

Fintech Group Project 01

Azuki, BAYC and Crypto Punks NFTs.

by Github users /[@dockingbay24](#) /[@angel-estrada7](#) and /[@mmsaki](#), 2022.

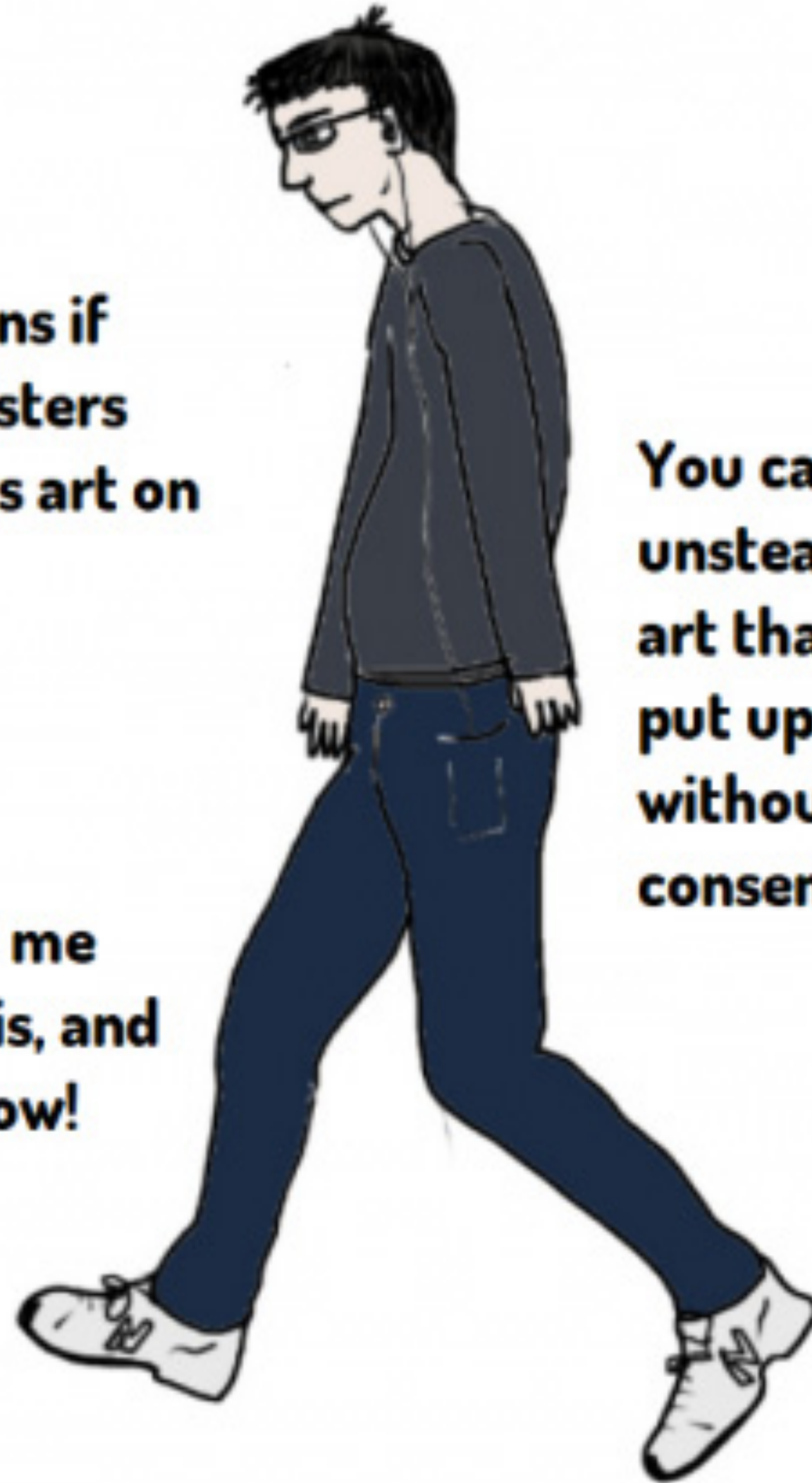
Table of Contents

- Hypothesis of Project
- Data Collection
- Data Cleanup & Exploration
- Data Analysis
- Discussion
- Postmortem
- Your Questions?



Azuki Token ID: 7558 | Current Owner: ron1n.eth | Sale: 27ETH

everyone else



So what happens if
somebody registers
somebody elses art on
Deloitte?

So what stops me
from doing this, and
saying mine now!

You cannot
unsteal that
art that was
put up
without
consent.

The NFT Alpha



Monetizes memes

Money goes
brrrrrrrrrrrrrr

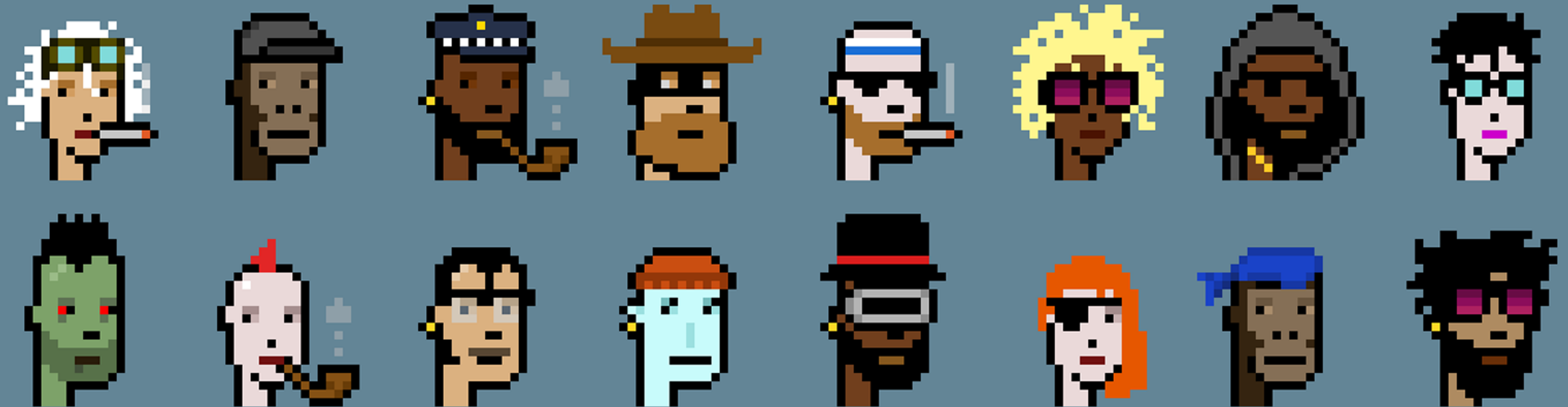
Creates something
for the community

1. Hypothesis

Hypothesis of Project

Our motivation and summary...

