



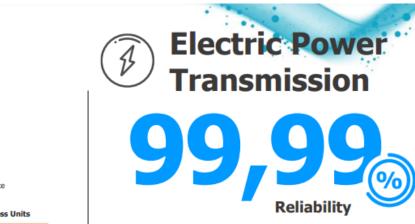




Elkin Cantor

From 2011 to 2014, Elkin Cantor worked at INTERCONEXIÓN ELÉCTRICA S.A (ISA) in Colombia, and he now works at ISA INTERCOLOMBIA S.A. as an asset manager in the maintenance department. He has led several projects involving power transformer maintenance strategy, failure analysis and condition assessments for high voltage substations. Elkin has a bachelor's degree in electrical engineering and a Master's degree in the same field. Currently, he is studying for a Master's degree in data science.





Presence

Business Units



- Management (XM) New Businesses
- |||) Roads
- Telecommunications and ICT's

Sustainability

Corporate Governance

Stakeholders

Financial Information

Strategy

In operation

47.358 Km and 92.720 MVA * Under construction

6.529 Km

and 15.603 MVA

Market share Revenues



70%









by 2030

USD 9.6 billion Investment target



It's the equivalent doing a lap and a half of planet Earth

United States

Mexico

Average time for concessions 25 years

In Colombia and Chile, concessions are perpetual





Artificial Intelligence Opportunities In Business

There are three key use cases for AI in business, which can overlap to some degree, but help to segment the opportunities.

For Example:

- 1. Change the way you understand and interact with customers.
- 2. Offer more intelligent products and services
- 3. Improve and automate business processes

[1]. Artificial Intelligence in Practice: How 50 Successful Companies Used AI and Machine Learning to Solve Problems Book by Bernard Marr

What is not a recommender system

A recommender system is NOT a system that "recommends" arbitrary values.

For Example:

- A system that "recommends" prices for a house you're selling is NOT a recommender system.
- A system that "recommends" whether a transaction is fraudulent is NOT a recommender system

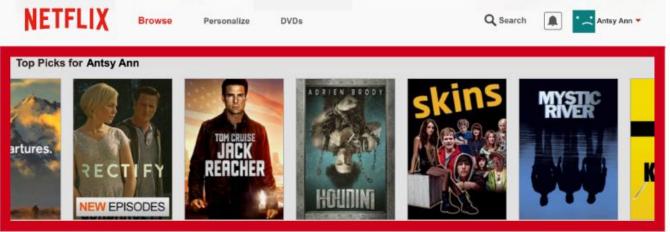
These are general machine learning problems, where you'd techniques such as Regression, deep learning or other techniques. apply

[2]. Building Recommender Systems with Machine Learning and AI: Help people discover new products and content with deep learning, neural networks, and machine learning recommendations. Book by Frank Kane

What is a recommender system

A system that predicts ratings or preferences a user might give to an item. Often these are sorted and presented as "top-N" recommendations

- Netflix's home page
- Google search
- Amazon's "people who bought also bought
- Online radio stations
- Youtube



Why not think in a recomender system for maintenance?

When people apply creativity to the things around them, innovation occurs.

Replicate – The ability to place existing things in a new context. Some of the best creative breakthroughs come from the simple act of placing a preexisting thing, idea or practice in a new context.

Associate – The ability to use analogous thinking to problem-solve. Analogies make the novel feel more familiar or, alternately, encourage people to think of familiar things in new ways.

[3]. The Creative Mindset Mastering the Six Skills That Empower Innovation . Book by Staney DeGraff and Jeff DeGraff

Replication and Association for a Maintenance Recommender system

E-Commerce and Contend Context

- More sales/consumption guiding the user based on their needs.
- Agility in the purchase/consumption process

Asset and Maintenance context

- Proper maintenance based on its failure mode.
 - Agility reducing response times when you have an equipment with a failure



What is the process that we are going to impact?



