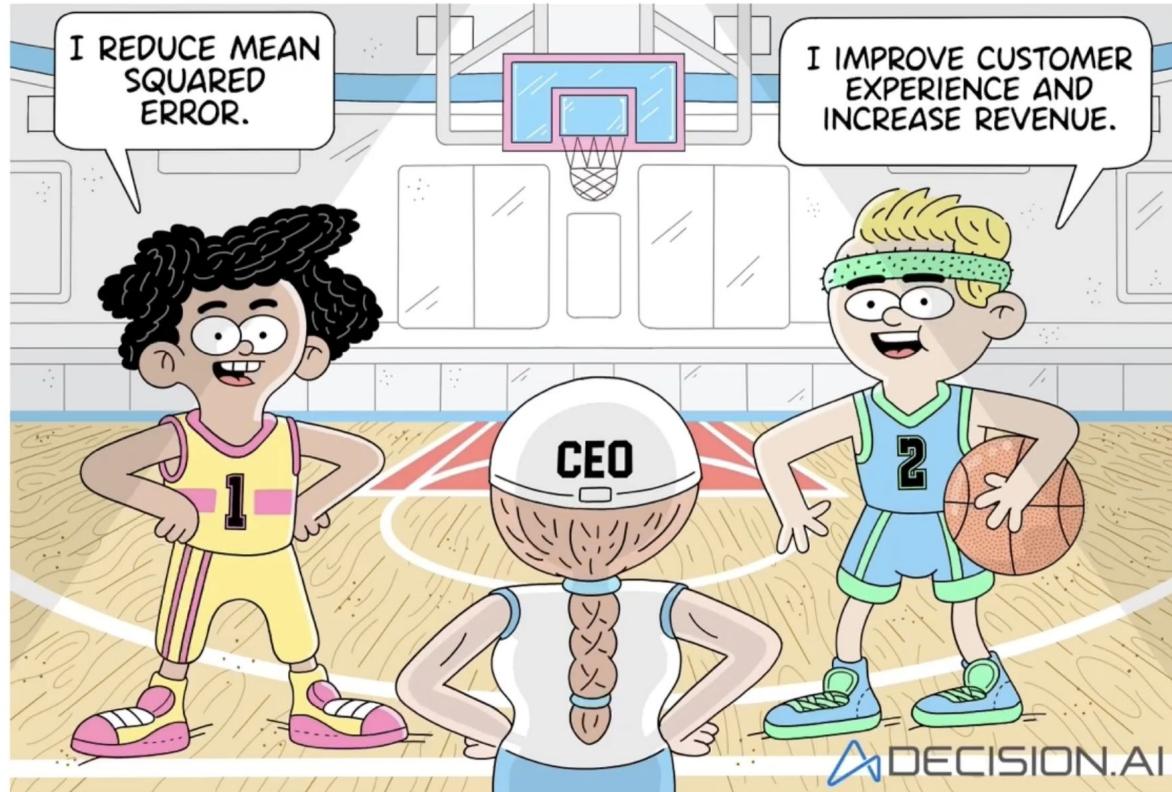


# Conectando métricas de negocio con nuestros modelos de ML.

**Por:** Jose Alberto Arango S.



# ¿Cuál DS crees que ella elegirá?



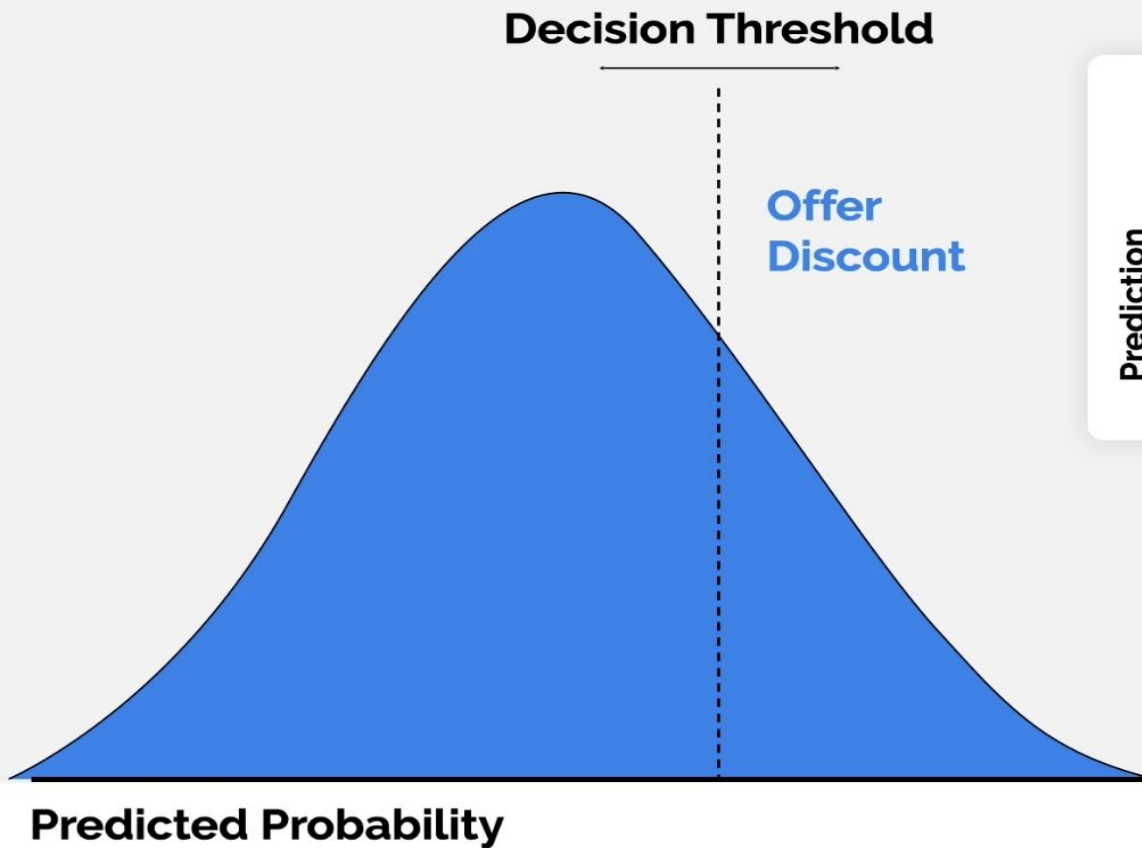
# Problema

## Standard Churn Model Scenario

 Churn	 Contract	 Years as Customer	 Home Phone
0	Monthly	7	0
0	Annual	1	0
1	Monthly	3	1
0	Monthly	3	0

