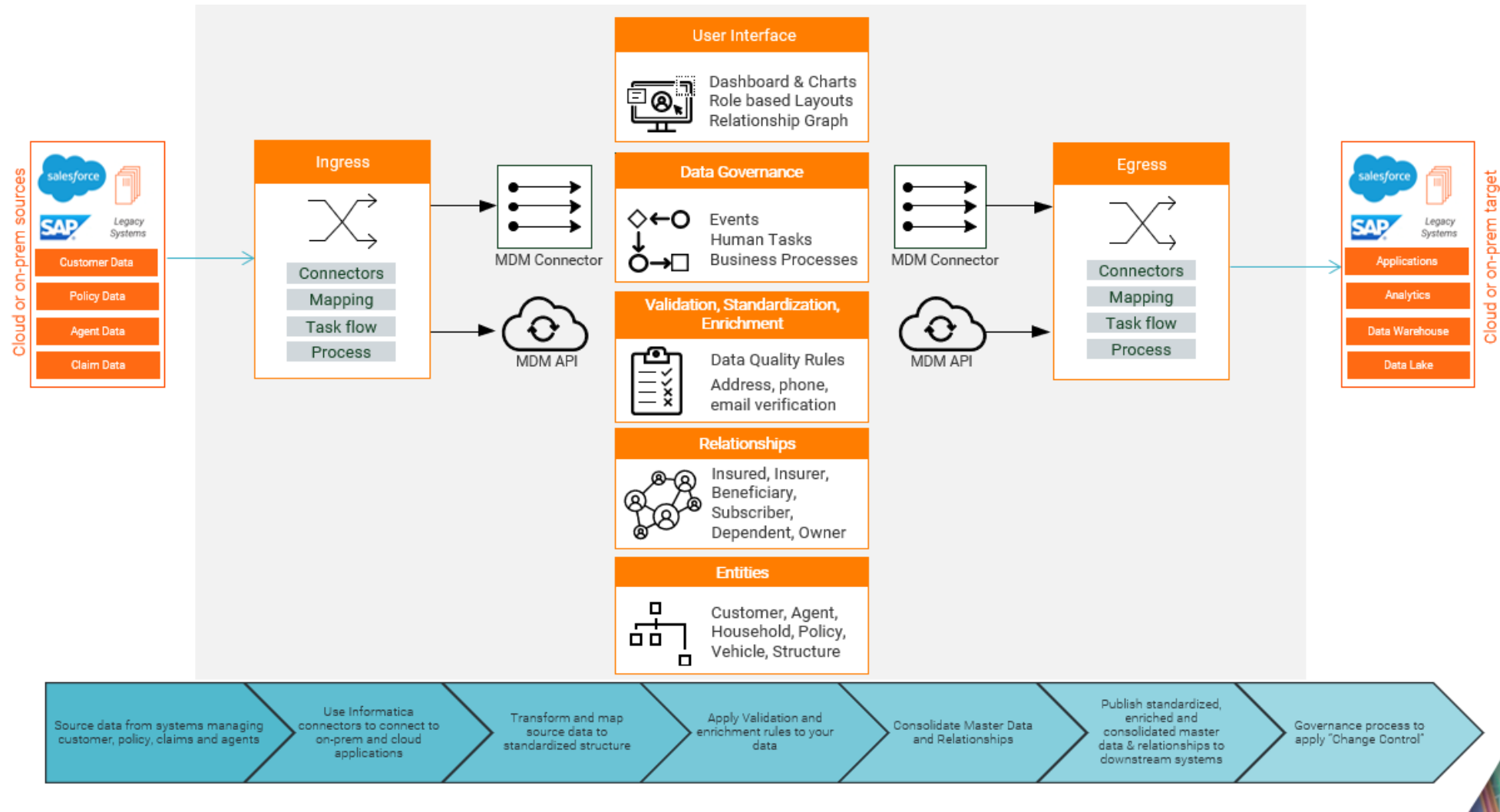
Identifier

Natioanl Provider Identifier
1225036270

```
graph TD; MM((Michigan Medicine)) --- DRA[Douglas A Rottmann]; MM --- SAK[Sara Kaufman Licata]; MM --- BAJ[Bethany Ann Giacobbe]; MM --- AST[Alexander Stuart Taylor]; MM --- DA[David Arnolds]; MM --- NLH[Nicole Louise Hall]; MM --- NAA[Naziheh Assarzadegan]
```

The diagram illustrates a network of healthcare providers. At the center is a yellow circle labeled "Michigan Medicine" with a building icon and a count of 7. It is connected by grey lines to seven green circles, each representing another provider with a person icon and a count of 1. The providers are arranged around the central node: Douglas A Rottmann (top), Sara Kaufman Licata (top-left), Bethany Ann Giacobbe (right), Alexander Stuart Taylor (bottom-right), David Arnolds (bottom), Nicole Louise Hall (bottom-left), and Naziheh Assarzadegan (left).