- Should you invest in [Azuki](#), [BAYC](#) or [Crypto Punks](#)?
- How can you tell which collection is performing well?



Data Collection

Describing what kinds of data we needed and where to find it.

2. Data Collection

Collecting data for NFT Collections

- Covalent API
- Etherscan Python Dependacy

0.1 Import dependancies

```
[10]: # Import dependancies
import os
import requests
import pandas as pd
import json
from dotenv import load_dotenv
from etherscan_py import etherscan_py
import plotly.express as px
```

```
[2]: # Loading .env containing our keys
load_dotenv()
```

[2]: True

```
[3]: # create variable for api key
api_key = os.getenv('COVALENT_API_KEY')
type(api_key)
```

[3]: str

0.2 Current value of ETH

```
[4]: # import dependancy
from etherscan_py import etherscan_py
client = etherscan_py.Client(os.getenv('ETHERSCAN_API'))

# Print current eth price and latest block height
eth_value = client.get_eth_price()
eth_value
```

3. Data Cleanup & Exploration

Exploring our collection data through APIs preparing it for analysis

0.3 Set variables

```
[5]: # Append url for our api
url = "https://api.covalenthq.com/v1"
chain_id = "/1"
azuki_address = "/0xED5AF388653567Af2F388E6224dC7C4b3241C544"
cryptopunks_address = "/0xb47e3cd837dDF8e4c57F05d70Ab865de6e193BBB"
BAYC_address = "/0xBC4CA0EdA7647A8aB7C2061c2E118A18a936f13D"
date_option = '/?quote-currency=USD&format=JSON&from=2017-01-01&to=2022-05-01'
page_option = '/transactions_v2/?'
    ↳quote-currency=USD&format=JSON&block-signed-at-asc=false&no-logs=false&page-number=0&page-s
api_option = "&key=" + api_key
api_no_option = '/?key=' + api_key
```

0.4 1. Azuki Daily Volume

```
[6]: # Create variables needed for owner data and add to url
historical_url = url + chain_id + "/nft_market/collection" + azuki_address +
    ↳api_no_option

# Get request
azuki_historical_json = requests.get(historical_url).json()

# Convert historical json data to a dataframe and view data
azuki_df = pd.DataFrame(azuki_historical_json['data']['items'])

# Set index to date
azuki_df = azuki_df.set_index('opening_date')

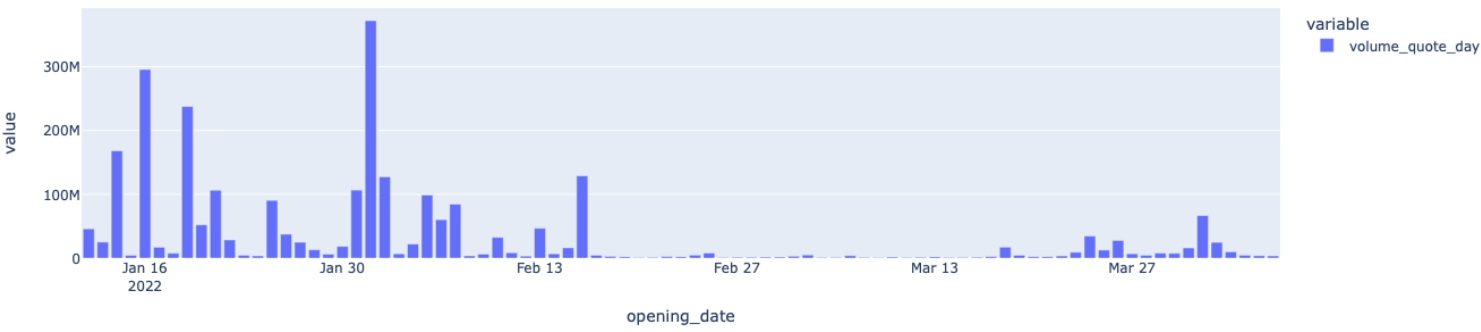
# Create Volume dataframe
azuki_vol_df = pd.DataFrame(azuki_df, columns = ['volume_quote_day',
    ↳'unique_token_ids_sold_count_day']).sort_index()
azuki_vol_df.head()
```

[6]:

	volume_quote_day	unique_token_ids_sold_count_day
opening_date		
2022-01-12	45941404.0	2402
2022-01-13	25129178.0	1318
2022-01-14	168151840.0	470
2022-01-15	4408686.0	499
2022-01-16	295638336.0	368

```
[40]: # Plot Volume quote per day
azuki_volume = azuki_vol_df['volume_quote_day'].astype(int)

# Plot Historical daily volume
px.bar(azuki_volume)
```



0.5 1. Azuki Historical transactions

```
[8]: # Querying the API for transaction data
azuki_tx_url = url + chain_id + "/address" + azuki_address + page_option +
    ↳api_option
azuki_tx = requests.get(azuki_tx_url).json()

# Convert transactions data to dataframe
azuki_tx_df = pd.DataFrame(azuki_tx['data']['items'], columns =
    ↳['to_address_label', 'fees_paid', 'value_quote', 'block_signed_at']).
    ↳set_index('block_signed_at').sort_index()

azuki_tx_df.head()
```

[8]:

	to_address_label	fees_paid	value_quote
block_signed_at			
2022-05-02T17:19:43Z	LooksRare: Exchange	19172817560393340	84346.297729
2022-05-02T17:20:39Z	None	7005576883041388	0.000000
2022-05-02T17:23:13Z	None	16639408965196144	0.000000
2022-05-02T17:24:03Z	LooksRare: Exchange	26602204157466837	85775.895996
2022-05-02T17:27:41Z	None	2184930000000000	0.000000