Maintenance inspection report Preventive and corrective maintenance **Monitoring systems**











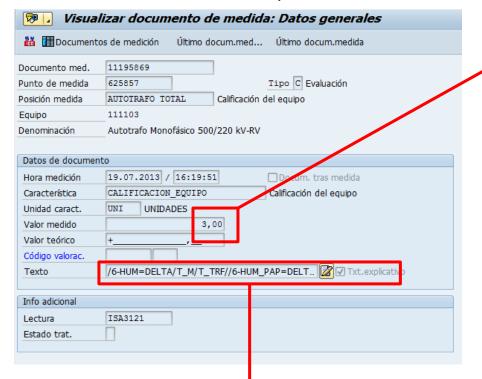
Maintenance Actions to take care ou assets

Recommender System support the making-decision

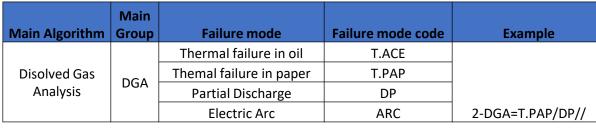


Dataset for the recommender system?

Historical Health Index, since 2010



Diffents levels of conditions



- **Bad Condition**
- Close follow-
- up Good Condition, no need maintenace actions

Dataset for the recommender system

Historical Maintenance Actions for each failure mode Since 2010

Example for a Failure mode in mechanical protection in power transformer

PRO-M



Code in the health index







	ld	
Id Equipment	Maintenance	Description
		Wiring insulation
112135	193845	measurement
123183	193844	Check Buchholz relay
		Calibrate winding
111566	193854	temperature indicator
111569	193862	Change pressure relief valve
111966	193859	Install seal relay

Text Normalization for Natural Language Processing (NLP)

When we normalize text, we attempt to reduce its randomness, bringing it closer to a predefined "standard".

Original Text

The technicians wrote 804 different maintenance

These were reduced to only 173 (80% less)





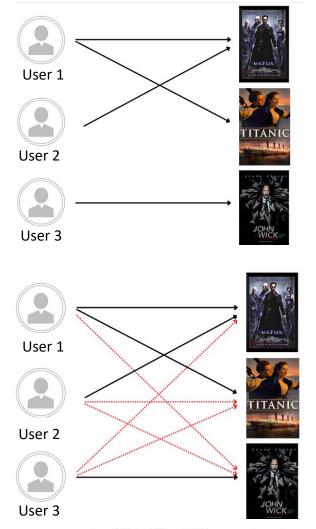
Original Text	Text Normalized
Change Silica gel Breather	
Change Silica gel Breather ATR Phase A	
Change Silica gel Breather ATR2	
Change Silica gel Breather Saturation>75%	
	Change Silica gel Breather
Change Silica gel Breather OLTC	
Change Silica gel Breather RL	
Change Silica gelbreather	

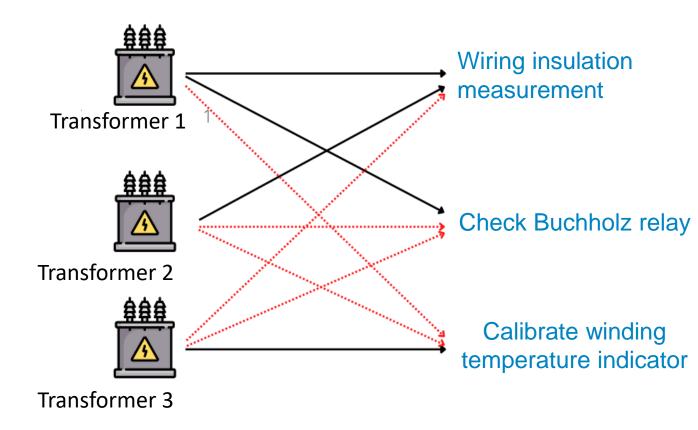
Text Normalization for Natural Language Processing (NLP)

Various techniques were used to standardize the text and reduce the amount of different maintenance.

