



SIX Payment Transaction Volumes Forecasting

DISCLAIMER_FROM_SIX

**Data has been
anonymised**

**No conclusion needs
to be derived out of it**

SIX_PACK_TEAM



Alžbeta Bohiniková = {
 'Bg': 'Applied Mathematics',
 'Past job': 'Researcher',
 'Loves': 'Baking gingerbread'
}



Luis Miguel Rodríguez Sedano = {
 'Bg': 'Environmental Consultant',
 'Past job': 'Forestry Engineer',
 'Loves': 'Hiking on rainy days'
}



Mukund Pondkule = {
 'Bg': 'Computational Engineering',
 'Past job': 'HPC Support',
 'Loves': 'Wandering aimlessly'
}



Michael Flury = {
 'Bg': 'Business & Finance',
 'Past job': 'Project Manager',
 'Loves': 'Tennis drop shots'
}

BUSINESS_UNITS

Stock Exchanges

Switzerland & Spain

Securities services

Financial instruments

Banking services

Payment processing

Financial information

Data provider



```
SIX_Banking_Services = ['ATM transactions',  
                        'Debit card transactions',  
                        'Point-of-sale transactions',  
                        'Interbank payments']
```

Why is forecasting transaction volumes important?

TRANSACTION FEE REVENUE MODEL

Fee earned for each payment transaction processed



DIRECT TRANSACTIONS FEE

i.e. contactless card payment in a shop



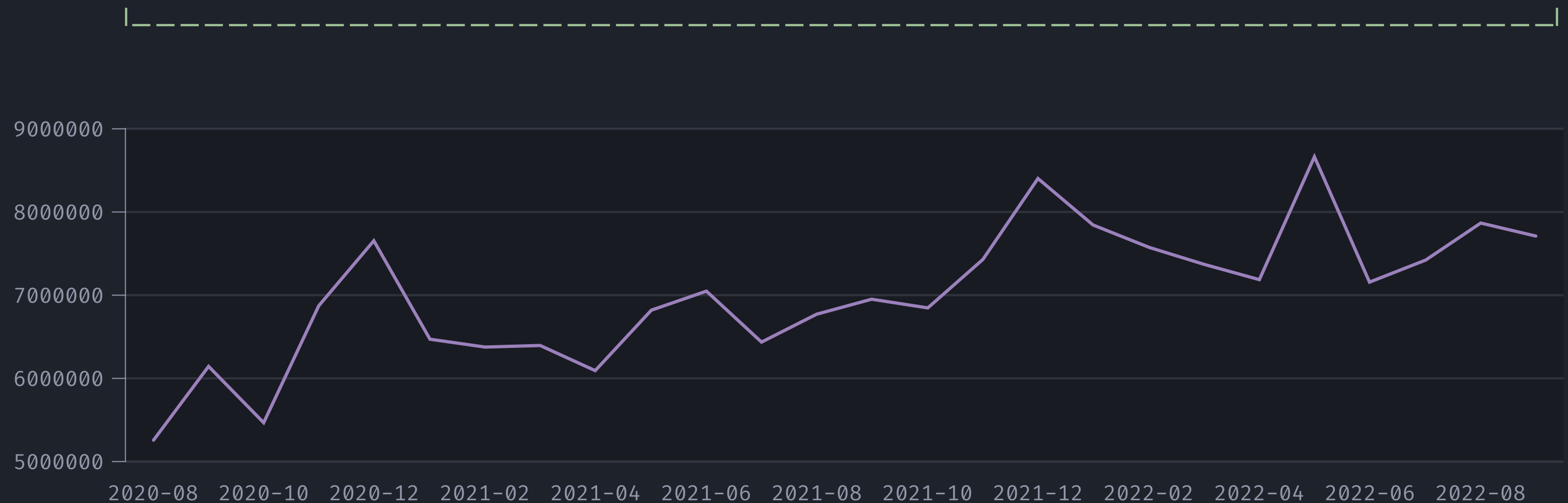
ADDITIONAL SERVICES (ADD-ONS)

i.e. 3-D secure and fraud check

What is time series data?

MEASUREMENTS OVER TIME

Month	08-2020	09-2020	10-2020	11-2020	12-2020	...	09-2022
Number of transactions	5'256'987	6'143'986	5'467'321	6'872'328	7'654'293	...	7'784'154