0.6 1.a Azuki Historical Sales

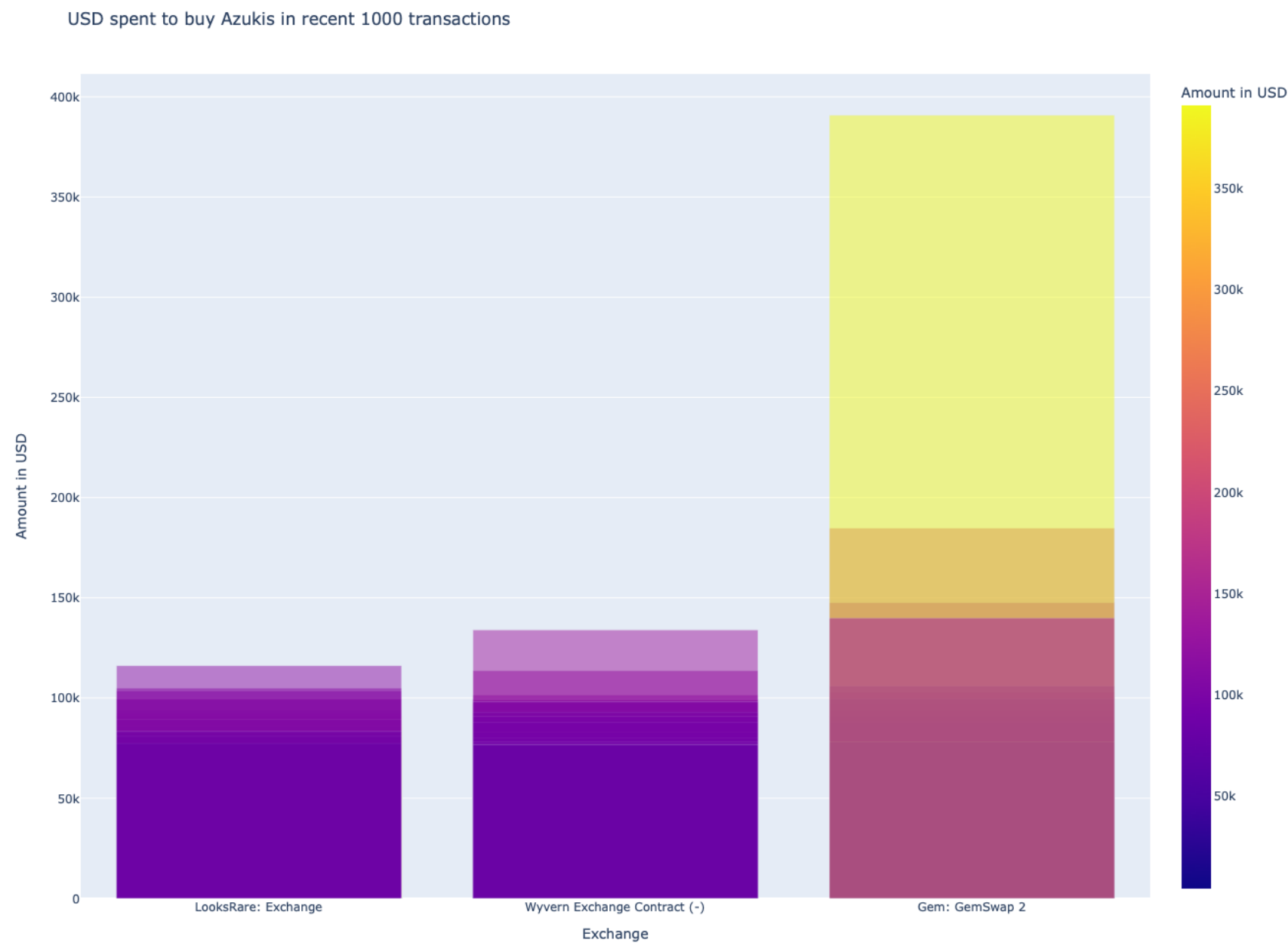
```
[41]: # Filter Through data for non null transactions
azuki_sales_df = azuki_tx_df[azuki_tx_df['value_quote'] != 0]
azuki_sales = azuki_sales_df[azuki_sales_df['to_address_label'].notnull()]

# Creating the plot using plotly express
azuki_fig = px.bar(azuki_sales,
    x='to_address_label',
    y='value_quote',
    color='value_quote',
    height=1020,
    width = 1000,
```

4. Data Analysis

Analyzing our data and developing figures to answer our questions

```
        barmode='overlay',
        labels={'value_quote':'Amount in USD', 'to_address_label':␣
↵'Exchange'}},
        title='USD spent to buy Azukis in recent 1000 transactions'
    )
azuki_fig.show()
```