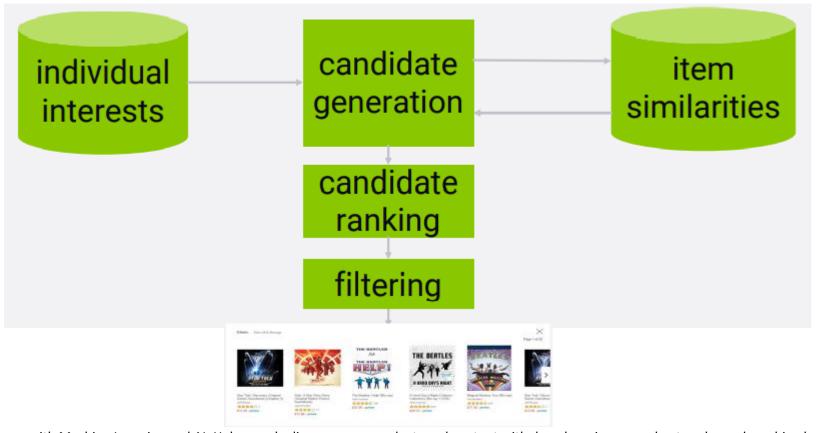
- The creation of dictionaries with keywords that were repeated and allowed to identify the maintenance.
- The Levenshtein distance algorithm measures differences in words that are difficult to recognize by dictionaries.
- Text normalization involves removing words that are deemed unnecessary for understanding the information.

How Do Recommender Systems Work



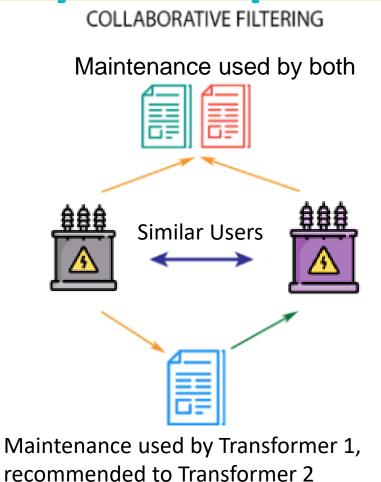


Anatomy of a top-N recommender

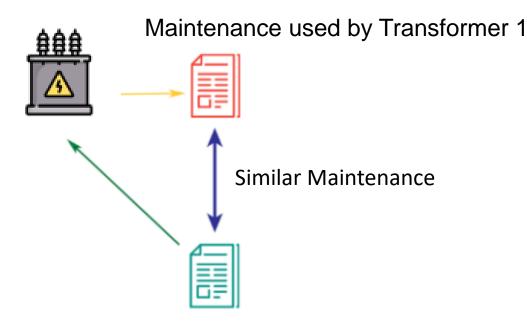


[1]. Building Recommender Systems with Machine Learning and AI: Help people discover new products and content with deep learning, neural networks, and machine learning recommendations. Book by Frank Kane

Anatomy of a top-N recommender



CONTENT-BASED FILTERING



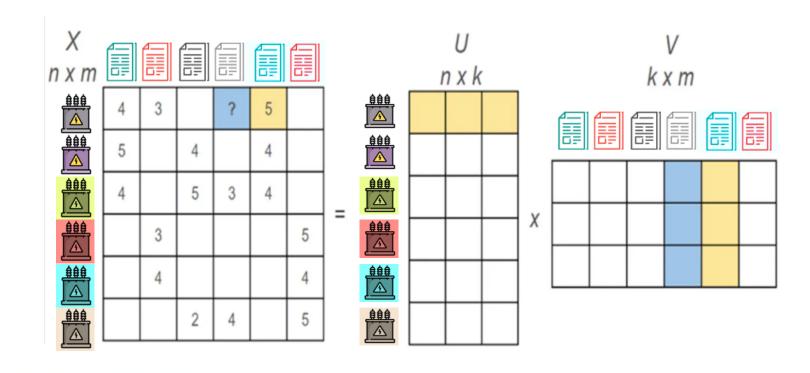
Recommended to Transformer

Singular Value Decomposition (SVD)

It is the model used to obtain the classifications

It is based on the ratings that other users give to dictate new ratings.

n: Users who recommend m: Recommendations to make k: Depth (How many similarly rated users does the model need)



Gathering Ratings

Explicit

• Ask people "Power Transformer" to rate items.

