visualizing_the_10000_pizza_bitcoin_network

September 28, 2021

```
[]: # 10,000 bitcoin network
```

This lab will dig into the fates of the bitcoin transactions tied to the infamous 10,000 bitcoin pizza.

This code is based on code originally written by Allen Day and modified by Sohien Dane and Meg Risdal from these Kaggle kernels (parts 1, 2, 3). It will be used to visualize a directed graph representing Bitcoin transactions that follow the first known exchange of Bitcoin for goods on May 17, 2010 made by Laszlo Hanyecz.

In this lab, we will: 1. Retrieve as many transactions as possible from BigQuery within 2 degrees of separation from the pizza exchange. - Post-process the transactions to remove excess transactions from step 1 because the query was overly-greedy to ensure the number of table scans equals the degrees of separation. - Visualize the directed graph.

0.1 Extract all the related transactions from BigQuery

In these next steps you retrieve as many transactions as possible from BigQuery within 2 degrees of separation from the pizza exchange.

```
[17]: import pandas as pd
      from google.cloud import bigquery
[18]: bq = bigquery.Client()
[20]: QUERY_TEMPLATE = """
      SELECT
          timestamp,
          inputs.input_pubkey_base58 AS input_key,
          outputs.output pubkey base58 AS output key,
          outputs.output satoshis as satoshis
      FROM `bigquery-public-data.bitcoin_blockchain.transactions`
          JOIN UNNEST (inputs) AS inputs
          JOIN UNNEST (outputs) AS outputs
      WHERE inputs.input_pubkey_base58 IN UNNEST({0})
          AND outputs.output_satoshis >= {1}
          AND inputs.input_pubkey_base58 IS NOT NULL
          AND outputs.output_pubkey_base58 IS NOT NULL
      GROUP BY timestamp, input_key, output_key, satoshis
      0.00
```

```
[22]: def trace_transactions(target_depth, seeds, min_satoshi_per_transaction):
          Trace transactions associated with a given bitcoin key.
          To limit the number of BigQuery calls, this function ignores time.
          If you care about the order of transactions, you'll need to do_{\sqcup}
       \hookrightarrow post-processing.
          May return a deeper graph than the `target_depth` if there are repeated \sqcup
       \hookrightarrow transactions
          from wallet a to b or or self transactions (a \rightarrow a).
          MAX SEEDS PER QUERY = 500
          query = QUERY_TEMPLATE.format(seeds, min_satoshi_per_transaction)
          #print(f'Estimated total query size: {int(bq_assist.
       →estimate_query_size(query)) * MAX_DEPTH}')
          results = []
          seeds_scanned = set()
          for i in range(target_depth):
              seeds = seeds[:MAX_SEEDS_PER_QUERY]
              print("Now scanning {} seeds".format(len(seeds)))
              query = QUERY_TEMPLATE.format(seeds, min_satoshi_per_transaction)
              transactions = bq.query(query).to_dataframe()
              results.append(transactions)
              # limit query kb by dropping any duplicated seeds
              seeds_scanned.update(seeds)
              seeds = list(set(transactions.output key.unique()).
       →difference(seeds_scanned))
          return pd.concat(results).drop_duplicates()
[24]: MAX DEPTH = 2
      BASE_SEEDS = ['1XPTgDRhN8RFnzniWCddobD9iKZatrvH4']
      SATOSHI PER BTC = 10**7
[25]: df = trace_transactions(MAX_DEPTH, BASE_SEEDS, 0)
     Now scanning 1 seeds
     Now scanning 27 seeds
[26]: df.size
```

