



# Merchandising Analytics



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Subgerente de Analítica Comercial

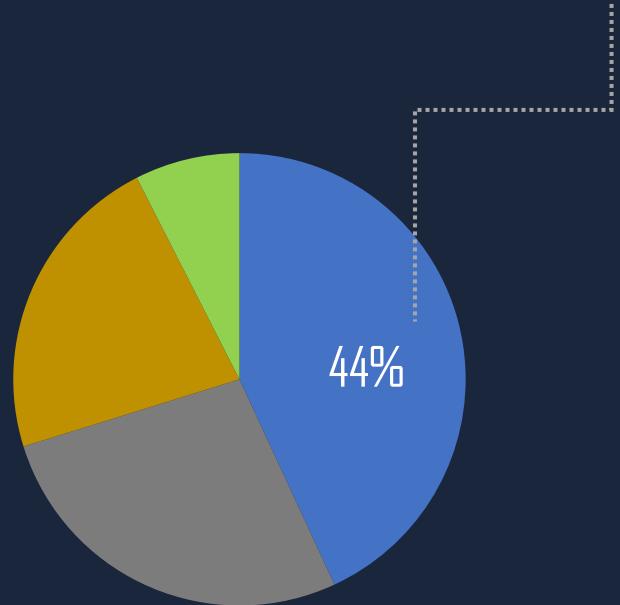


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Senior Data Scientist



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Senior Data Scientist

# Walmart Chile



*Lider* \*

*express*  
de LIDER

SuperBodega  
*aCuenta*®

\**central*  
mayorista

*Lider.cl*

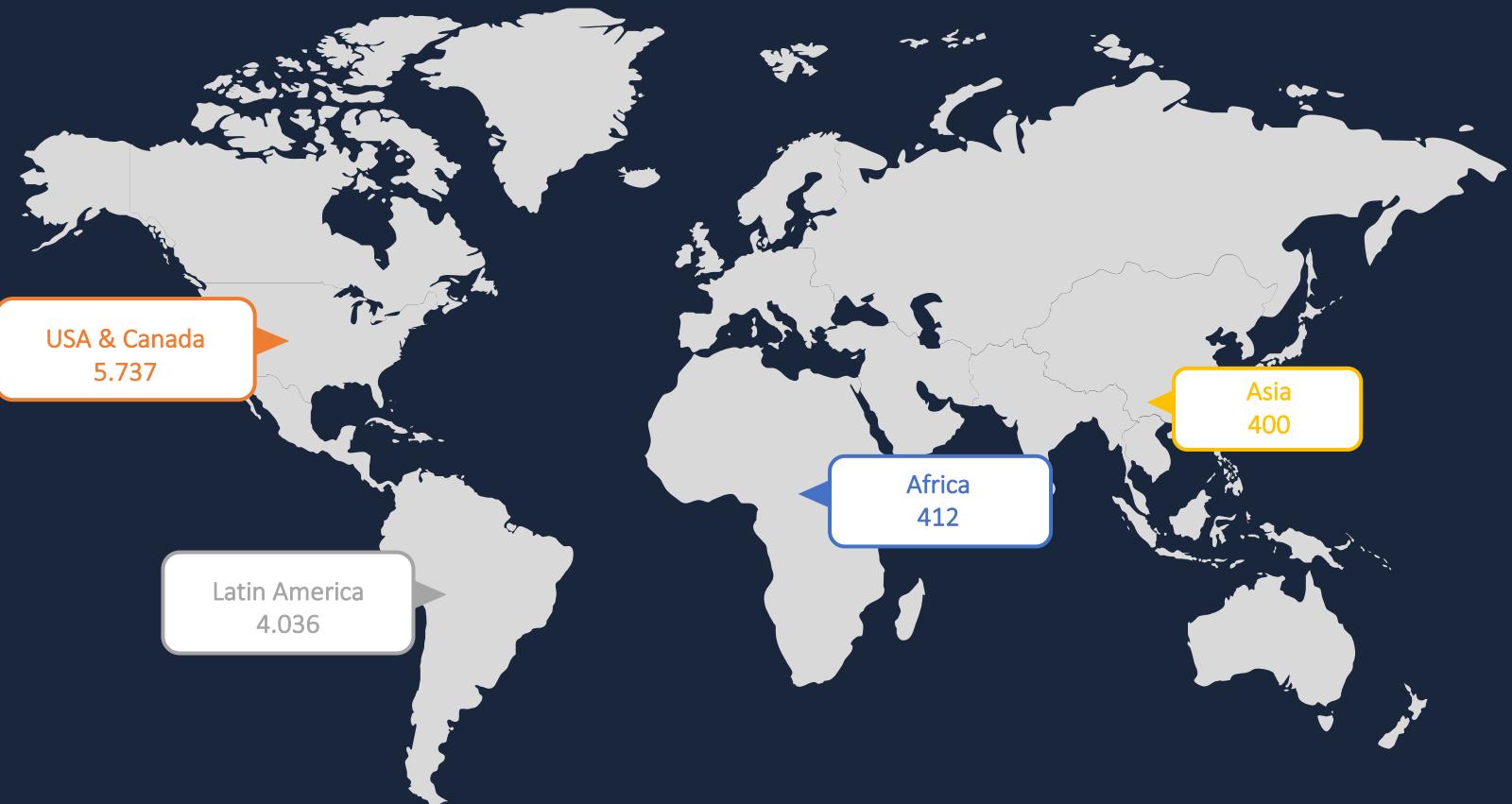


**Walmart**  
Save money. Live better.

10,585 stores

2.3 million associates

Revenue of US\$572 billions



# ¿Qué es el Retail?

1

**BUY**

- Sourcing & Negotiation
- Planning & Revenue
- Space & Assortment
- Pricing & Promotions
- Item & Catalog