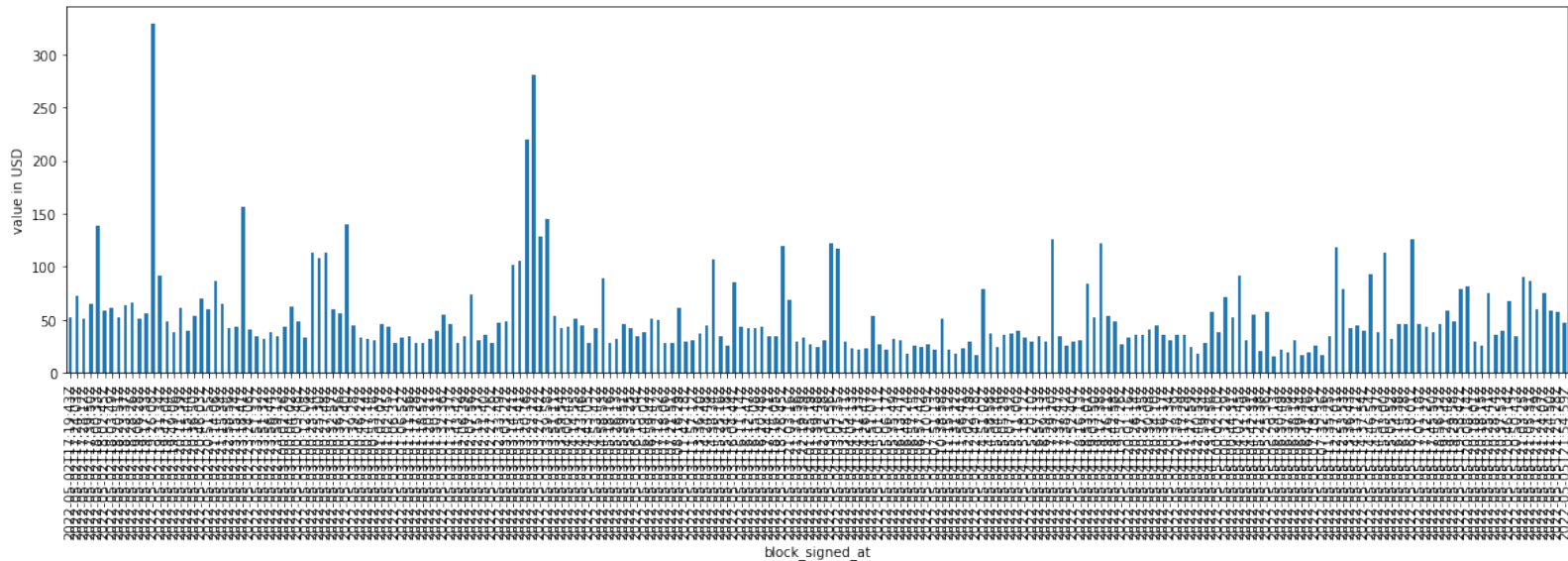


0.7 1.b Azuki transaction fees paid

```
[30]: # Filter Through data for non null transactions
azuki_fees = azuki_sales_df['fees_paid'].astype(int)/10**18*eth_value

azuki_fees.plot.bar(rot = 90, figsize = (20,5), ylabel = 'value in USD')
```

[30]: <AxesSubplot:xlabel='block_signed_at', ylabel='value in USD'>



0.8 2. Cryptopunks Daily Volume

```
[13]: # Create variables needed for owner data and append to url
cryptopunks_historical_url = url + chain_id + "/nft_market/collection" +␣
↵cryptopunks_address + api_no_option

# Get request
cryptopunks_historical_json = requests.get(cryptopunks_historical_url).json()

# Convert historical json data to a dataframe and view data
cryptopunks_df = pd.DataFrame(cryptopunks_historical_json['data']['items'])

# Set index to date
cryptopunks_df = cryptopunks_df.set_index('opening_date')

# Create Volume dataframe
cryptopunks_vol_df = pd.DataFrame(cryptopunks_df, columns =␣
↵['volume_quote_day', 'unique_token_ids_sold_count_day']).sort_index()
cryptopunks_vol_df.head()
```

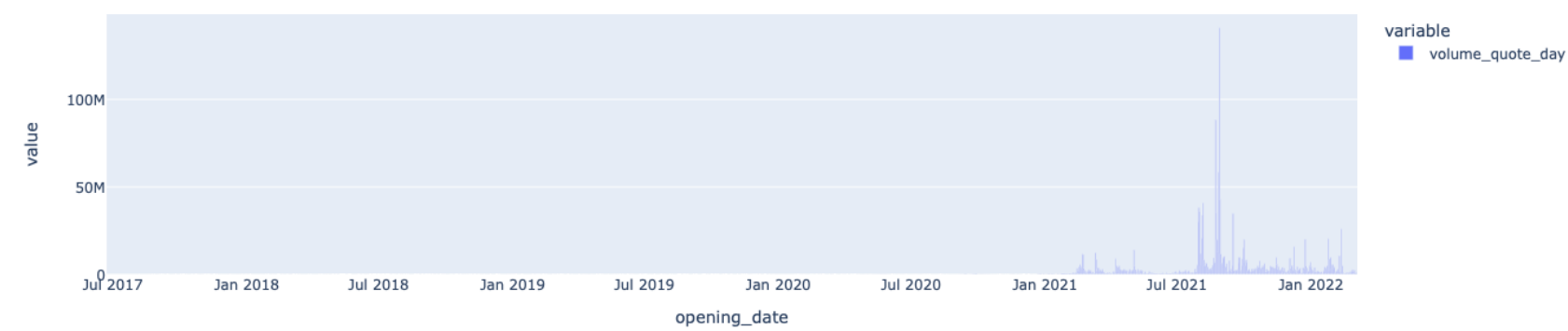
```
[13]:          volume_quote_day  unique_token_ids_sold_count_day
opening_date
2017-06-23                0.0                             19
2017-06-24                0.0                             22
2017-06-25                0.0                             11
2017-06-26                0.0                             18
2017-06-27                0.0                             35
```

```
[42]: # Plot Volume quote per day
cryptopunks_volume = cryptopunks_vol_df['volume_quote_day'].astype(int)

# cryptopunks_volume.plot.line(figsize = (20,4))
```