Time series decomposition

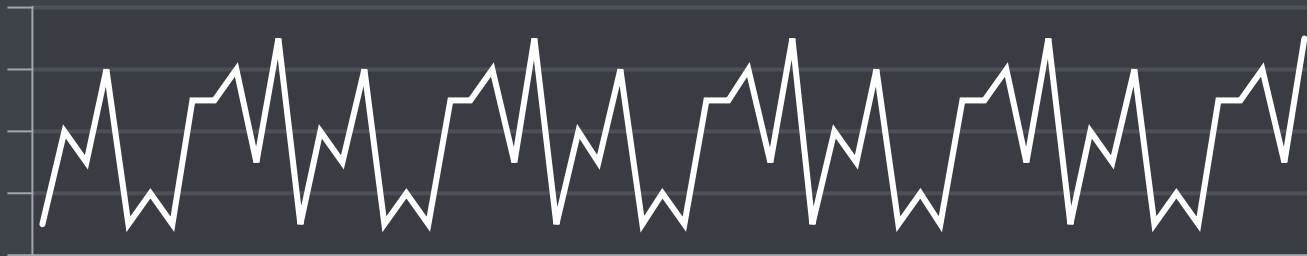
BREAKDOWN INTO SIMPLE SIGNALS

TREND



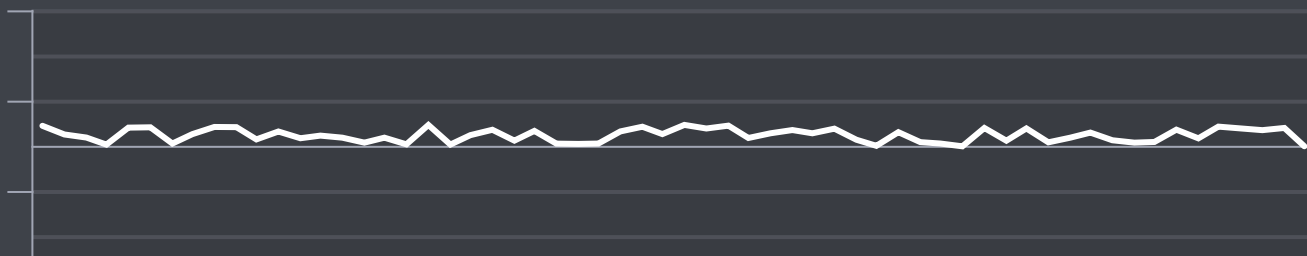
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SEASONALITY



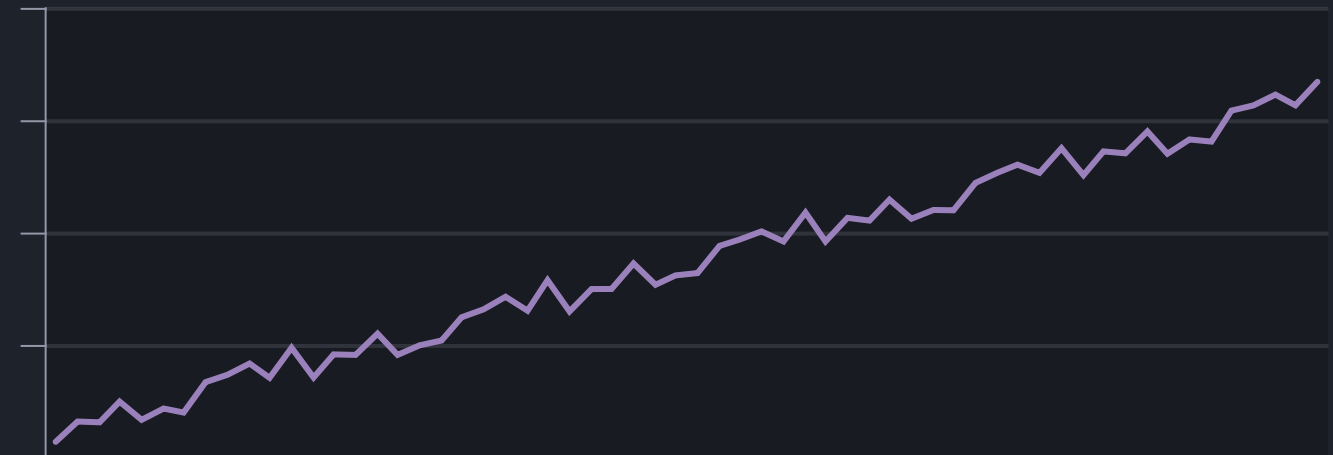
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RANDOM EFFECT



=

ACTUAL DATA (SUM OF SIGNALS)



Time series decomposition

BREAKDOWN INTO SIMPLE SIGNALS

TREND



+

SEASONALITY



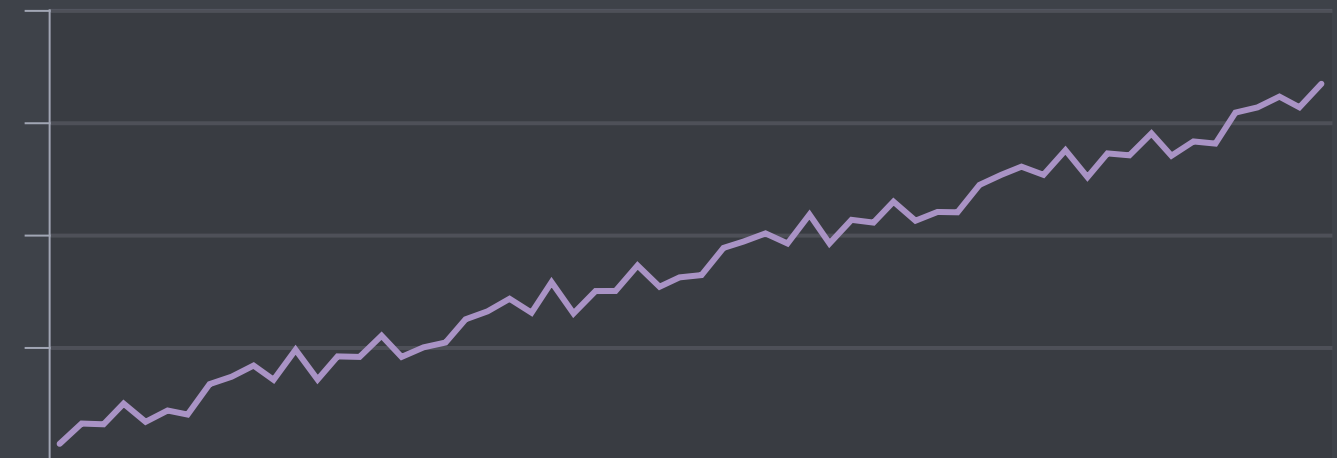
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RANDOM EFFECT