Insurance Data Model- OOB Extension for IDMC



Key Business Entities

Organization
(Customer, Agent)

Person
(Customer, Agent)

Household

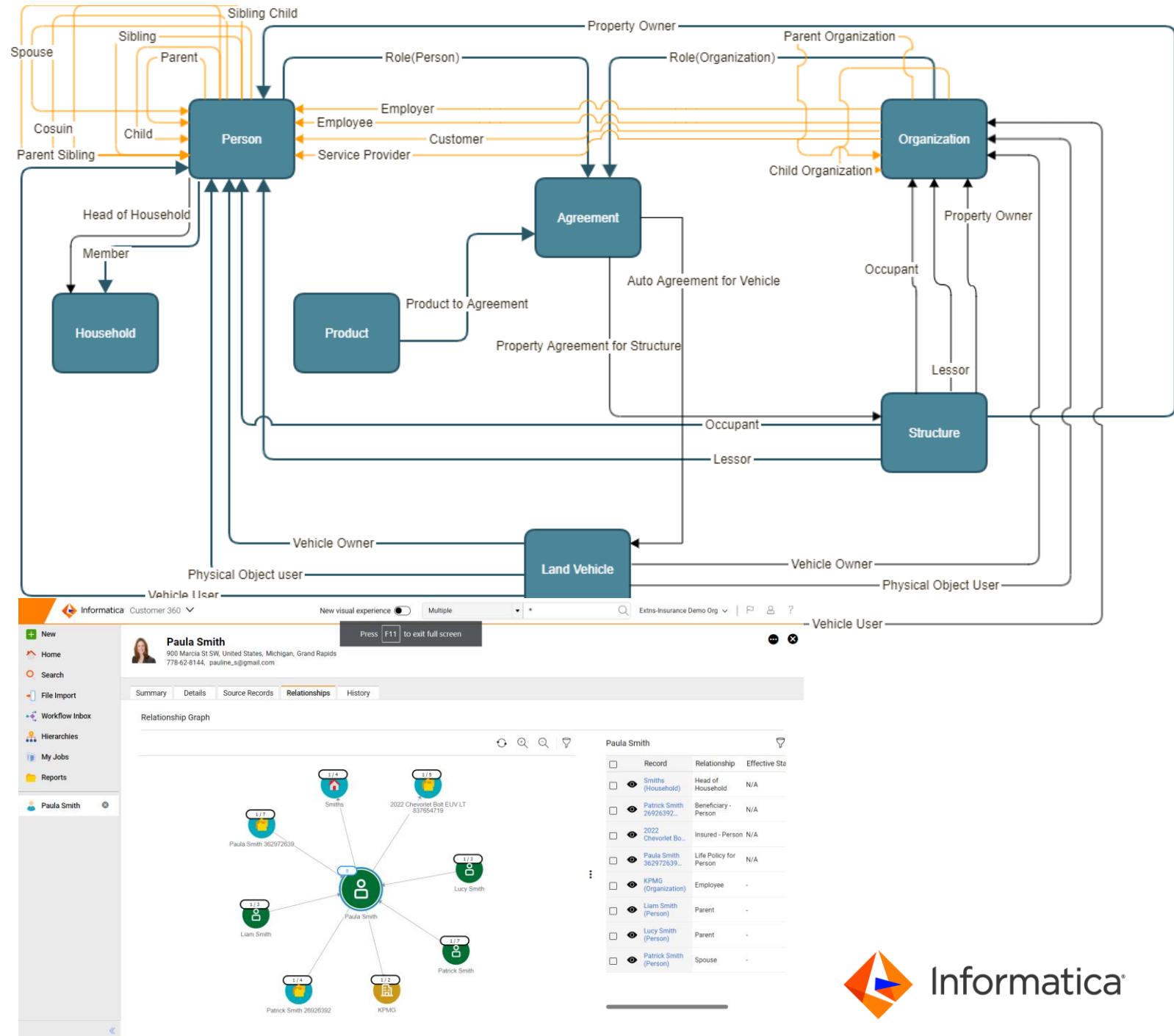
Agreement

Product

Land Vehicle

Structure

Insurance Model-Key Relationships



Match Model –customer/Agent

Model Configuration Declarative Rules Machine Learning Model					
Declarative Rules (5) Edit					
Rank	Merge Strategy	Match Strategy	Match Criterion	Match Level	Description
1	Automated	EXACT	-	-	Exactly similar name and address.
2	Automated	FUZZY	Person_Name	Typical	Strongly similar person name and exactly similar government ID.
3	Automated	FUZZY	Resident	Typical	Strongly similar name, address, and exactly similar alternate ID
4	Manual	FUZZY	Person_Name	Typical	Similar person name and same address
5	Manual	FUZZY	Address	Typical	Similar address and same last name

Match/Merge- Key Pointers

Avoid Hotspots

- Must investigate the high frequency values during the profiling & fix it before matching i.e. Must not use garbage/Test data without enforcing DQ execution for matching ,to avoid job hitting 10k match candidates blocking limit and max 1k Xref for each master record
- Avoid loading transactional Information
- Must not randomly use exact fields like "DOB","CITY","STATE","ZIP" may results in hotspots
- Follow Iterative Process- Fine tune match model with smaller data sets

Match Configuration

- Keep exact match rules even when you configure Fuzzy match Strategy aim to send reduce records for fuzzy matching
- Per profiling results, keep highest matching percentage rule with top Rank.
- Leverage Segment Matching
- Keep only Entity relevant Field Groups , higher number of field groups will slow the match and merge process,

Benefit - Enhance match accuracy + optimal jobs performance

Questions?



Speaker: Kamal Abrol, Sourya
Das



Thank You!

• Customer Success Architect

References

<https://informatica.csod.com/LMS/UserTranscript/OnlineClassView.aspx?qs=%5e%5e%5eYjseWjceYYIHv1Aa%2bnK1ilhZcf2c8x7o7uAmmwFLfpsr2qxeOtv%2bC6pdEoY4VTFnWdXNbFNB0SVEZ6H4Pkcpg68%2bhi4Flx2zS7SJ%2bjOM5GIIVpvYxl1yU%2fJyoRp%2b%2byc>

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<https://docs.informatica.com/master-data-management-cloud/business-360-console/current-version/configure-match-and-merge/preface.html>

<https://infawiki.informatica.com/display/DF/Extension+Releases>