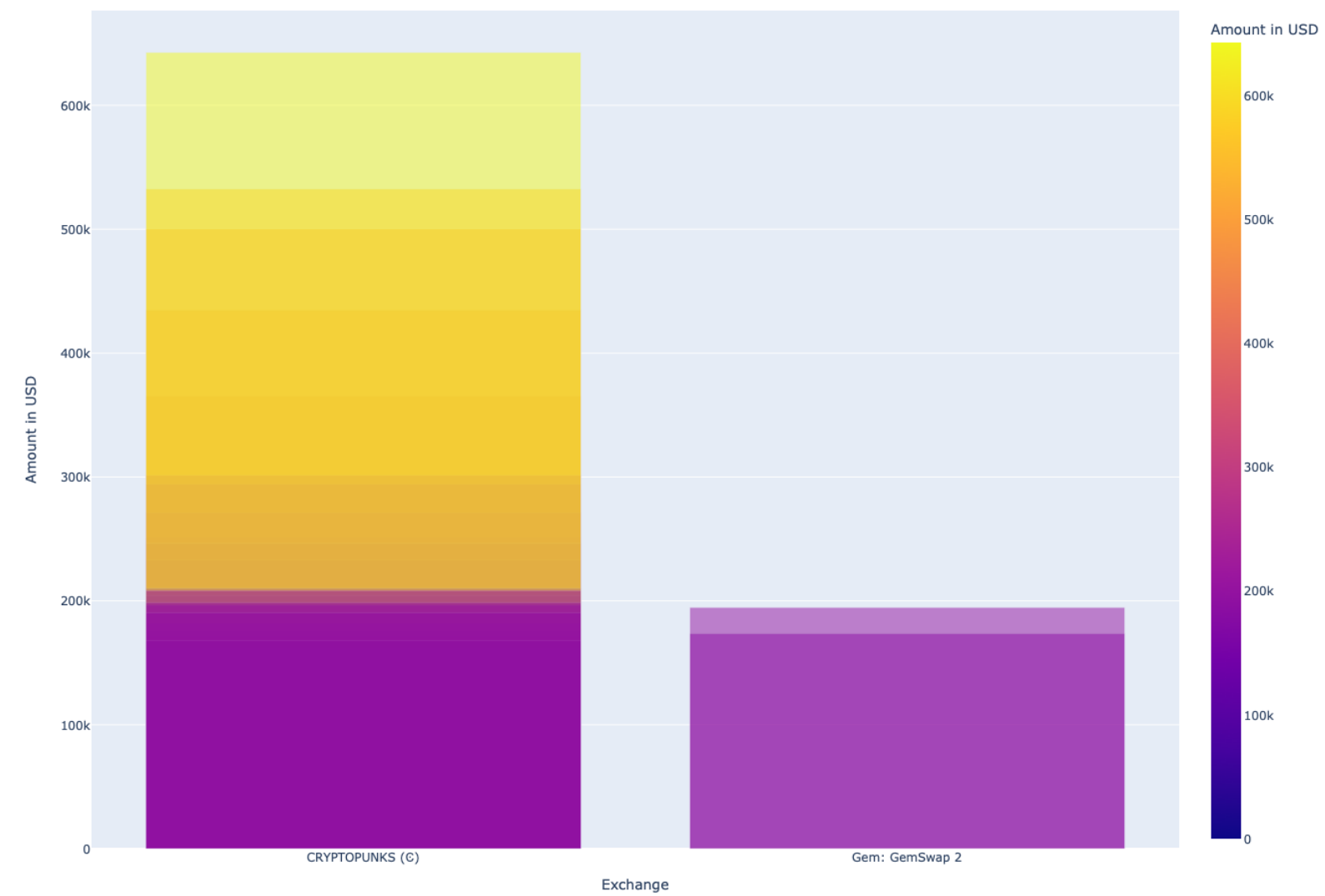


```
px.bar(cryptopunks_volume)
```



```
cryptopunks_fig = px.bar(cryptopunks_sales,
                        x='to_address_label',
                        y= 'value_quote',
                        color='value_quote',
                        height=1020,
                        width = 1000,
                        barmode = 'overlay',
                        labels={'value_quote': 'Amount in USD',
                                'to_address_label': 'Exchange'},
                        title='USD spent to buy Cryptopunks in recent 1000',
                        transactions'
                        )
cryptopunks_fig.show()
```

USD spent to buy Cryptopunks in recent 1000 transactions



0.9 2a Cryptopunks Historical transactions

```
[15]: # Querying the API for transaction data
cryptopunks_tx_url = url + chain_id + "/address" + cryptopunks_address +
    page_option + api_option
cryptopunks_tx = requests.get(cryptopunks_tx_url).json()

# Convert transactions data to dataframe
cryptopunks_tx_df = pd.DataFrame(cryptopunks_tx['data']['items'], columns =
    ['to_address_label', 'fees_paid', 'value_quote', 'block_signed_at']).
    set_index('block_signed_at').sort_index()

cryptopunks_tx_df.head()
```

```
[15]:
```

	to_address_label	fees_paid	value_quote
block_signed_at			
2022-04-27T17:02:10Z	CRYPTOPUNKS ()	9248158760553474	0.0
2022-04-27T17:02:42Z	CRYPTOPUNKS ()	9375445908544005	0.0
2022-04-27T17:03:33Z	None	46953125223110119	0.0
2022-04-27T17:05:29Z	None	48539446229017550	0.0
2022-04-27T17:20:09Z	CRYPTOPUNKS ()	6381749930928020	0.0

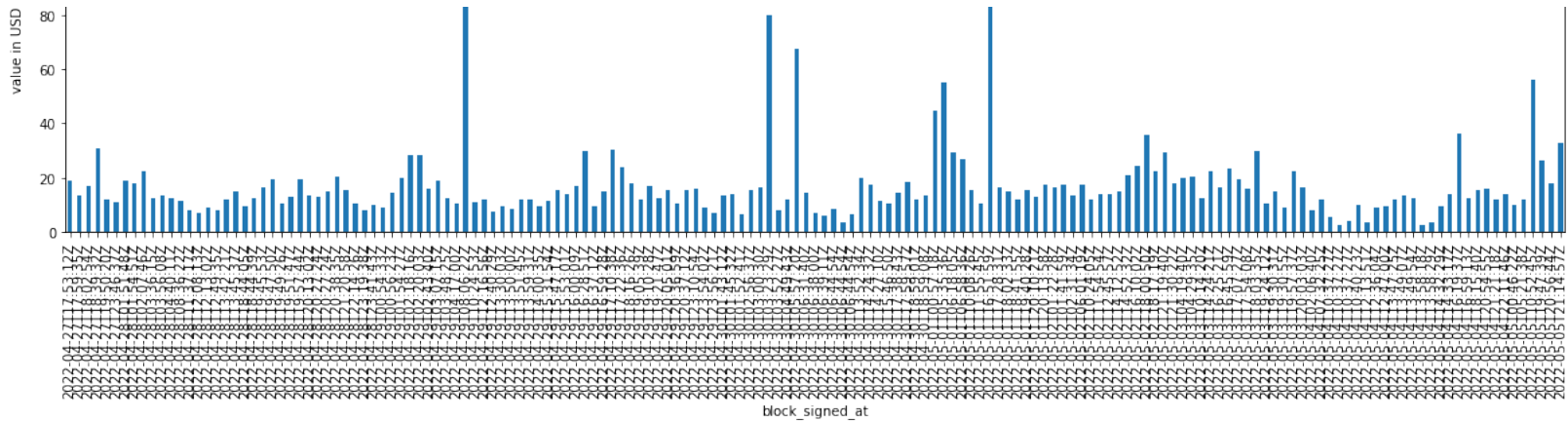
0.10 2.a Cryptopunks Historical Sales

```
[36]: # Filter Through data for non null transactions
cryptopunks_sales_df = cryptopunks_tx_df[cryptopunks_tx_df['value_quote'] != 0]
cryptopunks_sales =
    cryptopunks_sales_df[cryptopunks_sales_df['to_address_label'].notnull()].
    dropna()

# Creating the plot using plotly express
```