2

**MOVE**

- Imports
- Warehouses
- Logistics & Routing
- Supply chain
- Last mile delivery

3

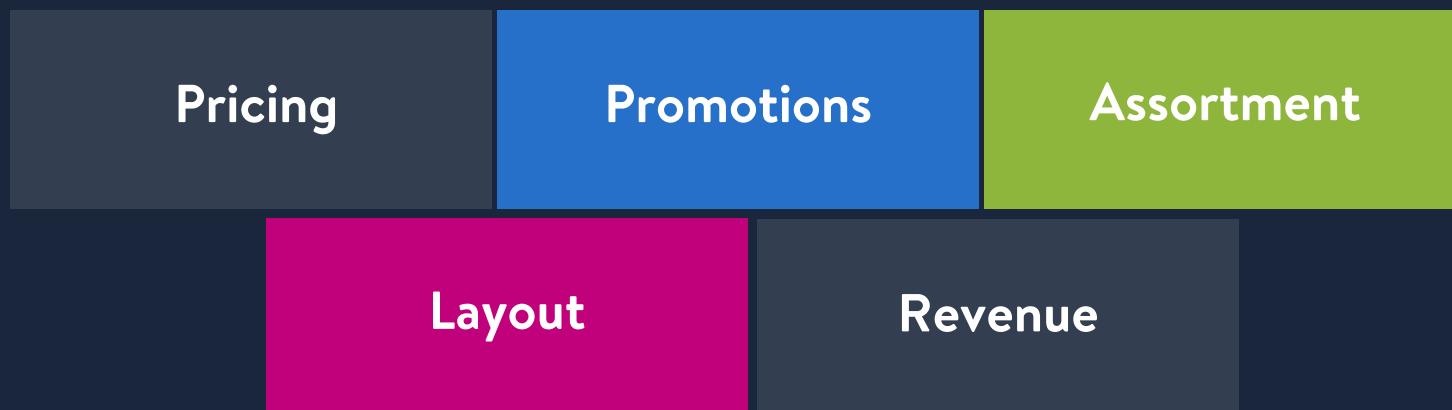
**SELL**

- Store location
- Workforce scheduling
- Digital experience
- Customer Relationship Management (CRM)

385 stores, 60000 products, 50000 associates  
4.5 millions customers  
8000 millions scans

# Merchandising Analytics

To provide digital solutions for Walmart Chile business issues,  
by using **new technologies, mathematical modeling,**  
and the ability to process **large amounts of data.**



# Association rules

Association of  
products or  
sets of products

7804642160138	PAN ARABE..	\$ 990
7809611707444	PECHUGA DESH	\$ 3.690
7802575201218	NIDO FETTUC	\$ 1.030
7808743602146	AVENA INSTAN	\$ 770
780504500144	JABON LAVAR	\$ 3.490
7803468003025	MOLDE LIGHT	\$ 1.390
7802920000916	YOGHURT LIGH	\$ 220
7802920000916	YOGHURT LIGH	\$ 220
7802920000916	YOGHURT LIGH	\$ 220
7613036756440	NUBE PINA	\$ 380
7613036756440	NUBE PINA	\$ 380
7613036756440	NUBE PINA	\$ 380
7613036756440	NUBE PINA	\$ 380
7802200400047	MERM DURAZNO	\$ 970
7613036443166	NUBE COCO	\$ 380
7613036443166	NUBE COCO	\$ 380
7613036443166	NUBE COCO	\$ 380
7613036443166	NUBE COCO	\$ 380
7802710350504	CASSATTA BRI	\$ 2.190
7802200400061	MERM DAMASCO	\$ 970
7613035180703	PACK CONDENS	\$ 1.490
7802615006551	ARROZ GRADO	\$ 1.150
7802926000682	HELADO AGUA	\$ 2.690

$$supp(X) = \frac{|\{t \in D; X \subseteq t\}|}{|D|} = \frac{C_x}{|D|} = P(X)$$

$$conf(X \Rightarrow Y) = \frac{supp(X \cup Y)}{supp(X)} = \frac{P(X \cap Y)}{P(X)} = \frac{C_{xy}}{C_x} = P(Y|X)$$