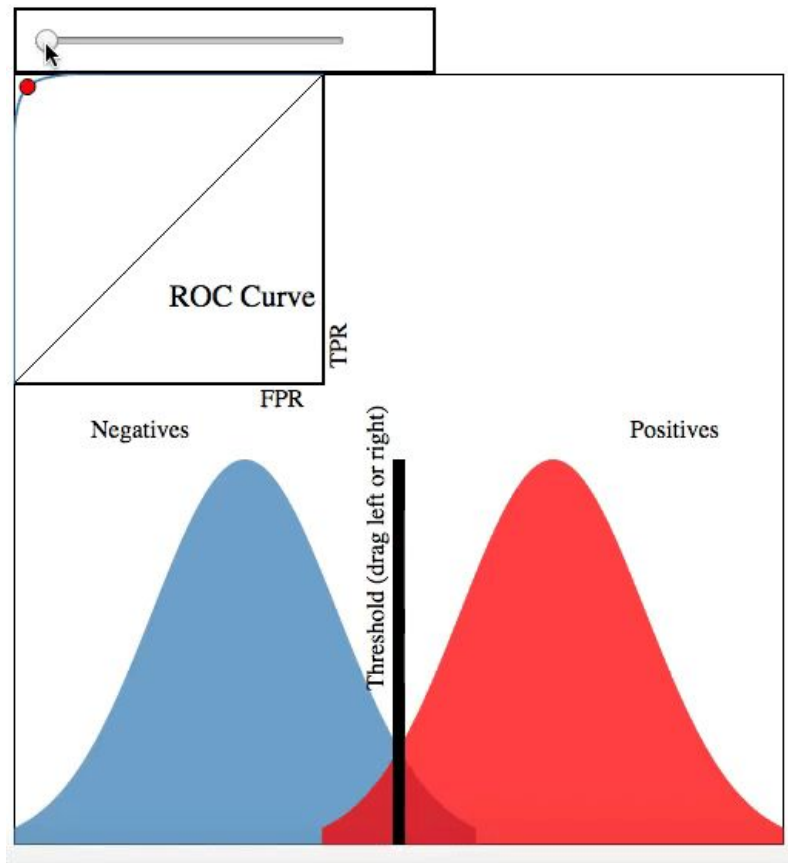
 Prediction Target

# Course Outline



		Churn	
		Yes	No
Prediction	Yes	150,000	80,000
	No	20,000	3,000,000

# ¿Cómo escogemos el umbral de decisión?



# ¿Cómo escogemos el umbral de decisión?

```
threshold = 0.5  
y_pred_binary = np.where(y_pred >= threshold, 1, 0)
```





# Profit Curve!

Medir el rendimiento de  
nuestros modelos ML en  
términos de financieros (\$).

# Profit matrix

		Churned	
		Yes	No
Discount	True	\$200	-\$30
	False	\$0	\$0



# Profit matrix

		Sale	
		Yes	No
Call	True	\$5	-\$2
	False	\$0	\$0

# Profit matrix

		Collection	
		Yes	No
Call	True	\$10	-\$2
	False	\$0	\$0

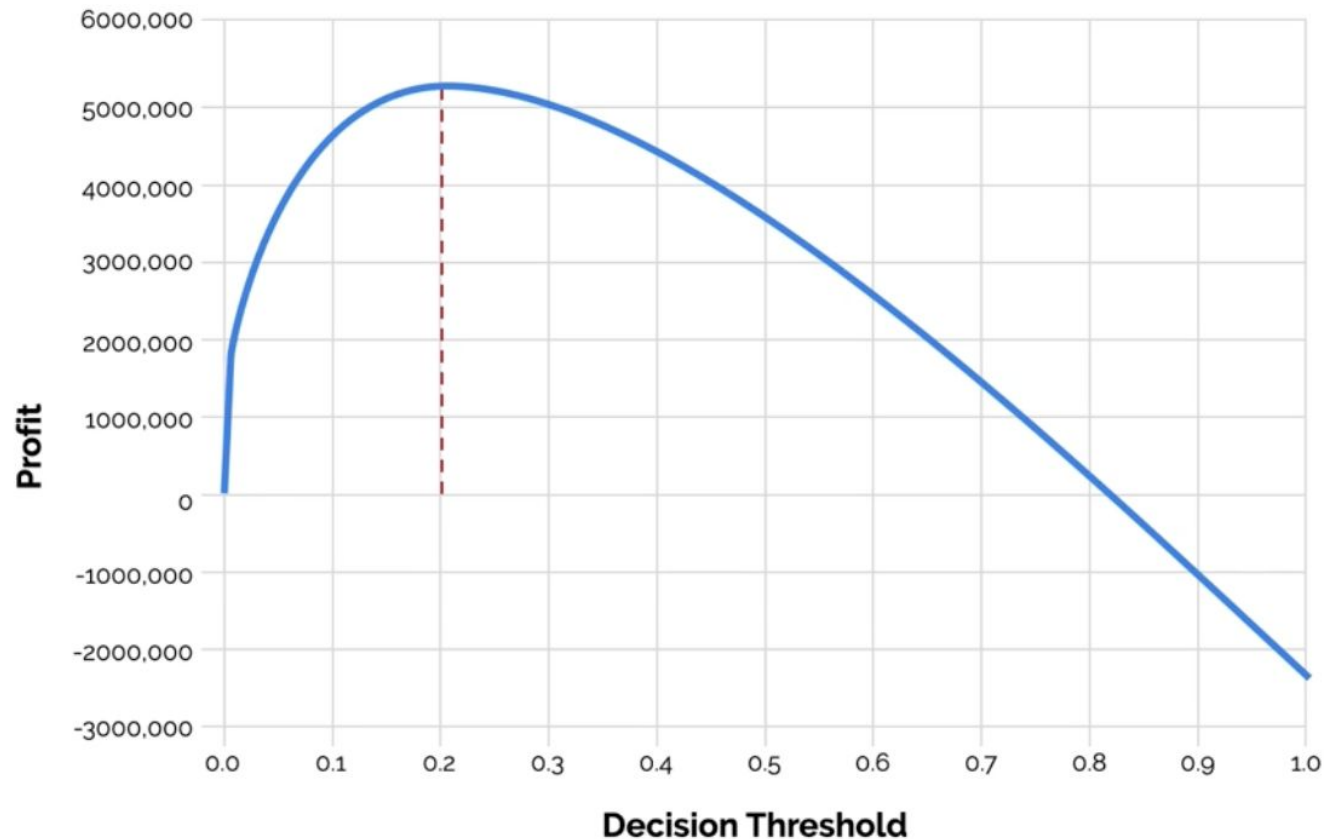
# Traducir los resultados en impacto financiero