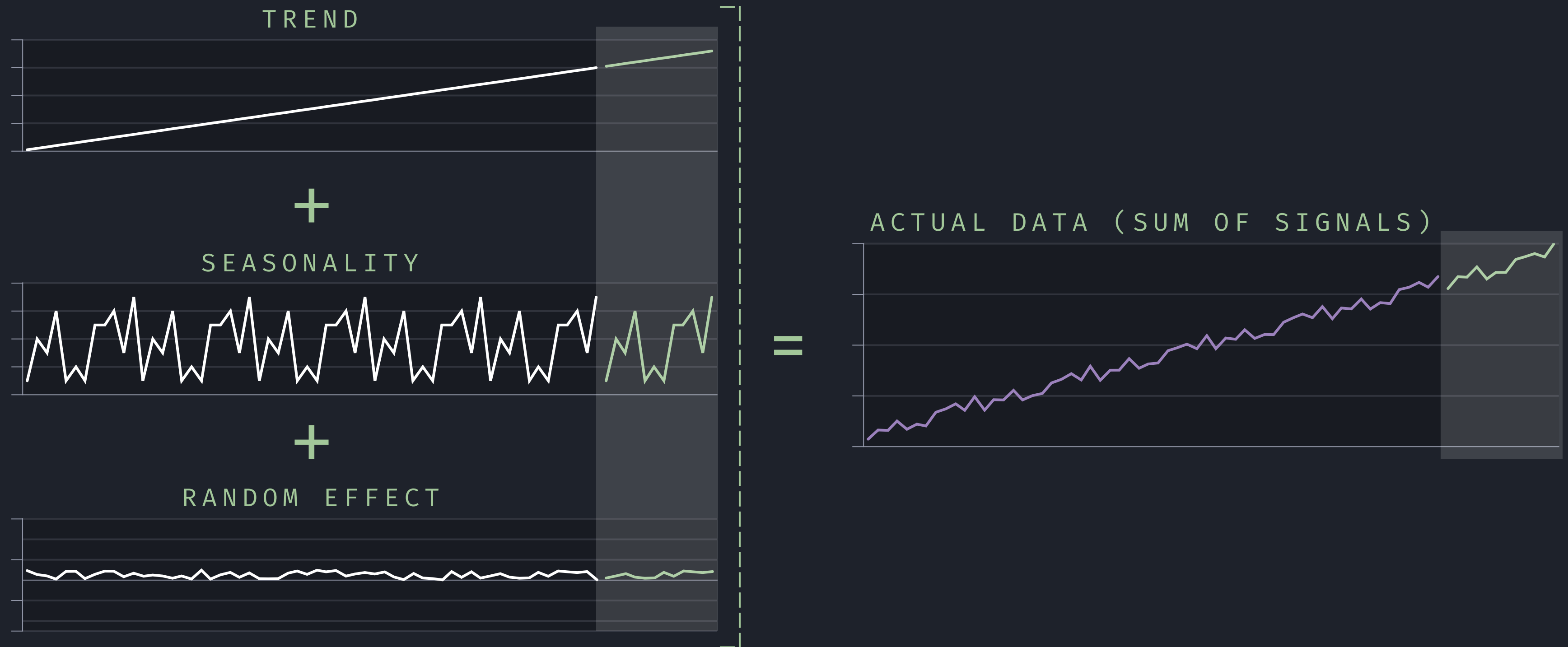
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ACTUAL DATA (SUM OF SIGNALS)



How does understanding simple signals help us?

FORECAST EACH SIGNAL INDIVIDUALLY AND SUM THEM TO PREDICT DATA



Time series models to estimate weights

- Special type of regression models
- For each "simple" signal, weights ("importance") are assigned to past values to predict the future
- Each model assigns weights in its own way

S(ARIMA): SEASONAL AUTOREGRESSIVE INTEGRATED MOVING AVERAGE

- Uses linear weights

ETS: EXPONENTIAL TRIPLE SMOOTHING

- Uses exponential weights

PROPHET

- Uses machine learning techniques to determine weights

What makes a good model?