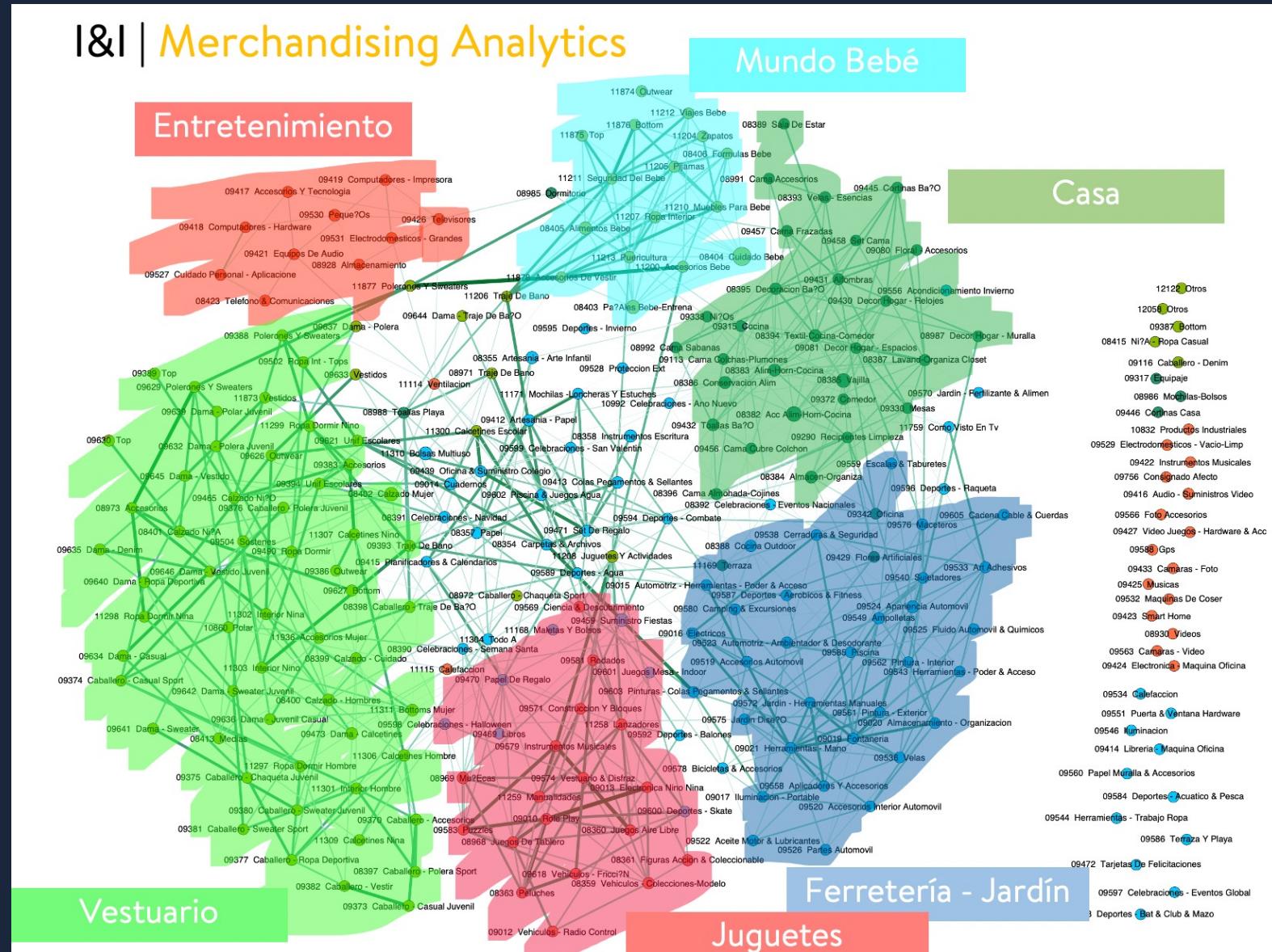
$$lift(X \Rightarrow Y) = \frac{conf(X \Rightarrow Y)}{supp(X)} = \frac{P(X \cap Y)}{P(X)P(Y)} = \frac{C_{xy}}{C_x C_y} |D|$$

# Association rules

STORE_NBR	DPFL_X	Fineline_Desc_X	DPFL_Y	Fineline_Desc_Y	C_xy	C_x	C_y	Lift
3	21100	ACOND FAMILIAR	23600	SHAMPOO FAMILIAR	14991	30140	43910	52.3
3	21075	ACOND TOD/TIPO CAB	23525	SHAMPOO CABELLO SECO	10035	62870	24799	29.7
3	924500	CREMAS	924525	SOPAS	24650	58176	78410	25
3	21075	ACOND TOD/TIPO CAB	23475	SHAMPOO CABELLO NORM	23979	62870	72489	24.3
8	922750	TODA OCASION	928350	TENTACION	17077	38506	54085	23.6
616	924500	CREMAS	924525	SOPAS	21864	42975	58263	23.5
616	21100	ACOND FAMILIAR	23600	SHAMPOO FAMILIAR	29500	52975	64427	23.3
3	921875	DOYPACK KETCHUP	921950	MOSTAZA REGULAR	13060	73408	39251	20.9
616	21075	ACOND TOD/TIPO CAB	23525	SHAMPOO CABELLO SECO	10182	57335	23325	20.5