0.11 2.b Cryptopunks Fees paid

```
[17]: # Filter Through data for non null transactions
cryptopunks_fees = cryptopunks_sales_df['fees_paid'].astype(int)/
    10**18*eth_value
```



0.12 3. BAYC Daily Volume

```
[37]: # Create variables needed for owner data and add to url
BAYC_historical_url = url + chain_id + "/nft_market/collection" + BAYC_address_
    ↪+ api_no_option

# Get request
BAYC_historical_json = requests.get(BAYC_historical_url).json()

# Convert historical json data to a dataframe and view data
BAYC_df = pd.DataFrame(BAYC_historical_json['data']['items'])

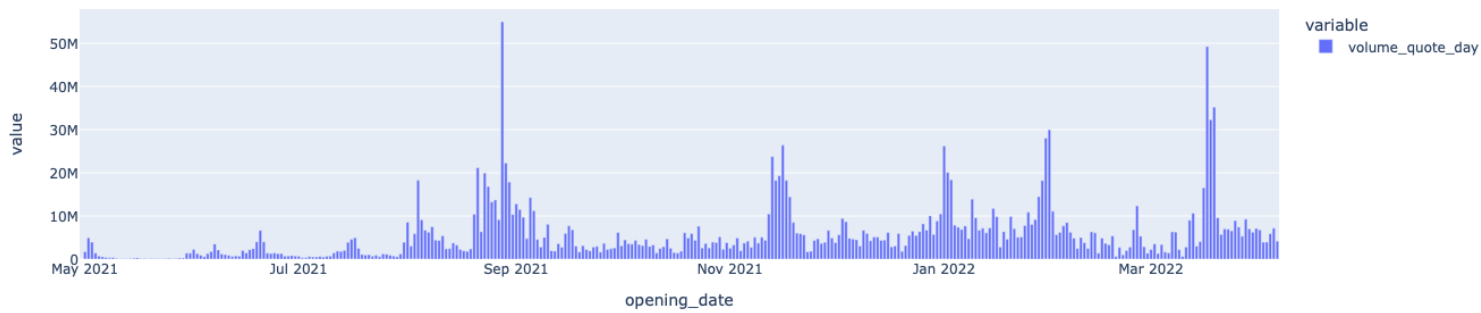
# Set index to date
BAYC_df = BAYC_df.set_index('opening_date')

# Create Volume dataframe
BAYC_vol_df = pd.DataFrame(BAYC_df, columns = ['volume_quote_day',
    ↪'unique_token_ids_sold_count_day']).sort_index()
BAYC_vol_df.head()
```

```
[37]:          volume_quote_day  unique_token_ids_sold_count_day
opening_date
```

```
[43]: # Plot Volume quote per day
BAYC_volume = BAYC_vol_df['volume_quote_day'].astype(int)

# BAYC_volume.plot.bar(figsize = (20,4))
px.bar(BAYC_volume)
```



0.13 3a BAYC Historical Sales

```
[20]: # Querying the API for transaction data
BAYC_tx_url = url + chain_id + "/address" + BAYC_address + page_option +
    ↪api_option
BAYC_tx = requests.get(BAYC_tx_url).json()

# Convert transactions data to dataframe
BAYC_tx_df = pd.DataFrame(BAYC_tx['data']['items'], columns =
    ↪['to_address_label', 'fees_paid', 'value_quote', 'block_signed_at']).
    ↪set_index('block_signed_at').sort_index()

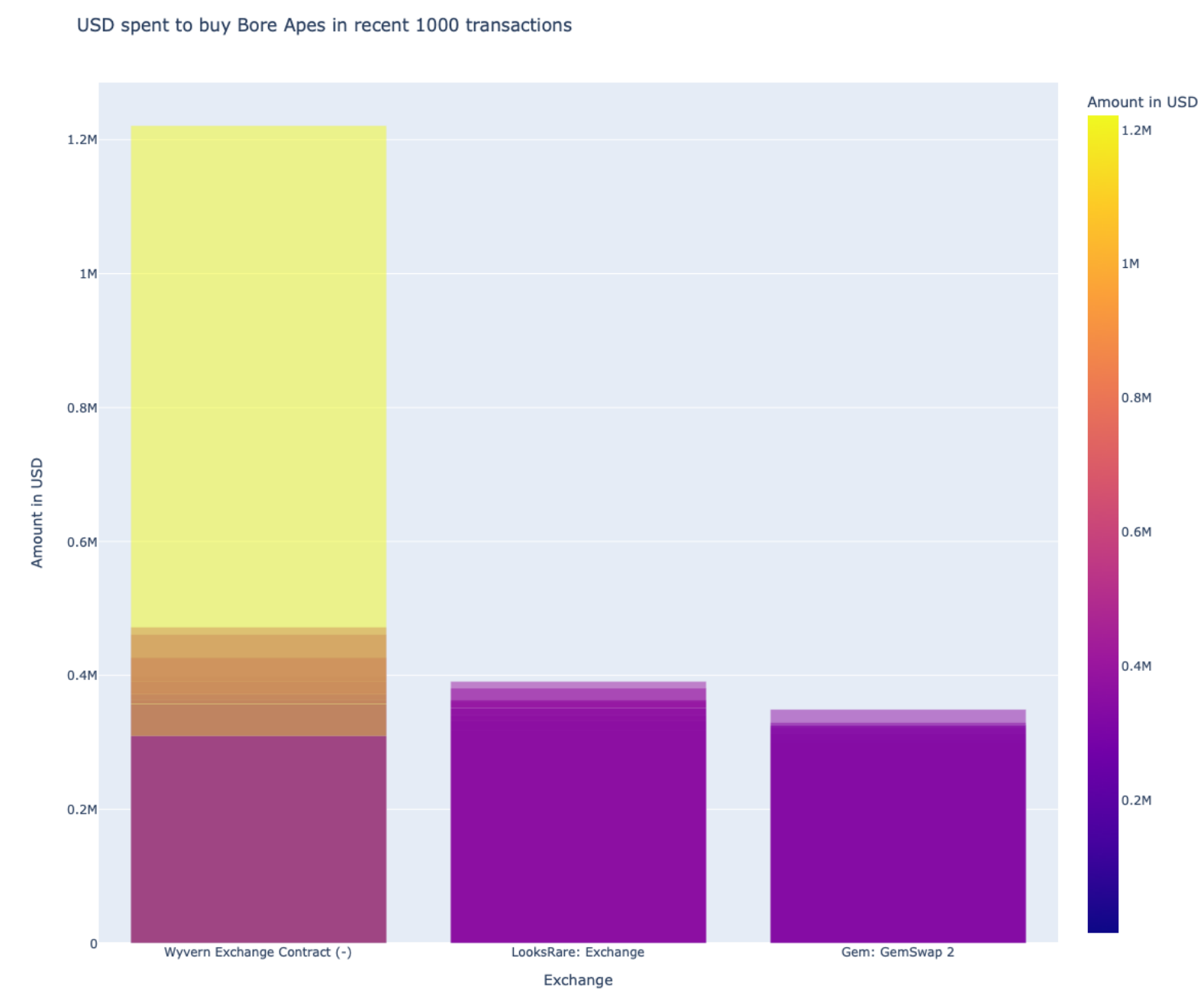
BAYC_tx_df.head()
```