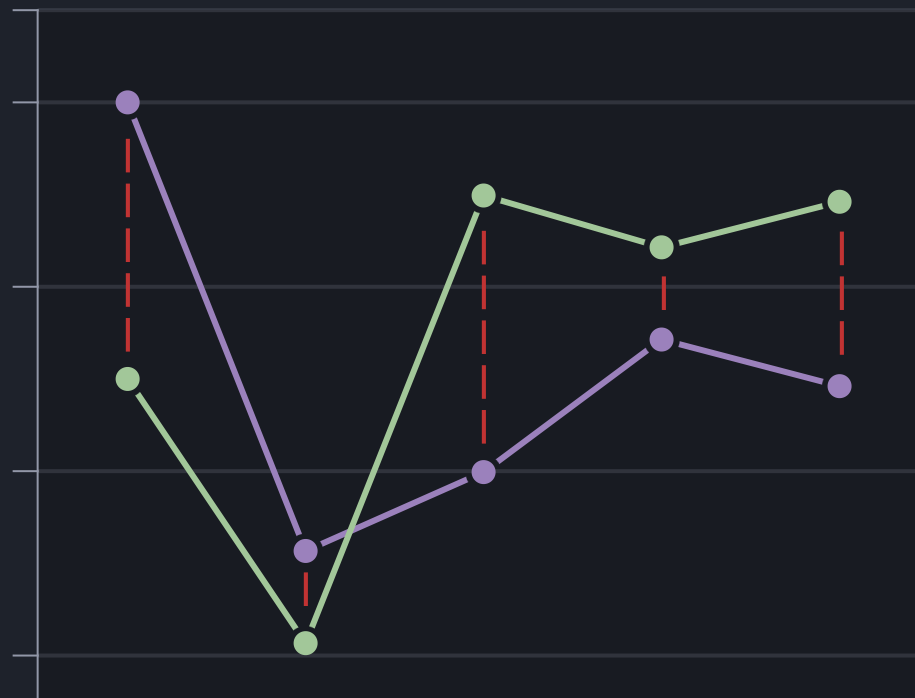
- Predicted data should try to predict actual data as close as possible

● Actual
● Predicted
| Difference

The **Error** calculates average % difference of the predictions from the real data

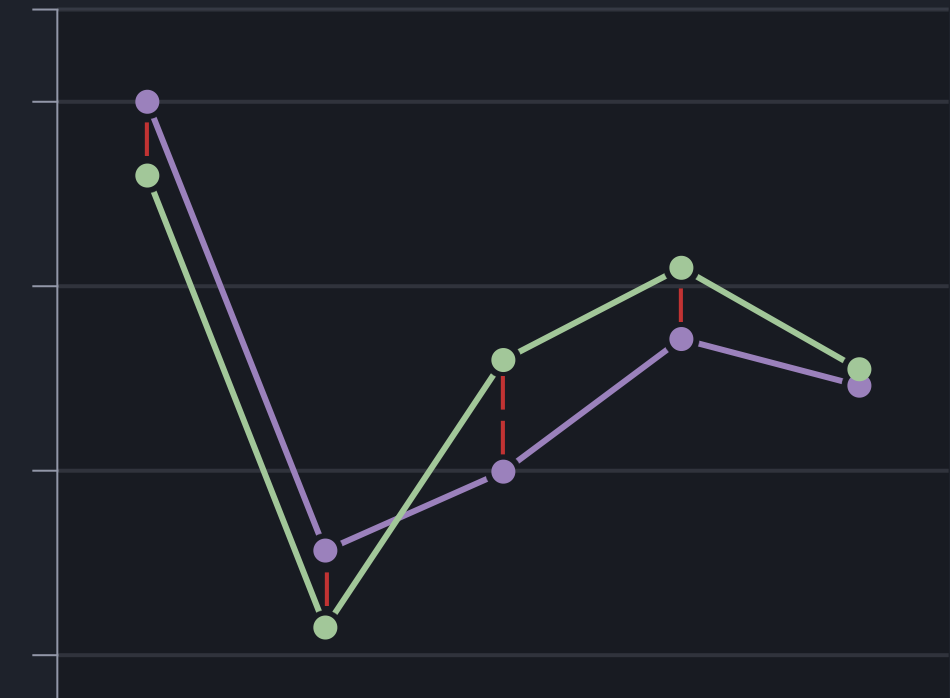
Example 1

5%



Example 2

1%



Better = lower error

PIPELINE_APPROACH

STEP_1

Transactions for
1913 merchants (26 months)

Merchant name	08-2020	09-2020	...	09-2022
Merchant 1	0	11	...	12
Merchant 2	915'976	1'002'153	...	978'198
...
Merchant 1913	57'282	70'174	...	62'109