# Association rules

I&I | Merchandising Analytics



# Pricing optimization

Optimal price  
for each product  
at each store.

$$\begin{aligned}
 & \max \sum_{i \in \mathcal{I}} \sum_{t \in \mathcal{T}} (p_{i,t} - c_{i,t}) q_{i,t} \\
 \text{s.t.} \\
 (1) \quad & \frac{\sum_{i \in \mathcal{I}_k} \frac{p_{i,t}}{cp_{i,t}} v_{i,t}}{\sum_{i \in \mathcal{I}_k} v_{i,t}} \leq pok_k \quad \forall k \in \mathcal{K}, t \in \mathcal{T}, \\
 (2) \quad & \frac{p_{i,t}}{cp_{i,t}} \leq poi_{i,t} \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, \\
 (3) \quad & p_{i,t} = p_{j,t} + \delta_{i,j}^1 \quad \forall (i, j) \in \mathcal{L}_p, t \in \mathcal{T}, \\
 (4) \quad & p_{i,t} = p_{j,t} \delta_{i,j}^2 \quad \forall (i, j) \in \mathcal{L}_f, t \in \mathcal{T}, \\
 (5) \quad & p_{i,t} = p_{j,t} \delta_{i,j}^3 \quad \forall (i, j) \in \mathcal{L}_{mp}, t \in \mathcal{T}, \\
 (6) \quad & p_{i,t} = fp_{i,t} \quad \forall i \in \mathcal{F}, t \in \mathcal{T}, \\
 (7) \quad & p_{i,t} \geq 1.19c_{i,t} \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, \\
 (8) \quad & \frac{op_{i,t} - p_{i,t}}{pp_{i,t}} \geq 0 \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, \\
 (9) \quad & p_{i,t} \leq \max \left\{ \frac{1.19c_{i,t}}{1 - (om_{i,t} + mg)}, pp_{i,t}(1 - mv_k^b) \right\} \quad \forall (i, t) \in \mathcal{NP}, k \in \mathcal{K}, v_{i,t} > 0, \\
 (9^*) \quad & p_{i,t} \geq \min \left\{ \frac{1.19c_{i,t}}{1 - (om_{i,t} + mg)}, pp_{i,t}(1 + mv_k^\#) \right\} \quad \forall (i, t) \in \mathcal{NP}, k \in \mathcal{K}, v_{i,t} = 0,
 \end{aligned}$$

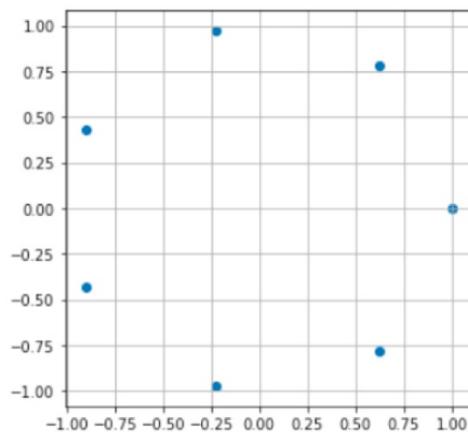
# Forecast modeling

1. ↵ Data pre-processing
2. 🚧 Feature Engineering (\*)
3. 📈 Trend Estimation (\*)
4. 💥 Model Training (\*)
5. 🔐 Post-Processing

Cyclical representation for day of week

```
day_of_week = np.array([0,1,2,3,4,5,6])
dw_sin = np.sin(day_of_week*(2.*np.pi/7))
dw_cos = np.cos(day_of_week*(2.*np.pi/7))

plt.figure(figsize=(5,5))
plt.grid()
plt.scatter(dw_cos, dw_sin)
plt.show()
```



$$y(t) = \underbrace{T(t)}_{\text{Trend}} + \underbrace{S(t)}_{\text{Seasonality}} + \underbrace{H(t)}_{\text{Holidays}} + \underbrace{E(t)}_{\text{Exogenous*}} + \epsilon_t$$