		Churned	
		Yes	No
Discount	True	150,000	80,000
	False	20,000	3,000,000

		Churned	
		Yes	No
Discount	True	\$200	-\$30
	False	\$0	\$0

$$\begin{aligned} & 150,000 * 200 \\ & - 80,000 * 30 \\ \hline & = \$27,600,000 \end{aligned}$$

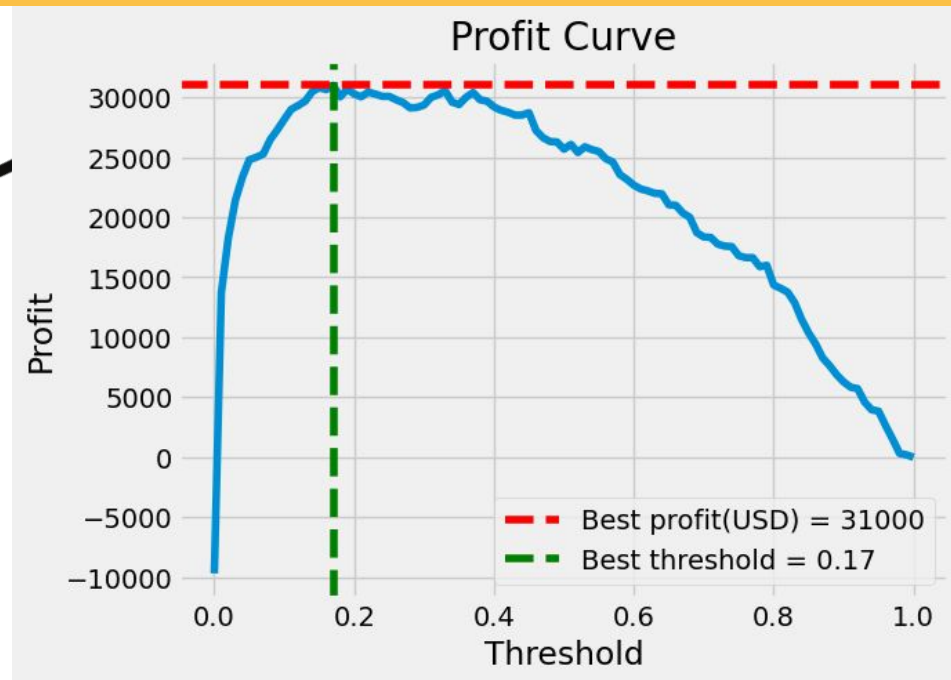
# Profit curve



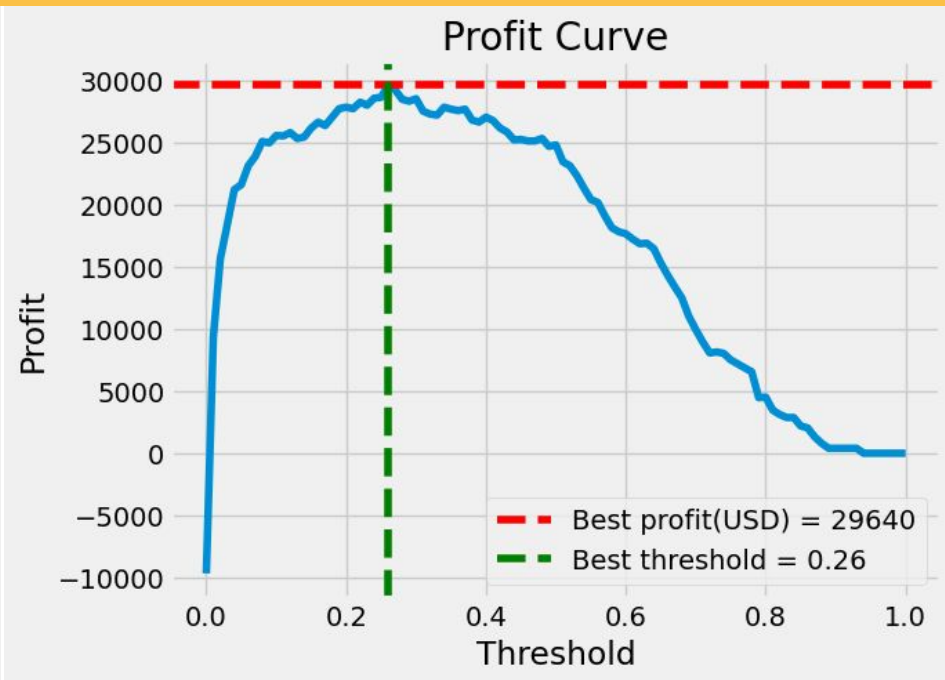
## Limitaciones:

- No saber los valores de los TP y FP
- Solo problemas de clasificación
- Valores heterogéneos