STEP_2

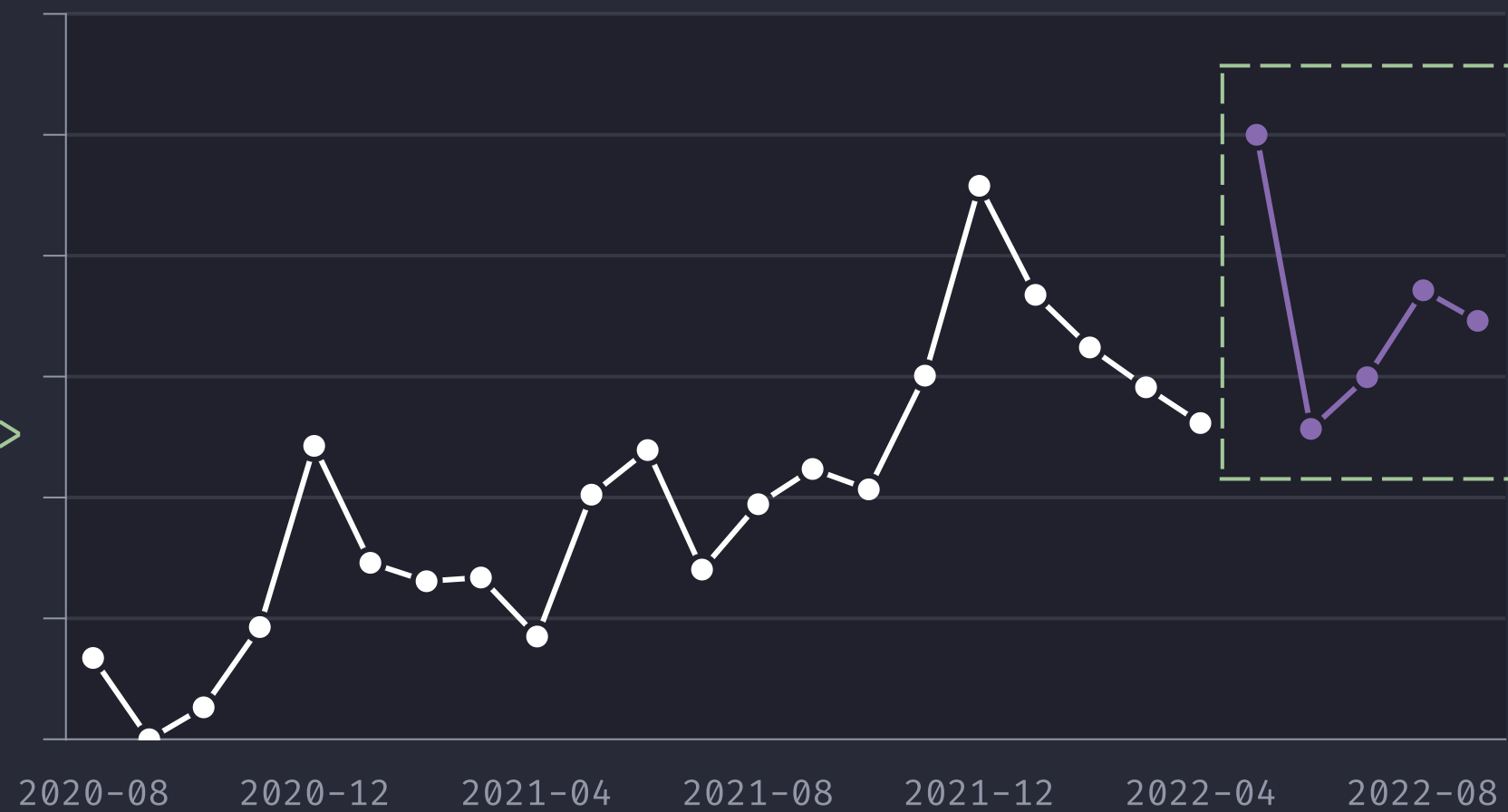
Sum of all transactions
per month (26 months)

Merchant name	08-2020	09-2020	...	09-2022
SUM	5'256'987	6'143'986	...	7'784'154

PIPELINE_APPROACH_NEXT

STEP_3

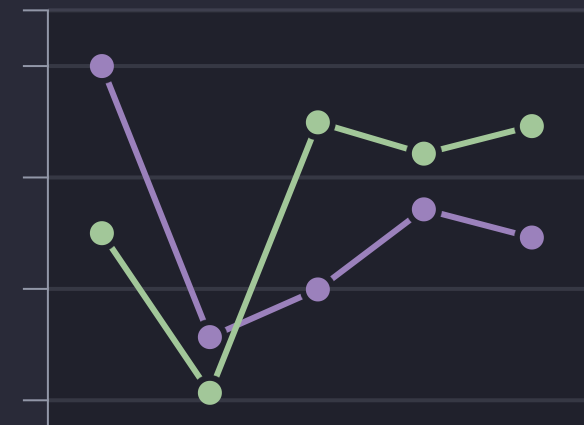
Model selection based on 21 months ●
Evaluation based on last 5 months ●



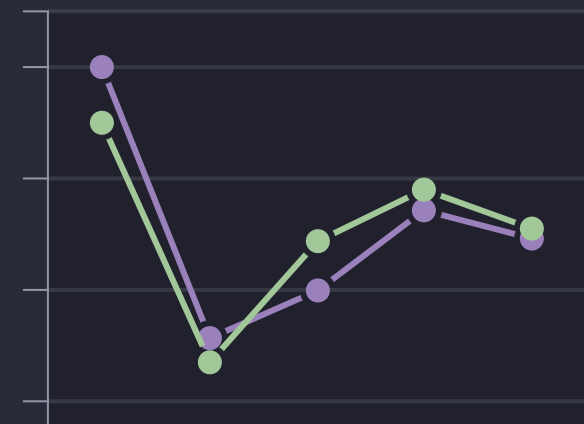
Which model would
you choose?