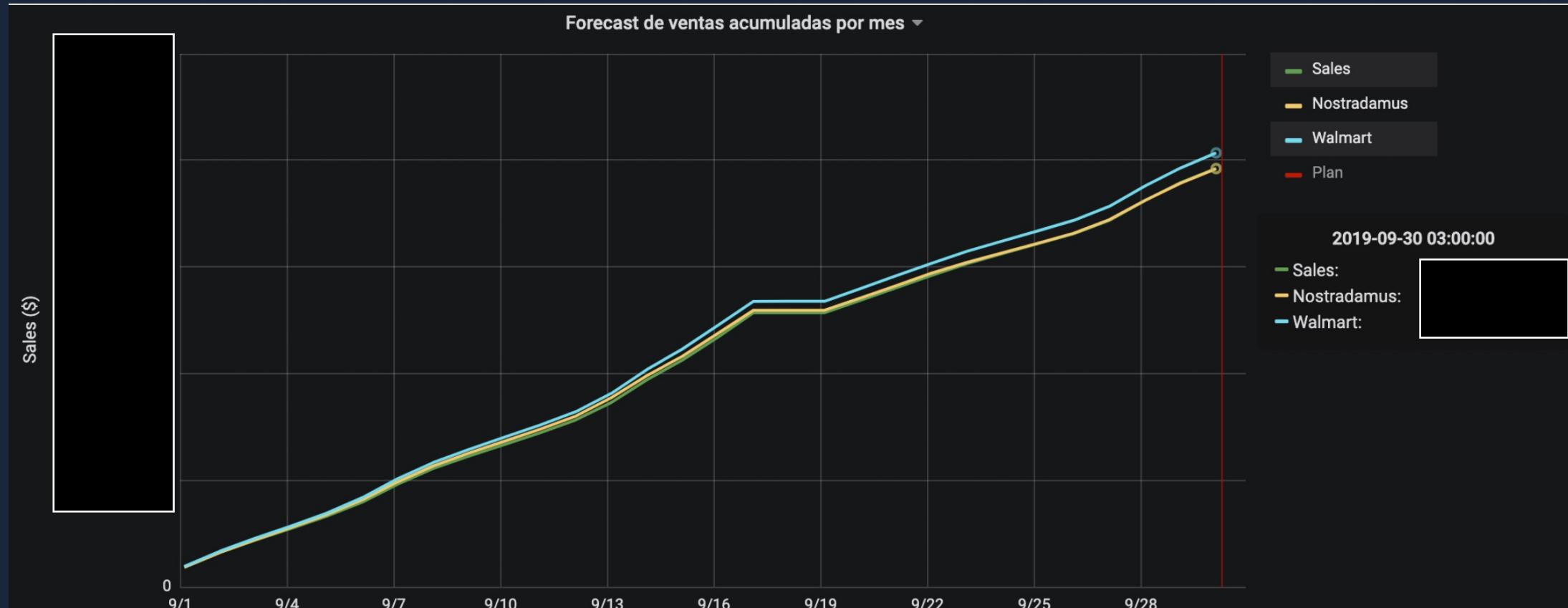


Training ends when error stops decreasing on validation set.  
**best\_iteration:** iteration where validation error reaches its minimum.



Trains over the full training set for **kappa \* best\_iterations** rounds.