# La importancia de una variables



Todas las variables



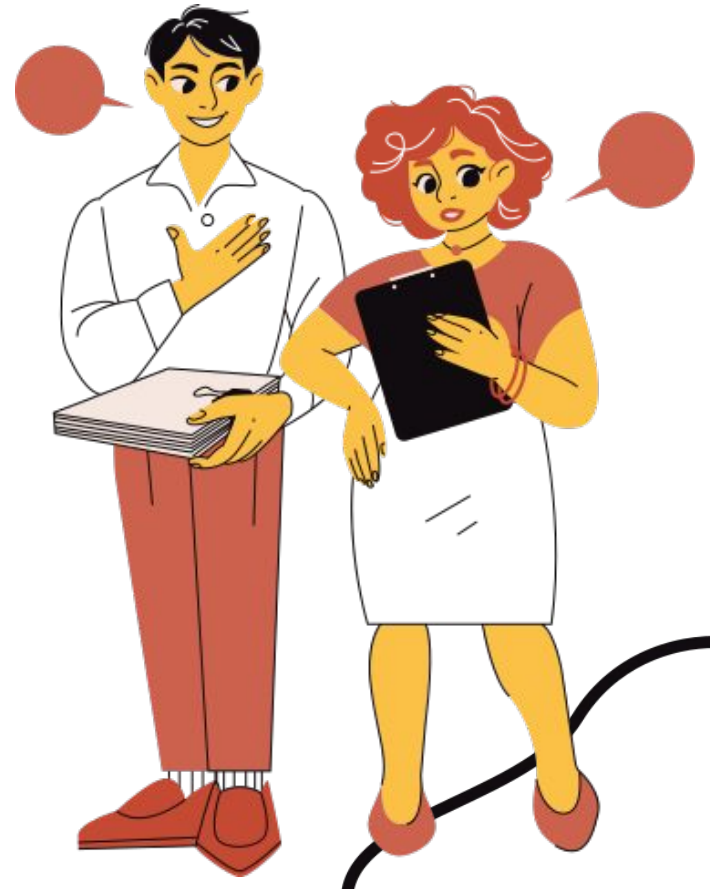
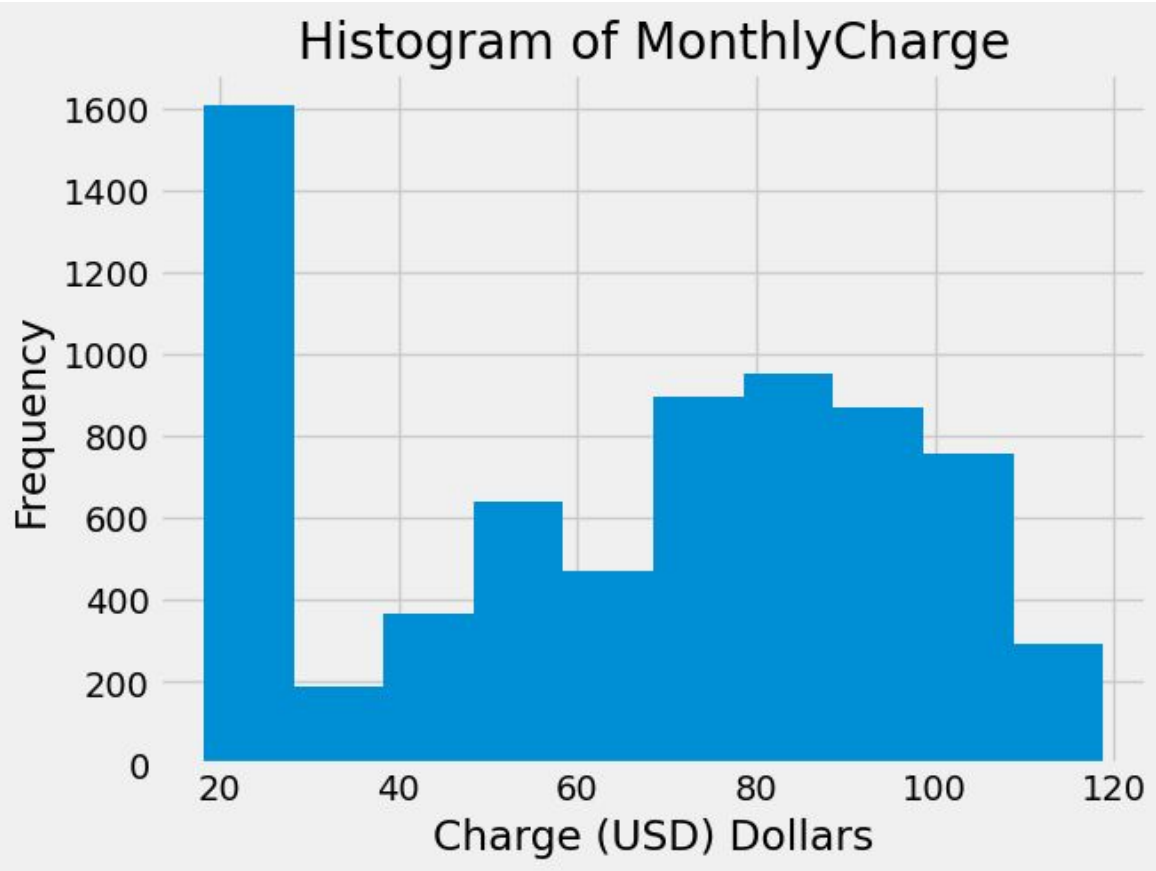
Falta 1 variable

**Total pérdida: 1.360 USD**

# Más allá de los umbrales de decisión



# Diferentes grupos de clientes



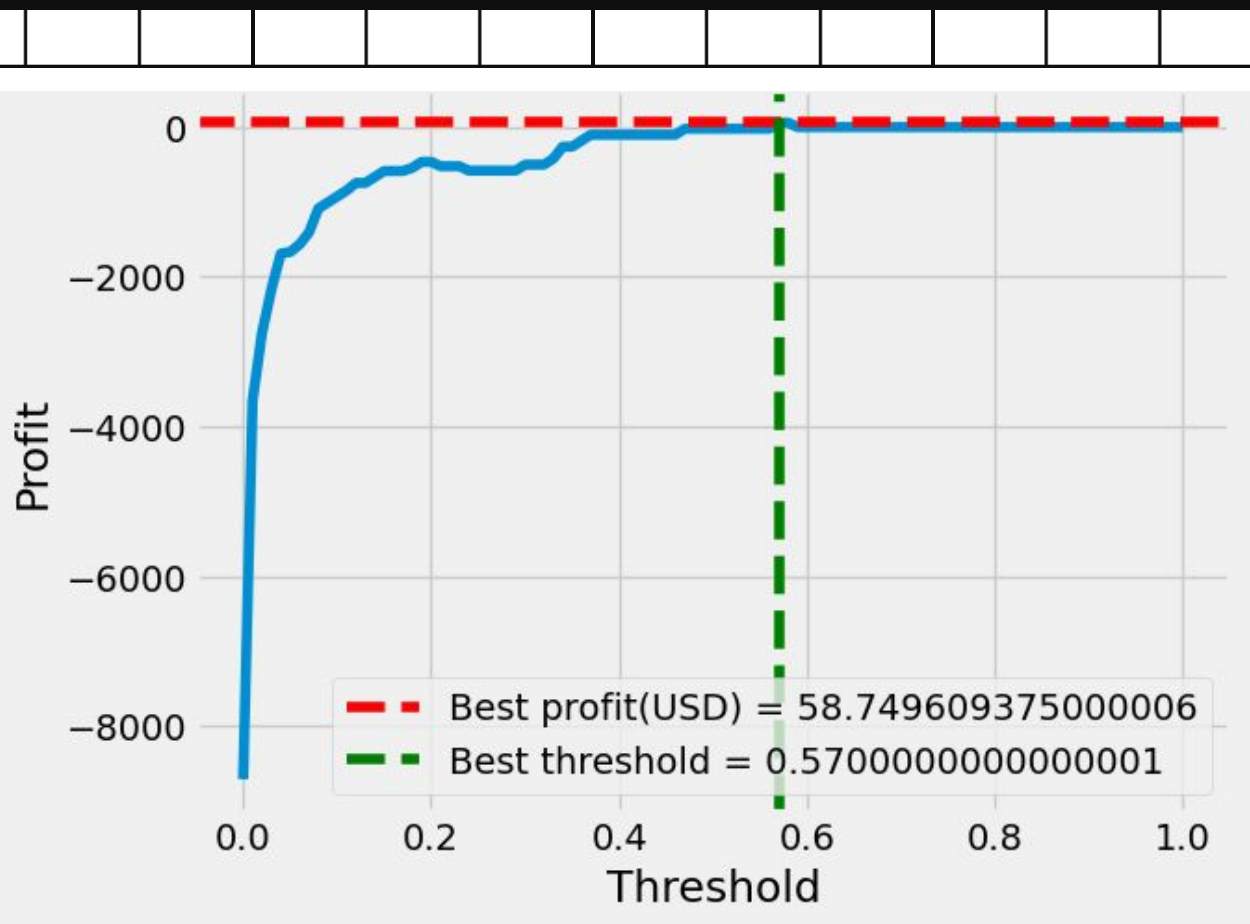
# Profit matrix

		Churned	
		Yes	No
Discount	True	$\$3 * \text{mean\_monthly\_charge}$	-\$80
	False	\$0	\$0