● Actual
● Predicted

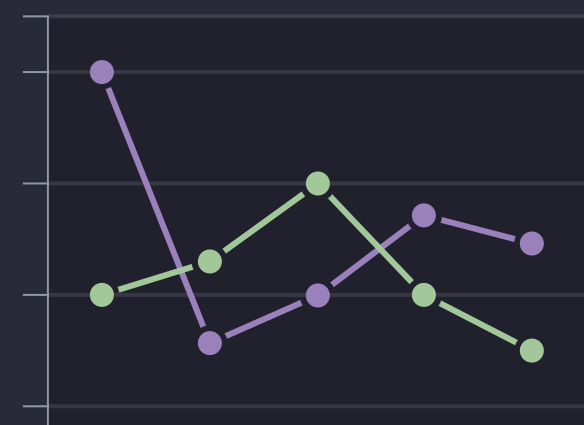
ARIMA



ETS



PROPHET

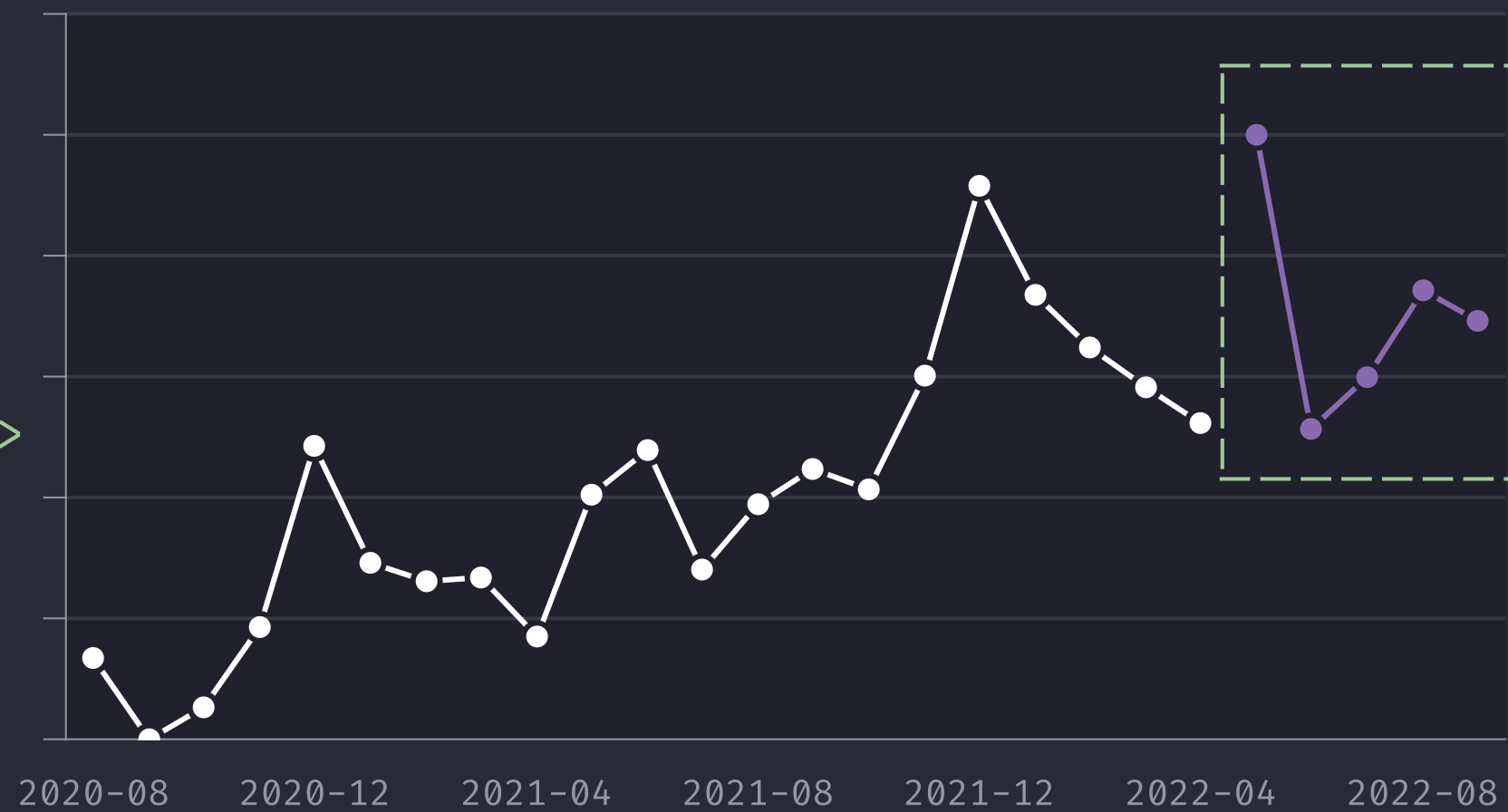


2022-05 2022-07 2022-09

PIPELINE_APPROACH_NEXT

STEP_3

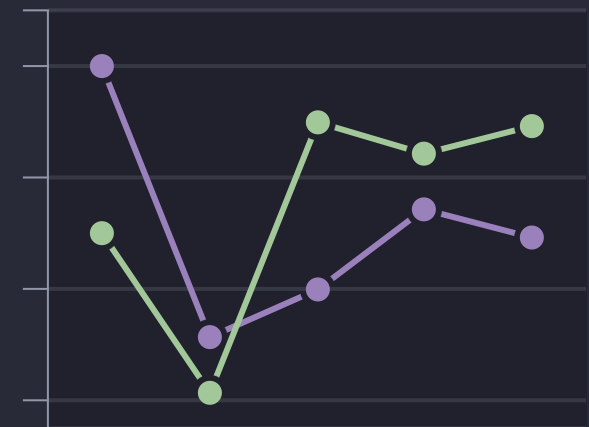