0.2 Post-processing the data pulled from BigQuery

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This stage handles the post-processing of data pulled from BigQuery in step 1. At the end, we'll have a graph of transactions for plotting in NetworkX.

```
[28]: df['date_time'] = pd.to_datetime(df.timestamp * 1000000)
[29]: df.head(3)
[29]:
             timestamp
                                                input_key \
      0 1511333577000
                        1XPTgDRhN8RFnzniWCddobD9iKZatrvH4
                        1XPTgDRhN8RFnzniWCddobD9iKZatrvH4
      1 1274888964000
      2 1281069800000
                        1XPTgDRhN8RFnzniWCddobD9iKZatrvH4
                                 output_key
                                                 satoshis
                                                                     date_time
      0 1MyQWqA8Ykcm9W5qKcYGokpYsV8tJDYguy
                                                 42419301 2017-11-22 06:52:57
      1 1JXFXUBGs2ZtEDAQMdZ3tkCKo38nT2XSEp 250000000000 2010-05-26 15:49:24
      2 1CQDMQ5HDY2gVoV17H5fptsbzTA6qxwbjT 156800000000 2010-08-06 04:43:20
[31]: df.head(3).to_csv('transactions.csv')
[33]: def dig_row(row, seeds, min_satoshis, trace_from_key):
          if row['satoshis'] < min_satoshis:</pre>
              return None
          if trace_from_key and row['input_key'] not in seeds:
          elif not trace_from_key and row['output_key'] not in seeds:
              return None
          seeds.add(row['output_key'])
          return row
      def single_pass_dig(initial_seeds, input_df, initial_datetime=None,u
       →min_satoshis=0, trace_from_key=True):
          df = input_df.copy()
          active seeds = set(initial seeds)
          if trace_from_key and initial_datetime is not None:
              df = df[df['date_time'] >= initial_datetime]
          elif not(trace_from_key) and initial_datetime is not None:
              df = df[df['date_time'] <= initial_datetime]</pre>
          df.sort_values(by=['timestamp'], ascending=trace_from_key, inplace=True)
          transactions = []
          for index, row in df.iterrows():
              rv = dig_row(row, active_seeds, min_satoshis, trace_from_key)
              if rv is not None:
                  transactions.append(rv)
          return pd.DataFrame(transactions)
[34]: future_transactions = single_pass_dig(BASE_SEEDS, df,
                                             initial_datetime=pd.to_datetime("May 16,__
       →2010"))
```

```
[35]: total_flows = future_transactions[['input_key', 'output_key', 'satoshis']].
       →groupby(
          by=['input_key', 'output_key']).sum().reset_index()
      total flows.head(3)
[35]:
                                  input_key
                                                                     output_key
        12hEZssdWGLS8UsavJzYAUgxiXLA4x4bu7
                                             1GAnVbQpkT2CFm9RAu13R22g5A9MXPR6vG
      1 13gWLbMdqyftxUw7mvtn4FQdGaAAWkErgF
                                             1GAnVbQpkT2CFm9RAu13R22g5A9MXPR6vG
      2 13pVVjBntbheW93GNCWKA1ZMdbQrRopNsp
                                             1GAnVbQpkT2CFm9RAu13R22g5A9MXPR6vG
             satoshis
       104687000000
      1 104687000000
      2 104687000000
[36]: total_flows.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 106 entries, 0 to 105
     Data columns (total 3 columns):
          Column
                      Non-Null Count
                                      Dtype
                      _____
      0
          input_key
                      106 non-null
                                      object
      1
          output_key 106 non-null
                                      object
                      106 non-null
                                      int64
      2
          satoshis
     dtypes: int64(1), object(2)
     memory usage: 2.6+ KB
[38]: total_flows.head(3).to_csv('total_flows.csv')
```

0.3 Visualizing the network

This code visualizes a directed graph representing Bitcoin transactions that follow Hanyecz's pizza buying transaction.

The figure generated here is similar to the one used in the "Bitcoin in BigQuery: Blockchain Analytics on Public Data" blog post by Allen Day and Colin Bookman.

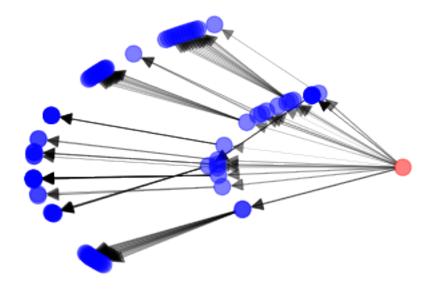
```
[39]: ## purchase address
hanyecz = "1XPTgDRhN8RFnzniWCddobD9iKZatrvH4"
```

We use the library networkx below to visualize a network of transactions following the pizza purchase by up to 2 degrees (As noted here, our code may return a deeper graph than the target depth if there are repeated transactions from wallet a to b or or self transactions (a -> a)).

Hanyecz's payment address is depicted as a red circle while other addresses are blue circles. Arrowheads indicate direction of Bitcoin flow following the pizza purchase transaction. Stroke width is approximately proportional to the amount of Bitcoin moving between addresses.

```
[40]: import networkx as nx
      import matplotlib.pyplot as plt
[42]: graph = nx.from_pandas_edgelist(total_flows,
        source = "input_key",
        target = "output_key",
        create_using = nx.DiGraph())
      pos = nx.kamada_kawai_layout(graph)
[43]: ## color the hanyecz red and the rest of the nodes blue
      node_colors = list(map(lambda x: "red" if x == hanyecz else "blue", graph.
       →nodes))
[45]: ## calculate width based on satoshis
      satoshi_stats = total_flows.describe()['satoshis']
      def get_width(x, stats):
          if x < stats['25%']:</pre>
              return 1
          elif x \ge stats['25\%'] and x < stats['50\%']:
          elif x \ge stats['50\%'] and x < stats['75\%']:
              return 3
          else:
              return 4
      edge_widths = total_flows['satoshis'].apply(lambda x: get_width(x,_
       ⇔satoshi_stats))
[46]: f = plt.figure()
      nx.draw_networkx(graph, pos,
        with_labels = False,
        alpha = 0.5,
       node_size = 150,
        node_color = node_colors,
       width = edge_widths / 4,
        arrowsize = 20,
        ax=f.add_subplot(111))
      plt.title("BTC inputs upstream of pizza purchase address (red)")
      plt.axis('off')
      f.savefig("graph.png")
```

BTC inputs upstream of pizza purchase address (red)



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