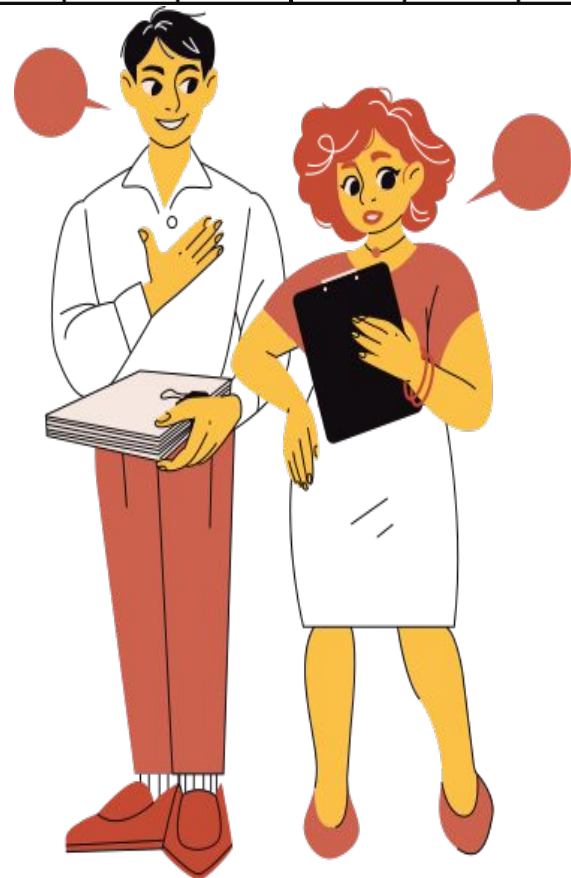
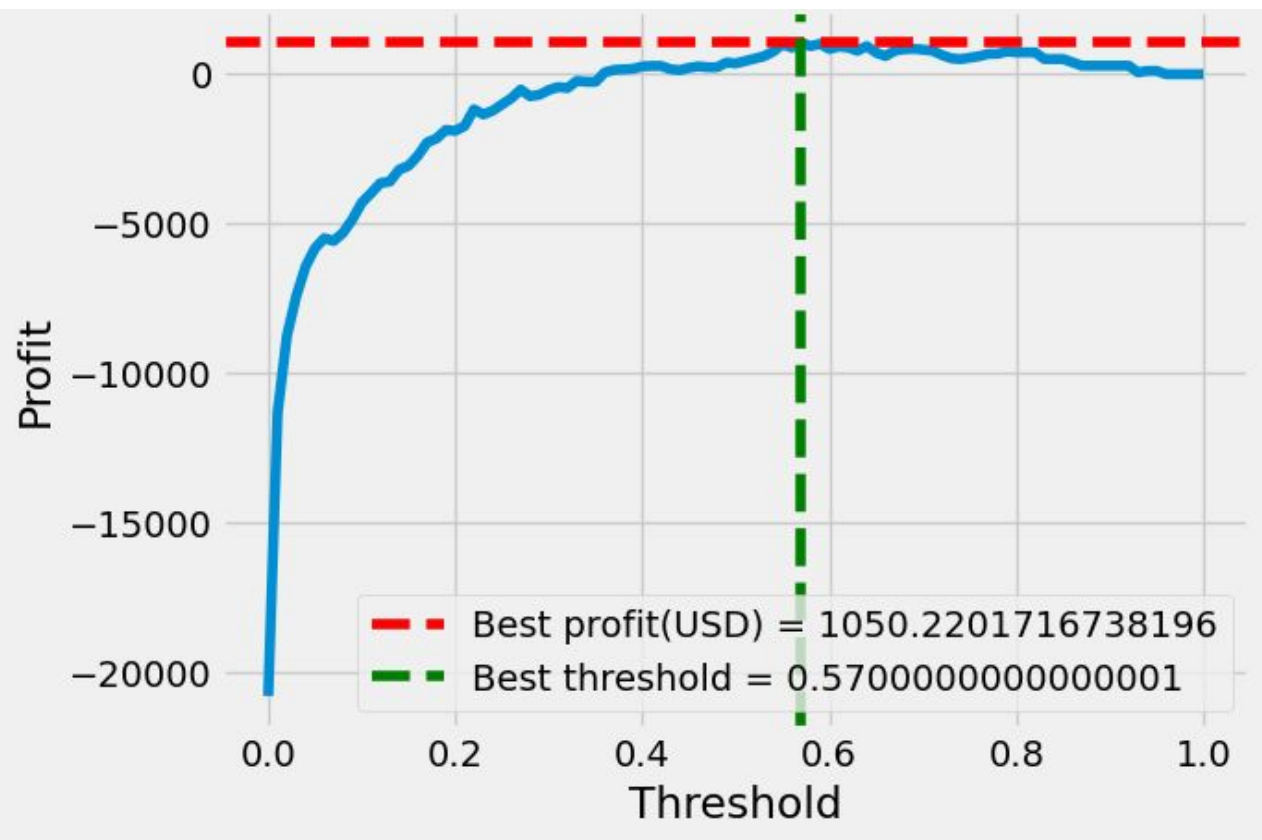