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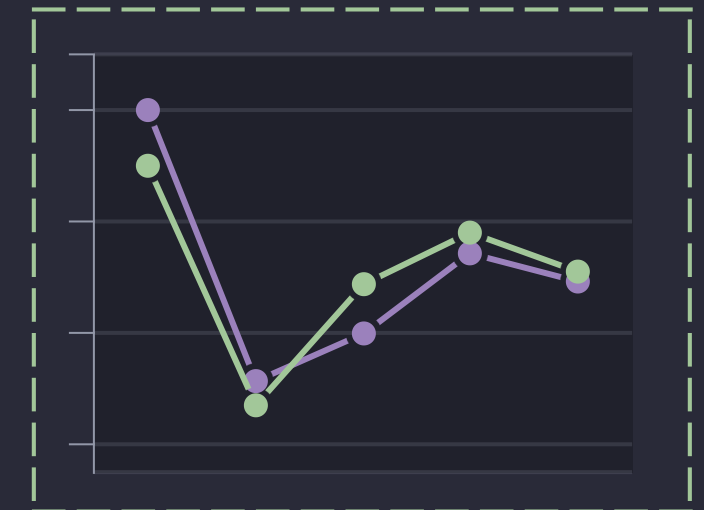
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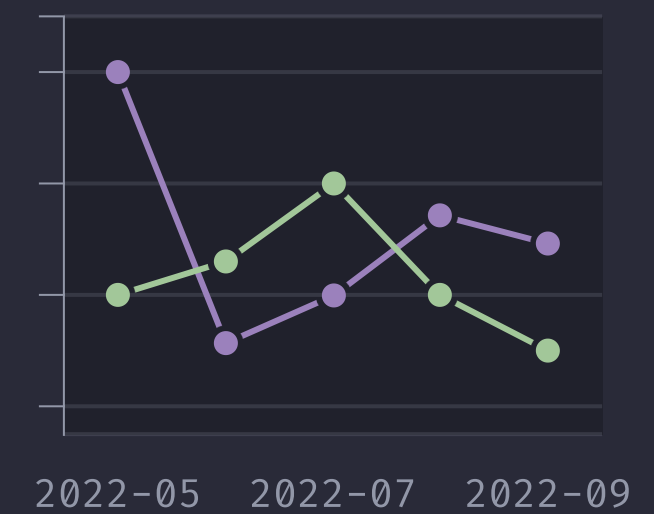
ARIMA