	to_address_label	fees_paid	\
block_signed_at			
2022-05-02T19:17:57Z	None	5514685994543960	
2022-05-02T19:22:45Z	None	10393902970606424	
2022-05-02T19:22:45Z	None	9087780139176796	
2022-05-02T19:23:32Z	Wyvern Exchange Contract (-)	25104762520975735	
2022-05-02T19:30:41Z	None	3462010508841040	

	value_quote
block_signed_at	
2022-05-02T19:17:57Z	0.000000
2022-05-02T19:22:45Z	0.000000
2022-05-02T19:22:45Z	0.000000
2022-05-02T19:23:32Z	328807.601318
2022-05-02T19:30:41Z	0.000000

```
[21]: # Filter Through data for non null transactions
BAYC_sales_df = BAYC_tx_df[BAYC_tx_df['value_quote'] != 0]
BAYC_sales = BAYC_sales_df[BAYC_sales_df['to_address_label'].notnull()].dropna()

# Creating the plot using plotly express
BAYC_fig = px.bar(BAYC_sales,
                  x='to_address_label',
                  y='value_quote',
                  color='value_quote',
                  height=1020,
                  width = 1000,
                  barmode = 'overlay',
                  labels={'value_quote':'Amount in USD', 'to_address_label':↵
↵'Exchange'}},
                  title='USD spent to buy Bore Apes in recent 1000 transactions'
                  )
BAYC_fig.show()
```

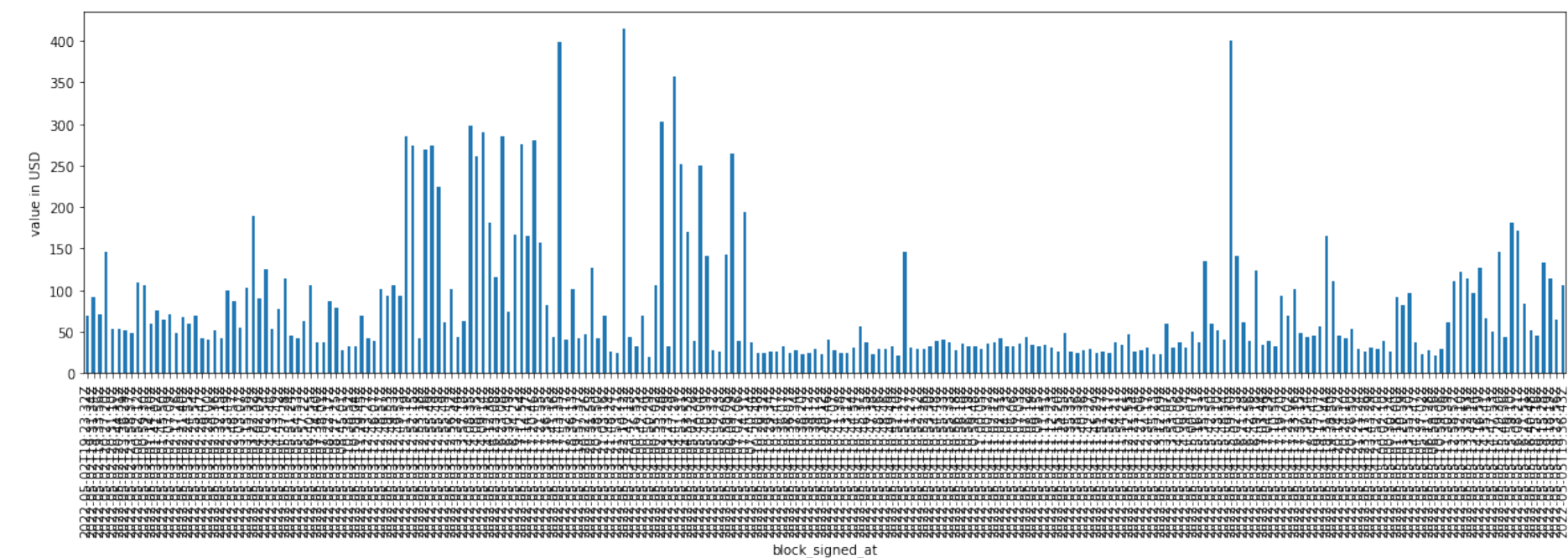


0.14 3.b BAYC Fees paid

```
[22]: # Filter Through data for non null transactions
BAYC_fees = BAYC_sales_df['fees_paid'].astype(int)/10**18*eth_value

BAYC_fees.plot.bar(rot = 90, figsize = (20,5), ylabel = 'value in USD')
```

```
[22]: <AxesSubplot:xlabel='block_signed_at', ylabel='value in USD'>
```