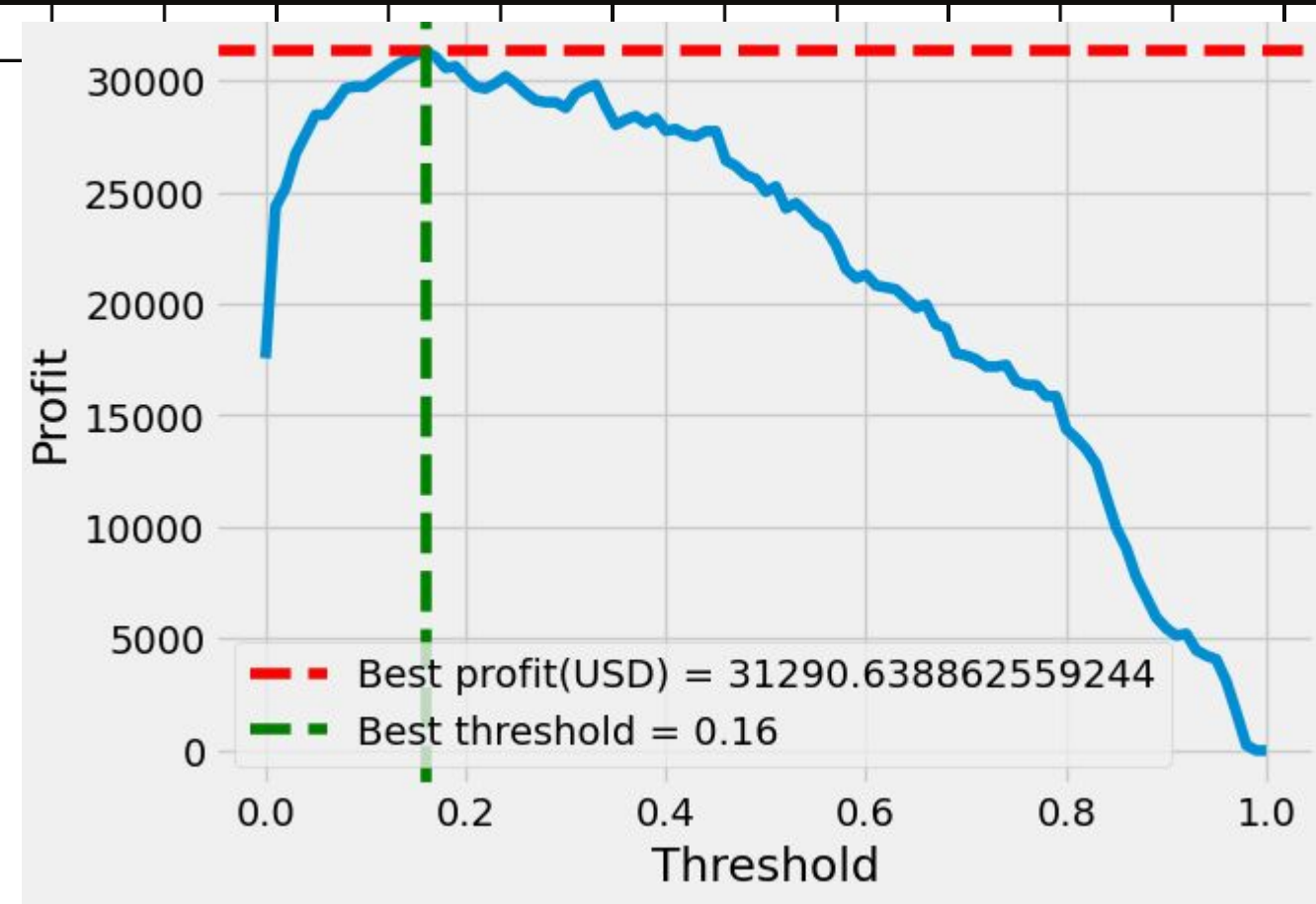
# Umbra para clientes que pagam menos de \$20 al mes



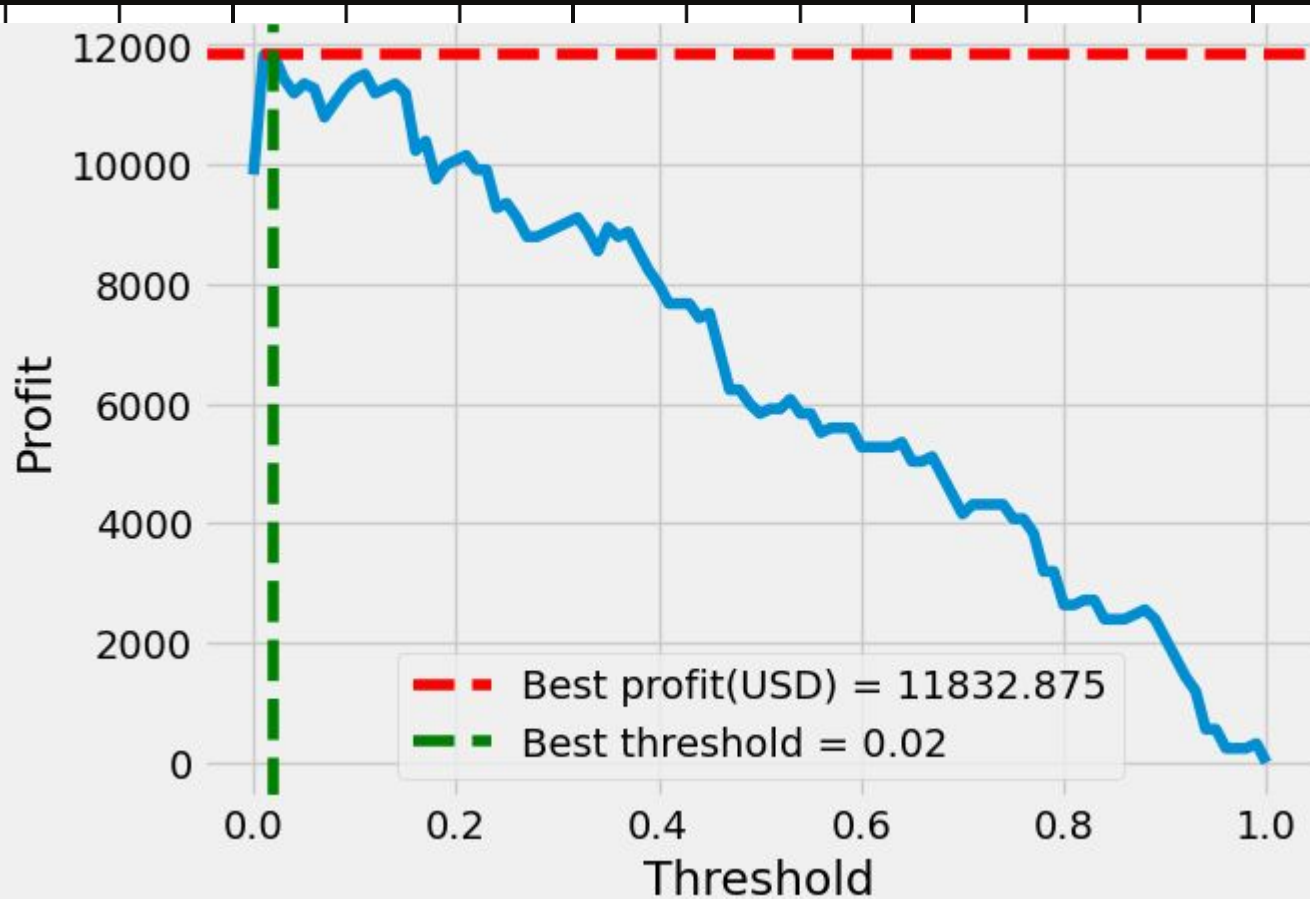
# Umbral para clientes que pagan entre \$20 y \$60/mes