ETS



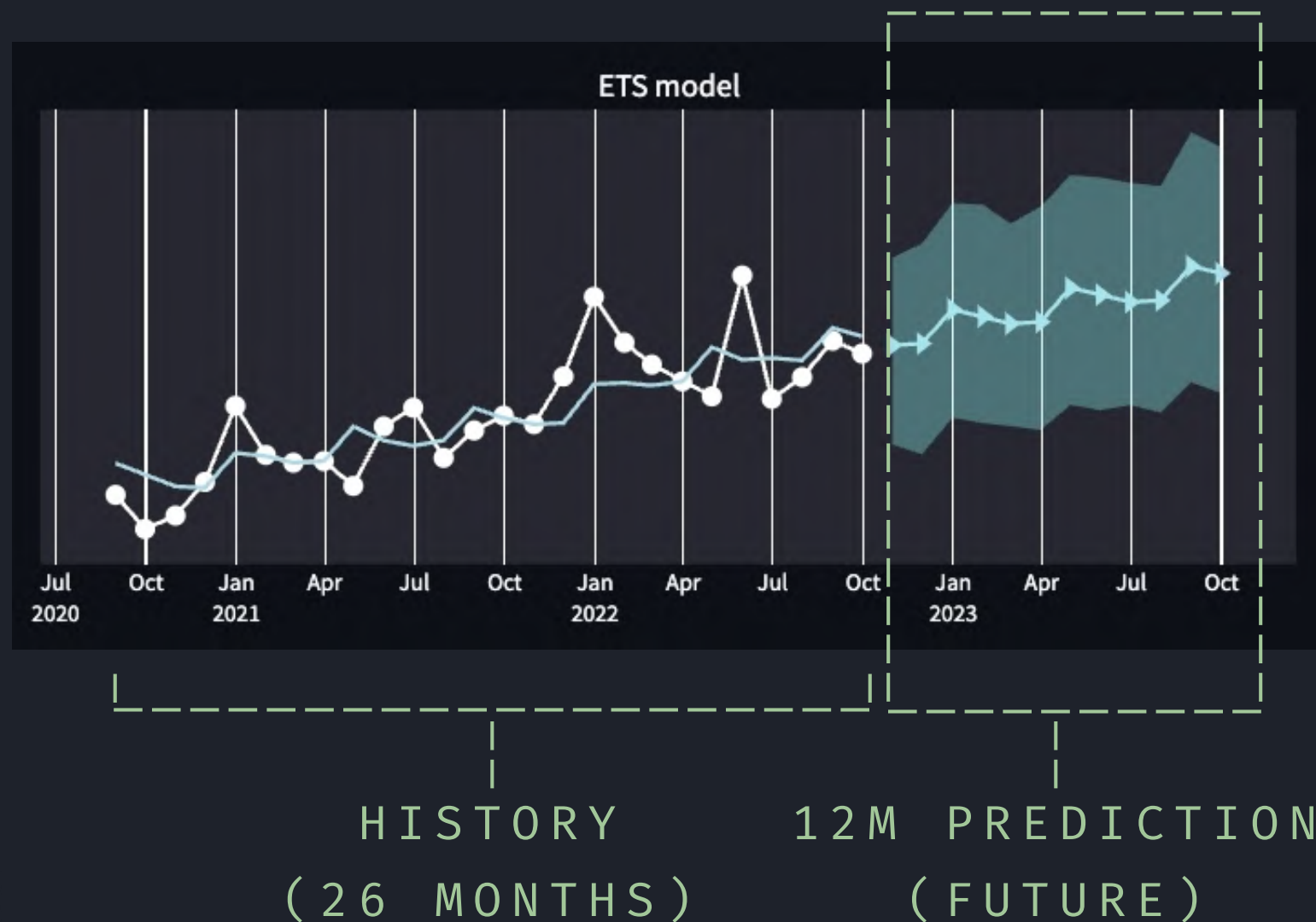
PROPHET



PIPELINE_APPROACH_END

STEP_4

Using the best model selected in step 3
we predict the next 12 months



STEP_5

Automatic monitoring of model
errors given new data

Model	10-2022	11-2022	12-2022	...
2022_09_best	0.75%	1.2%	2.4%	...
2022_10_best		1%	3.5%	...
2022_11_best			2.7%	...
...				...

ERROR = % DIFFERENCE
(PREDICTED VS ACTUAL FOR NEW MONTHS)

<TITLE> PIPELINE IN ACTION </TITLE>

^(^0^) / \ (^ _ ^)

Let's see our work in action

{ \ _ _ / }
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[HTTPS://MIKJF-SIX-STREAMLIT-APP-SRCAPP-XHHC83.STREAMLIT.APP/](https://mikjf-six-streamlit-app-srcapp-xhhc83.streamlit.app/)

Project conclusions

FORECASTING TRANSACTION VOLUMES => FORECASTING REVENUE

- Thanks to our pipeline SIX will be able to forecast transaction volumes more precisely

IDENTIFY MERCHANTS WITH HIGH VOLUME OF TRANSACTIONS

- SIX will be able to plan tailored solutions

<TITLE> ALMOST_THE_END </TITLE>

^ (^ 0 ^) / \ (^ _ ^)

**Thank you very much
for your time and consideration**

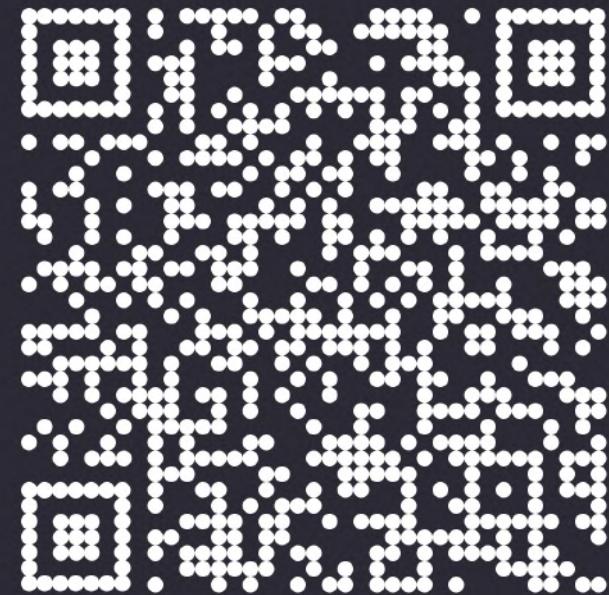
SIX_PACK TEAM



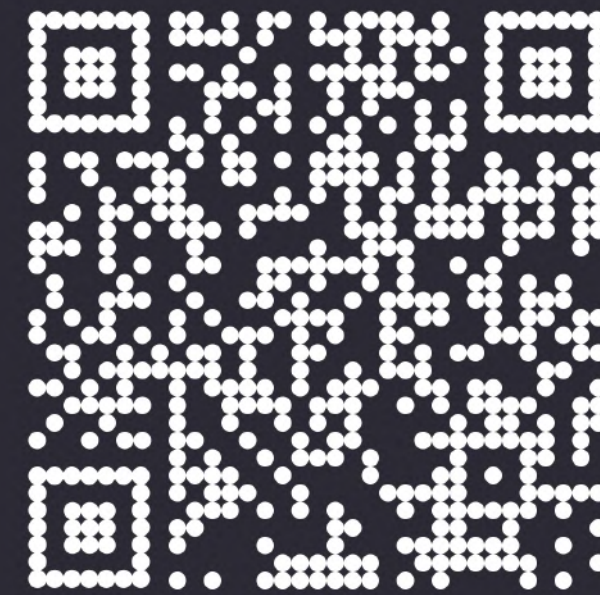
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