Implicit

Learn ratings from user actions.



In our case, it is based on the frequency of each maintenance action.

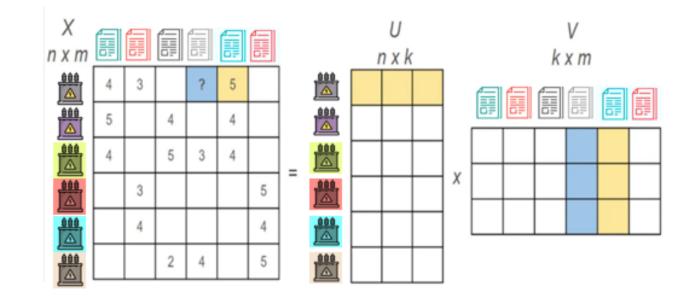


Singular Value Decomposition (SVD)

It is the model used to obtain the classifications

$M = U\Sigma V^T$

- M is the matrix that we want to decompose (m x n)
- U is a unitary matrix of eigenvectors (m x m)
- Σ rectangular matrix, diagonal with eigenvalues (m x n)
- V is a right unitary matrix of eigenvectors $(n \times n)$



The matrix V is calculated by solving the following system of known matrices:

$$V = \frac{1}{\Sigma * M^T * D}$$

Singular Value Decomposition (SVD)

It is the model used to obtain the classifications

$$M = U \Sigma V^T$$

 Our matrix M is going to have many missing values because the users have not qualified all the items. This is where we start to iterate to solve the system.

$$M = \begin{bmatrix} 1 & ? & 3 \\ 2 & 5 & ? \\ 7 & ? & ? \end{bmatrix}$$

To find these values, we can use:

- Traditional algorithms (computationally slow)
- convex optimization
- Neural networks (Simon Funk's algorithm)

The User Interface

Recomendador de Acciones de Mantenimiento





Top N,
Maintenance
are
Recommeded

	Recomendationes					
Equipo	Ubicac.técnica	Denominación •	Fabricante	Calificacion texto	Calificacion numerica	Recomendacion
808997	VAL_PAT	Valledupar 2 100 MVA 220/110 kV	PAUWELS	/9-OIL.TAP//2-INSP=9-ENF/2-PRO-M/8- P.CUB/	2	cambiar aspa motoventildor
808997	VAL_PAT	Valledupar 2 100 MVA 220/110 kV	PAUWELS	/9-OIL.TAP//2-INSP=9-ENF/2-PRO-M/8- P.CUB/	2	cambiar termometro temperatura
808997	VAL_PAT	Valledupar 2 100 MVA 220/110 kV	PAUWELS	/9-OIL.TAP//2-INSP=9-ENF/2-PRO-M/8- P.CUB/	2	cambiar transformador potencia
808997	VAL_PAT	Valledupar 2 100 MVA 220/110 kV	PAUWELS	/9-OIL.TAP//2-INSP=9-ENF/2-PRO-M/8- P.CUB/	2	proteccion mecanica
808997	VAL_PAT	Valledupar 2 100 MVA 220/110 kV	PAUWELS	/9-OIL.TAP//2-INSP=9-ENF/2-PRO-M/8- P.CUB/	2	revisar disparo deshabilitado