# Forecast modeling

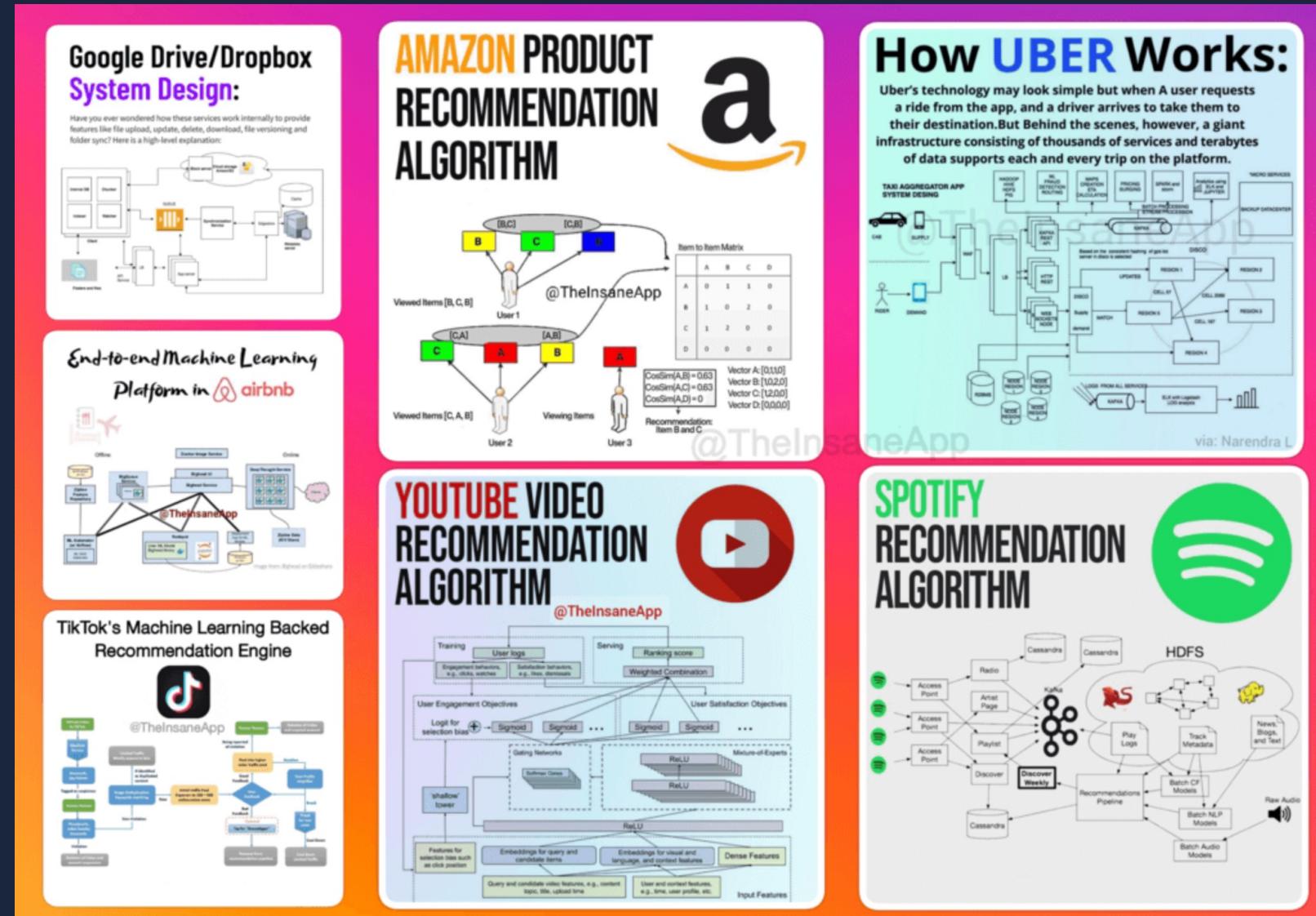


September 2019:

Legacy forecast accuracy: 3.46%  
Nostradamus accuracy: -0.10%

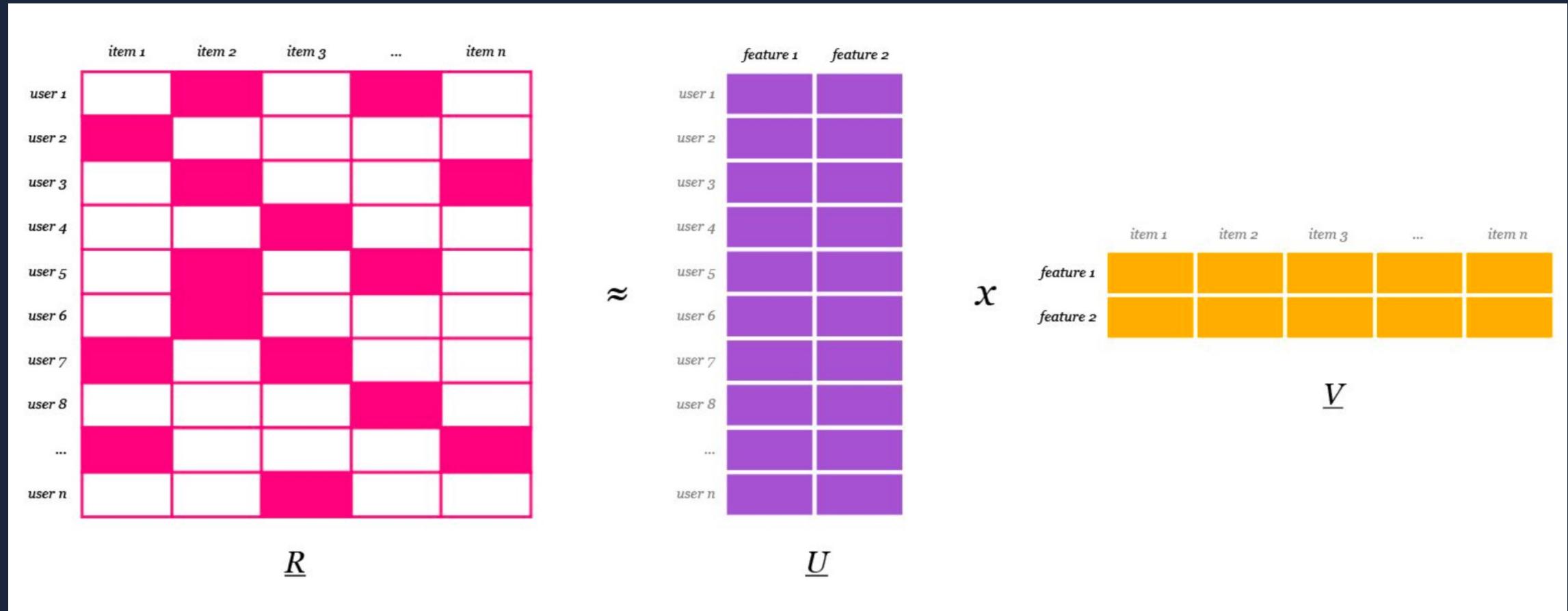
30x improvement  
In forecast accuracy

# Recommendation systems



Source:  
TheInsaneApp  
2021

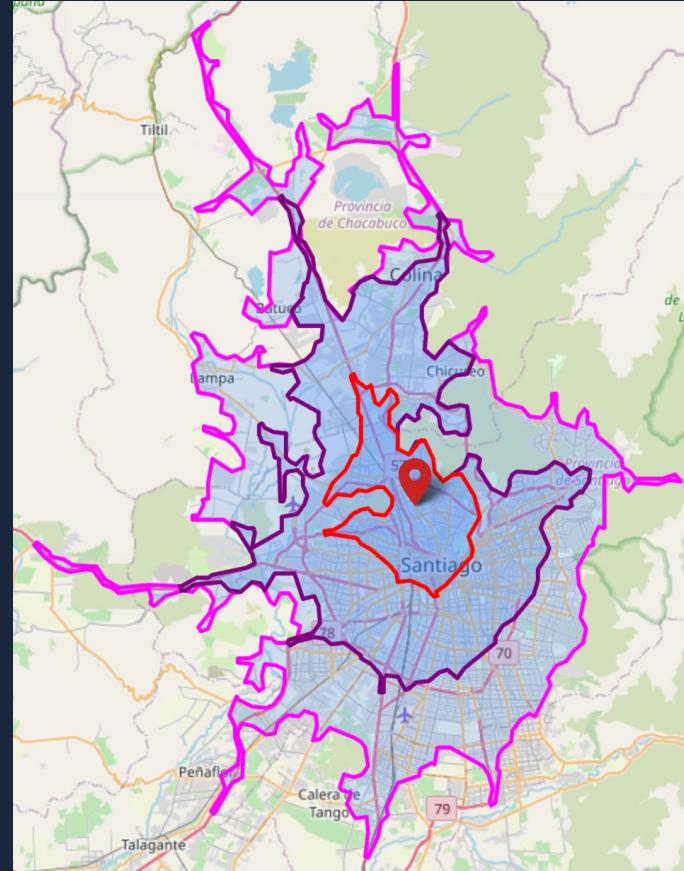
# Recommendation systems



# Georeferenced data

Use of publicly available data  
to leverage our solutions to  
georeferenced problems.

Fase	Acumulado	Porcentaje
Total compañía	15 878 851	90.8%
Total personas		17 482 610



THE  
**NEW YORKER**

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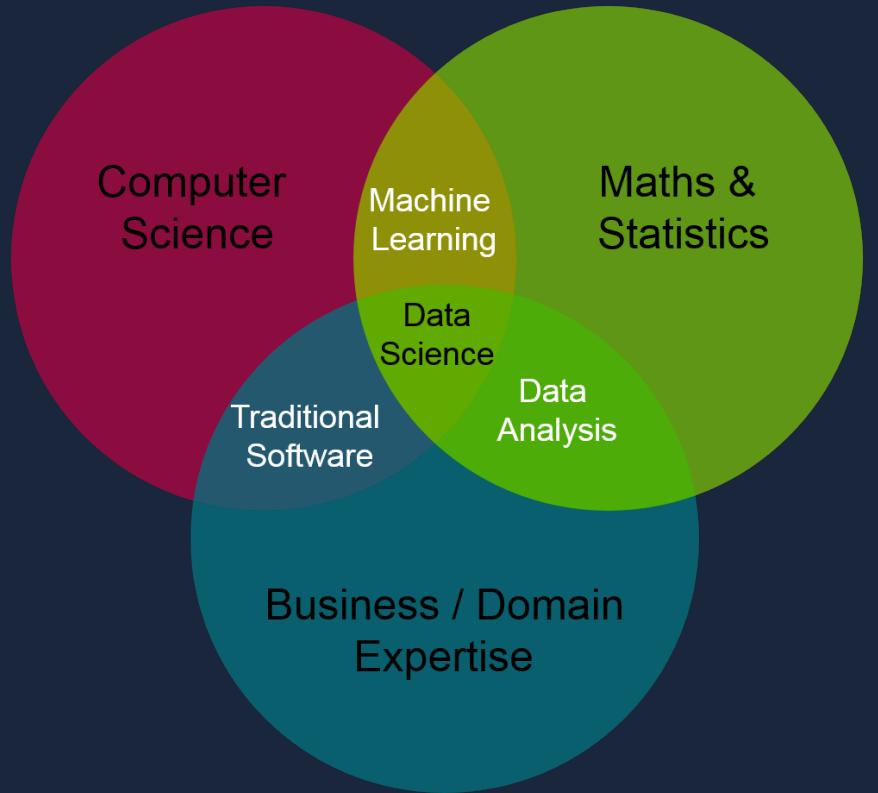
ANNALS OF COMMUNICATIONS