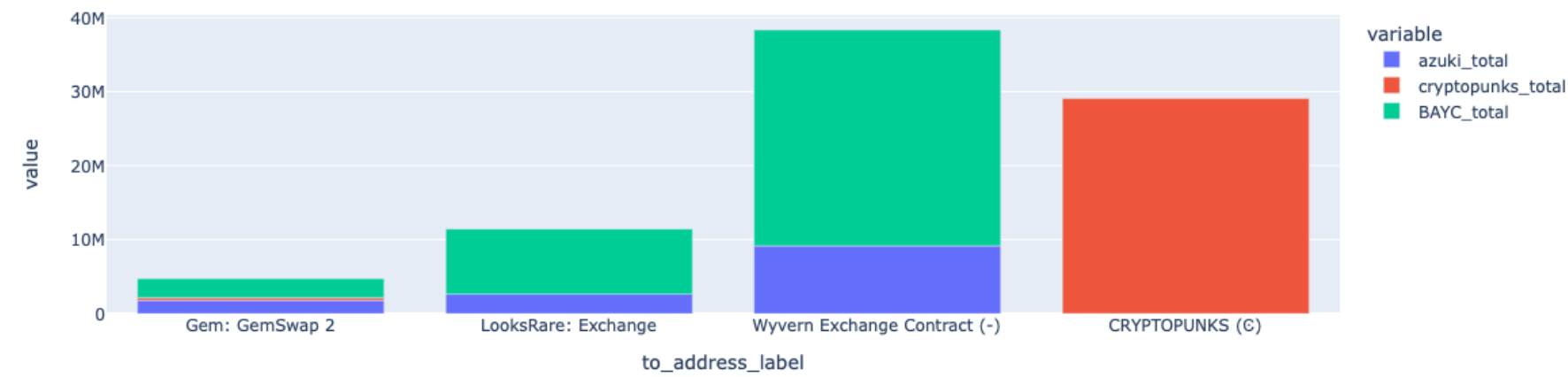
0.15 Combine Total Sales

```
[23]: # Group by address label and sum the value
azuki_total = azuki_sales.groupby('to_address_label').sum()
cryptopunks_total = cryptopunks_sales.groupby('to_address_label').sum()
BAYC_total = BAYC_sales.groupby('to_address_label').sum()

[24]: # Combine and rename columns for our total sales data
combined_totals = pd.concat([azuki_total,cryptopunks_total,BAYC_total], axis=1)
combined_totals.columns = ['azuki_total', 'cryptopunks_total', 'BAYC_total']

[25]: # Plot for combined figure
combined_total_fig = px.bar(combined_totals)

# Show Figure
combined_total_fig.show()
```



0.16 Combine Total Fees

```
[26]: # Group by address label and sum the value
combined_totals
```

```
[26]:
```

	azuki_total	cryptopunks_total	BAYC_total
to_address_label			
Gem: GemSwap 2	1.791827e+06	3.683260e+05	2.568193e+06
LooksRare: Exchange	2.623563e+06	NaN	8.814426e+06
Wyvern Exchange Contract (-)	9.173662e+06	NaN	2.918673e+07
CRYPTOPUNKS ()	NaN	2.909927e+07	NaN

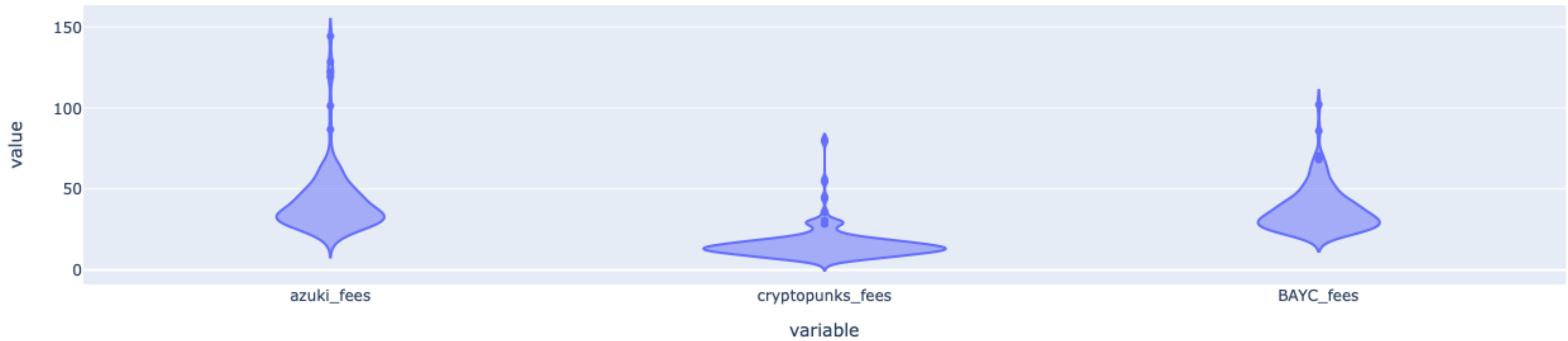
```
[38]: # Combine and rename columns for our total sales data
azuki_usd_fees = azuki_sales['fees_paid'].astype(int)/10**18*eth_value
cryptopunks_usd_fees = cryptopunks_sales['fees_paid'].astype(int)/
↪10**18*eth_value
BAYC_usd_fees = BAYC_sales['fees_paid'].astype(int)/10**18*eth_value

# Combine dataframe and drop nulls
combined_usd_fees = pd.concat([azuki_usd_fees.reset_index(drop=True),
                               cryptopunks_usd_fees.reset_index(drop=True),
                               BAYC_usd_fees.reset_index(drop=True)],
                               axis=1
                               ).dropna()

combined_usd_fees.columns = ['azuki_fees', 'cryptopunks_fees', 'BAYC_fees']
```

```
[39]: # Plot for combined figure
combined_fees_fig = px.violin(combined_usd_fees)

# Show Figure
combined_fees_fig.show()
```



5. Discussion

Combining our data and discussing our findings

Postmoterm

Did we find everything we expected to find?

- Our difficulties and how we dealt with them
- Additional questions that came up that we didn't answer
- What would we research next if we had more time?

8. Questions

Open floor Q&A with the audience

thank you,