Recomendaciones

Failure Mode

Aviso Cerrado

Avisos Previos Si

Maintenance s are already planned

Aviso	Equipo	Descripción	Modo Falla	Fecha de cierre	Creado el	Fin deseado	Inicio desea^
232252	808997	Cambiar empaques y silica ATR 1 Dispac	PRO-M	jueves, 31 de marzo de 2011	viernes, 25 de marzo de 2011	viernes, 25 de marzo de 2011	viernes, 25 c
231474	808997	Cambiar silicagel ATR R	PRO-M	viernes, 29 de abril de 2011	miércoles, 09 de marzo de 2011	miércoles, 06 de abril de 2011	miércoles, 0
232227	808997	CAMBIAR SILICA GEL ATR2	PRO-M	viernes, 06 de mayo de 2011	jueves, 24 de marzo de 2011	miércoles, 20 de abril de 2011	miércoles, 2
468313	808997	Cambiar sílica gel por saturación > 75%	PRO-M	martes, 31 de mayo de 2011	martes, 19 de abril de 2011	jueves, 26 de mayo de 2011	martes, 17 c
702529	808997	Cambiar sílica del por saturación > 75%	PRO-M	martes. 31 de mavo de 2011	martes. 19 de abril de 2011	viernes. 27 de mavo de 2011	martes. 17 cv

What problems are going to be solved?



Support decisions in maintenance always taking into account the data



Knowledge <u>management (KM)</u>



Improve risk management by reducing response times to a failure.

18 days on average to create a maintenance after a failure



Better efficiency in defining maintenance actions

277 days on average to change the condition



Next Steps

1) Automate data extraction from our ERP













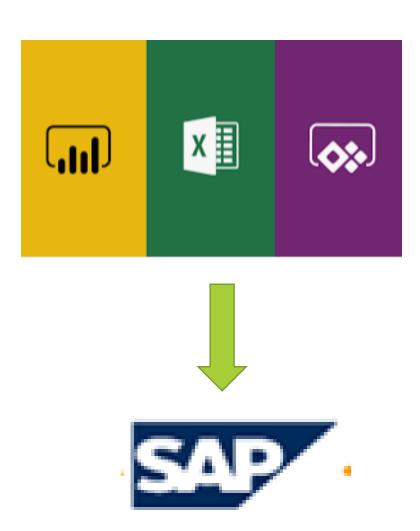




Next Steps

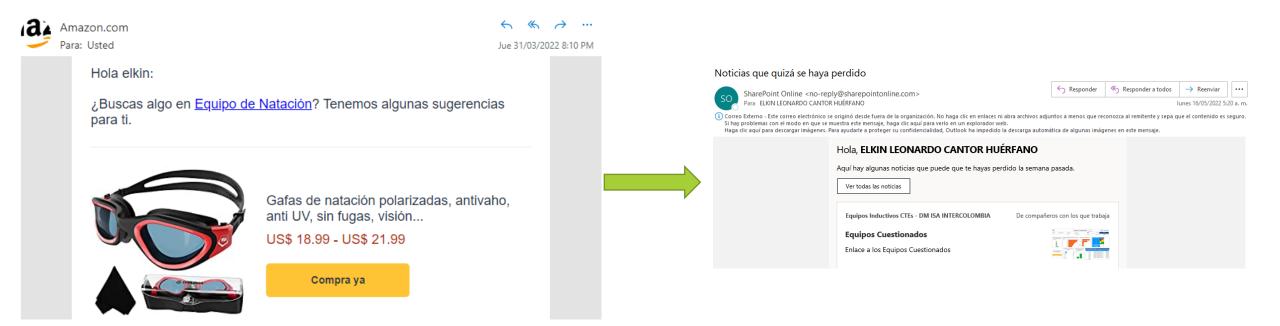
2) Automate data extraction from our ERP

The recommendation and maintenance processes are easily linked.



Next Steps

3) Personalization of the recommendation can help to improve the model.



Questions?

Thanks Elkin Cantor ecantor@intercolombia.com