# HOW THE MATH MEN OVERTHREW THE MAD MEN


 viernes Ediciones anteriores La revista de laSegunda laSegunda

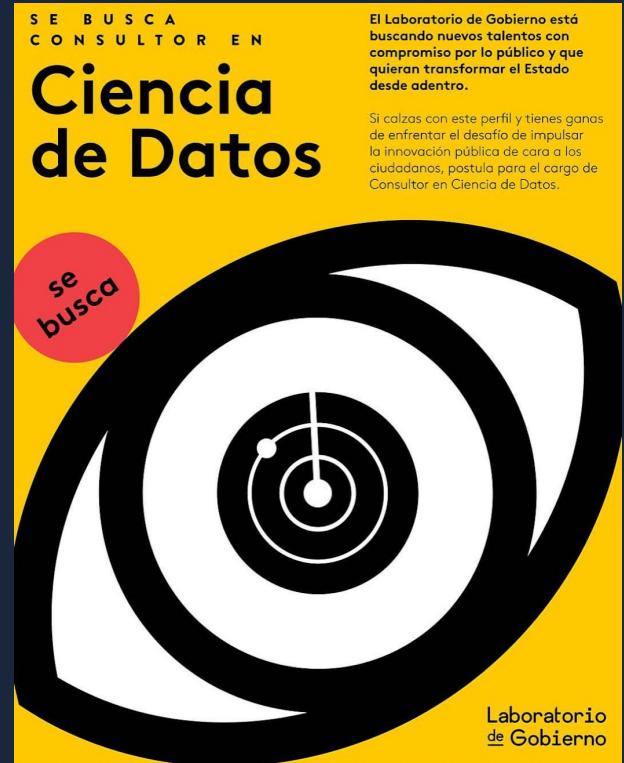
**viernes\_reportaje**

# LA NUEVA INFLUENCIA DE LOS MATEMÁTICOS

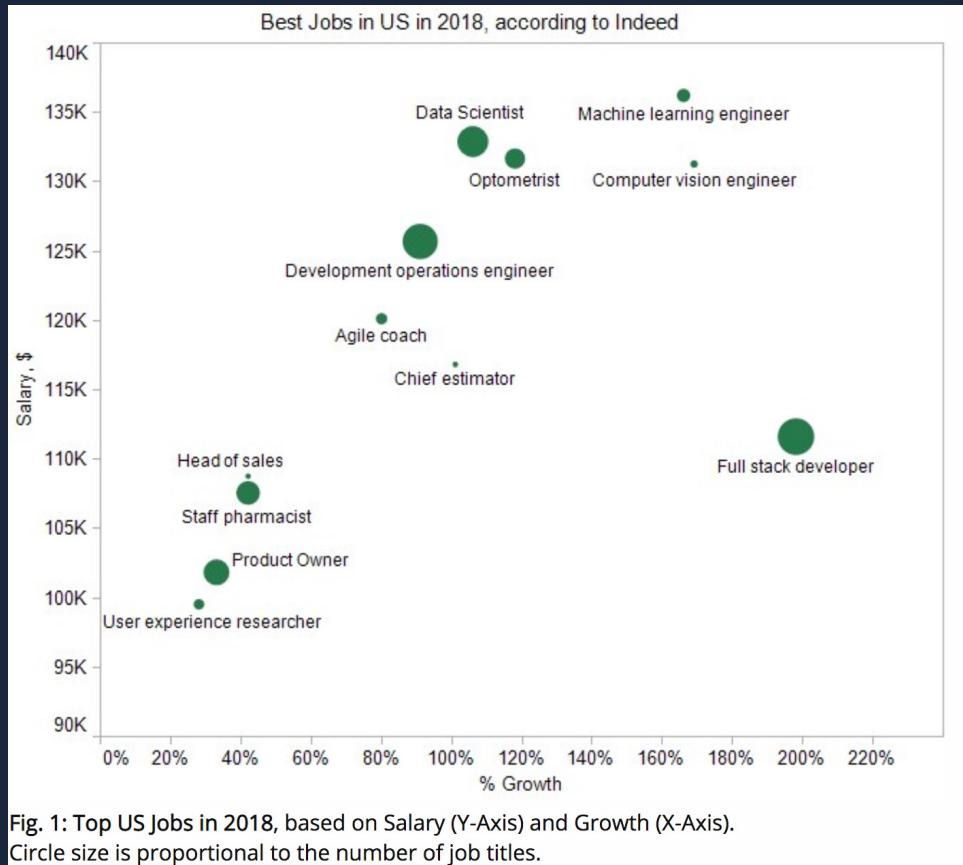


Hay empleo!

	<b>Data Scientist Área Analítica Avanzada (Las Condes)</b> Coca-Cola Emboron Gran Santiago, Región Metropolitana de Santiago, Chile
	<b>Data Scientist</b> EY Gran Santiago, Región Metropolitana de Santiago, Chile
	<b>Data Scientist - Santiago</b> Teck Resources Limited Las Condes, Región Metropolitana de Santiago, Chile
	<b>Digital Data Scientist</b> Nestlé Gran Santiago, Región Metropolitana de Santiago, Chile
	<b>Data Scientist</b> Banco de Chile Gran Santiago, Región Metropolitana de Santiago, Chile
	<b>CL_CIENTIFICO/A DE DATOS</b> Telefónica Gran Santiago, Región Metropolitana de Santiago, Chile
	<b>Data Scientist</b> IFG Capital Viña del Mar, Región de Valparaíso, Chile (Presencial)
	<b>Data Scientist</b> Gimnasios Energy Club Estación Central, Región Metropolitana de Santiago, Chile



[www.somoswalmartchile.cl](http://www.somoswalmartchile.cl)



Recursos:

<https://www.kdnuggets.com/>

<https://datasciencecl.slack.com/>

<https://www.datasciencecentral.com/>

<http://blog.kaggle.com/>

# Teamwork



Optimization:  
Mathematical modeling and optimization of product assortment.  
Price-demand elasticity and price optimization.

Time series:  
Hierarchical or grouped forecast of time series.

Data mining:  
Asymmetrical association rules.



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