# Umbral para clientes que pagan entre \$60 y \$100/mes

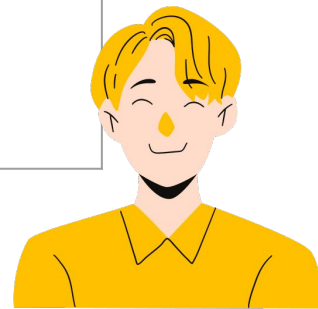


# Umbral para clientes que pagan más de \$100/mes



# Tipos de umbral

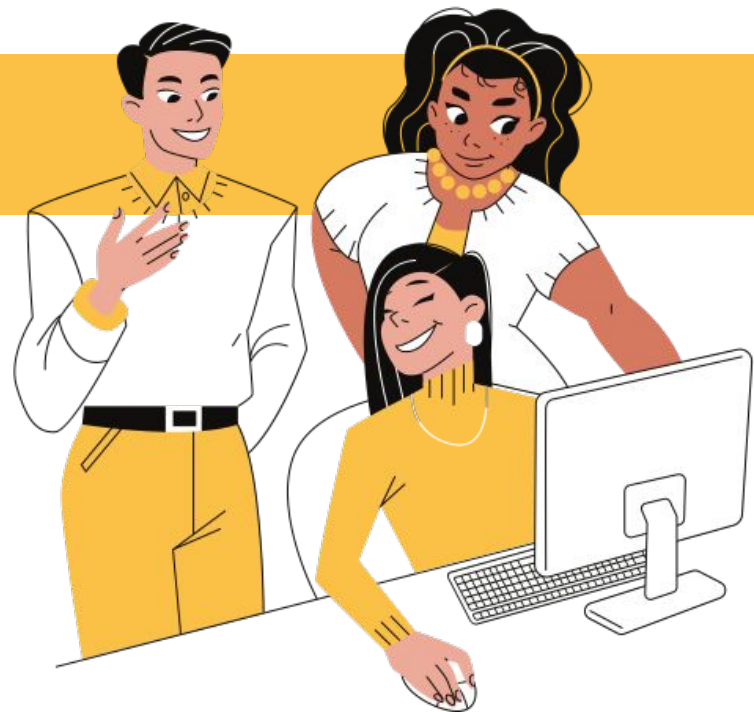
Tipo de Umbral	Ganancia esperada
Umbrales flexibles	44.232 USD
Umbral estático	31.000 USD
Diferencia	13.232 USD





# Mejoras del modelo

¿Debo seguir mejorando este modelo o pasar a un nuevo proyecto?



# “Datos esperados” que sean precisos.

```
errors = y_pred - y_true
```



# “Datos esperados” que sean precisos.

```
synth_data = y_pred - errors * improvement_factor
```





# “Datos esperados” que sean precisos.

```
synth_data_AUC = roc_auc_score(y_true, synth_data)
```



```
1 def make_synth_accurate_data(y_pred, y_true, improvement_factor=0.1):
2     """Create synthetic data that is accurate.
3
4     Args:
5         y_pred (array-like): Predicted probabilities.
6         y_true (array-like): True labels.
7         improvement_factor (float): A measure of how much to increase the accuracy of the synthetic data.
8
9     Returns:
10         array: Synthetic data.
11     """
12     errors = y_pred - y_true
13     synth_data = y_pred - errors * improvement_factor
14     synth_data_AUC = roc_auc_score(y_true, synth_data)
15     print(
16         f"improvement_factor: {improvement_factor:.2f}.    AUC of synthetic data: {synth_data_AUC:.2f}"
17     )
18     return synth_data
```

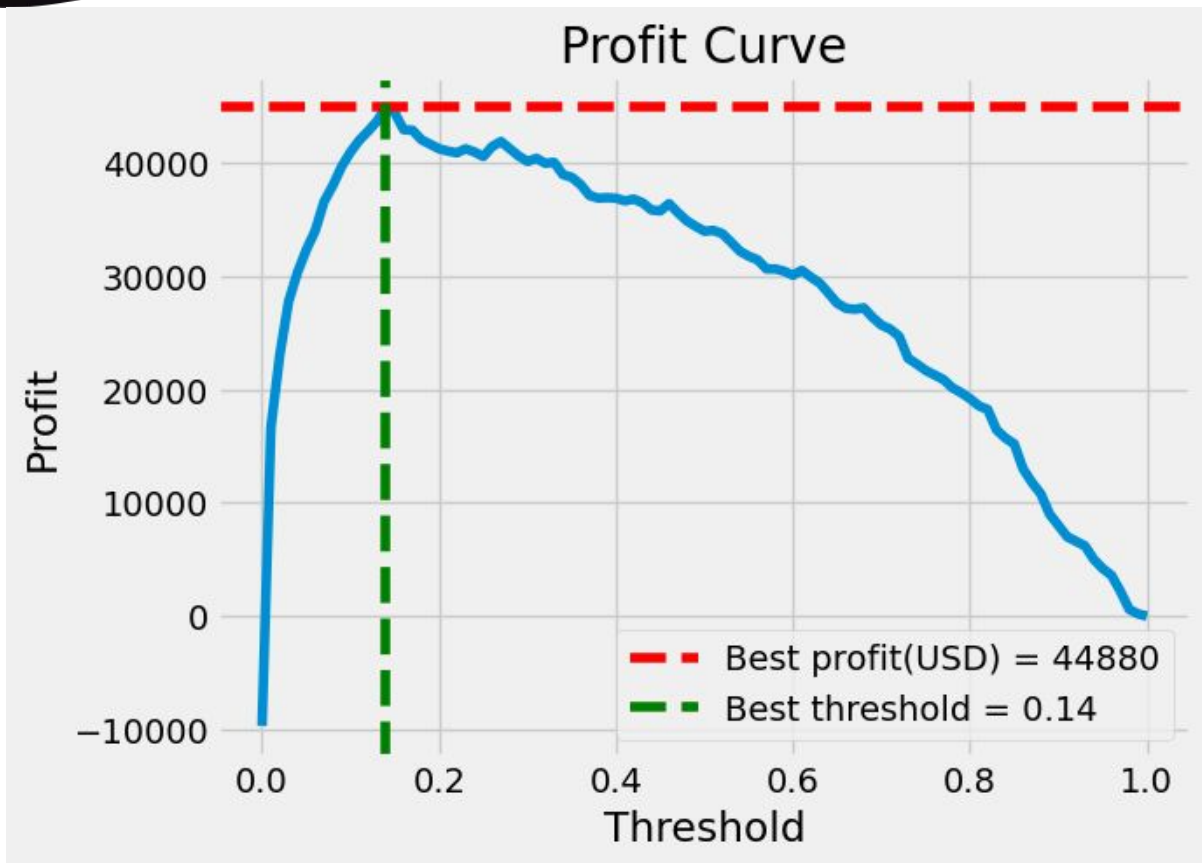
# Diferentes valores de Improvement\_factor

```
for i in np.linspace(0, 0.2, 11):  
    make_syntn_accurate_data(y_pred, y_test, improvement_factor=i)
```

improvement_factor: 0.00.	AUC of synthetic data: 0.81
improvement_factor: 0.02.	AUC of synthetic data: 0.83
improvement_factor: 0.04.	AUC of synthetic data: 0.84
improvement_factor: 0.06.	AUC of synthetic data: 0.85
improvement_factor: 0.08.	AUC of synthetic data: 0.86
improvement_factor: 0.10.	AUC of synthetic data: 0.87
improvement_factor: 0.12.	AUC of synthetic data: 0.88
improvement_factor: 0.14.	AUC of synthetic data: 0.89
improvement_factor: 0.16.	AUC of synthetic data: 0.90
improvement_factor: 0.18.	AUC of synthetic data: 0.91
improvement_factor: 0.20.	AUC of synthetic data: 0.92

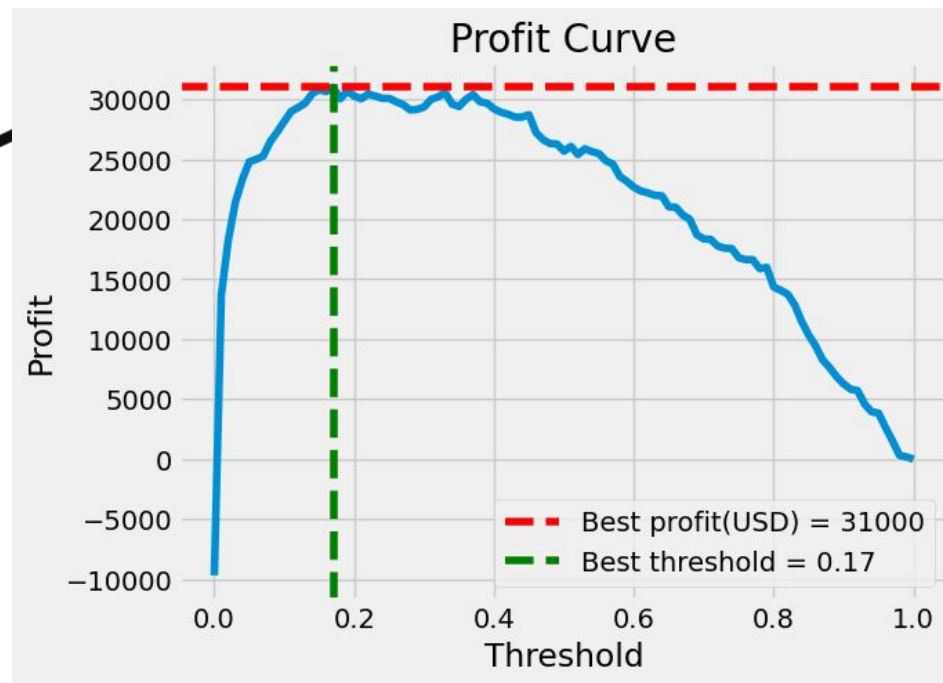


# Profit curve esperada con la mejora

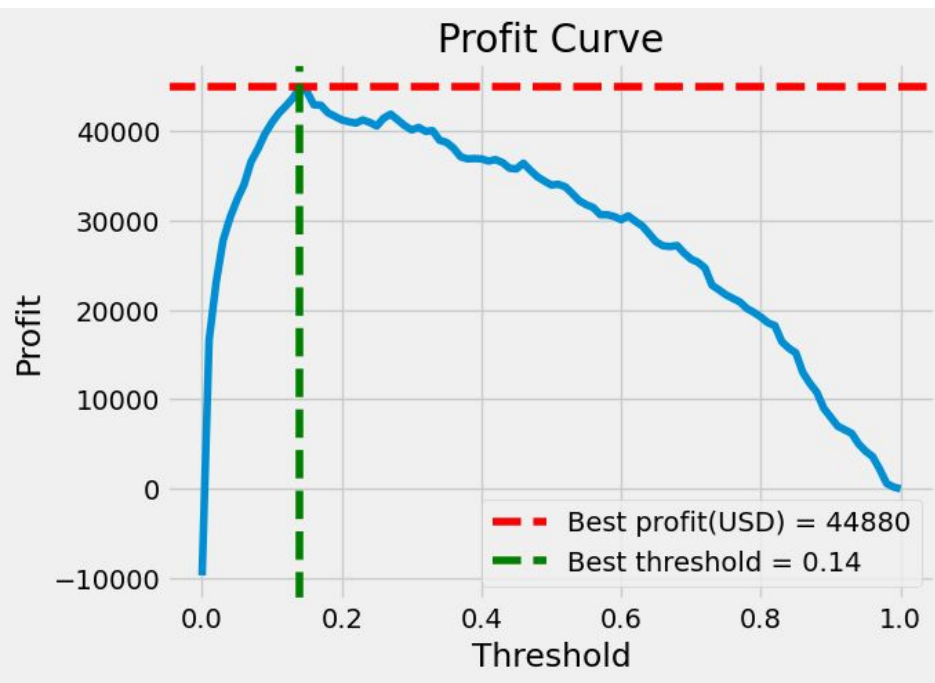


- Improvement factor: 0.14.
- AUC of synthetic data: 0.89

# Profit curve Actual vs Esperado




Actual



Esperado

Total ganancia: **13.880 USD**

# Referencias

- 
1. Dataset usado: [Telco Customer Churn](#)
  2. [Machine Learning for Business Decision Optimization W&B AI Academy](#)
  3. [Decision.AI](#)




# Muchas Gracias!!!

¿Alguna pregunta?

