Trading Network Performance for Cash in the Bitcoin Blockchain

Enrico Tedeschi, Håvard D. Johansen and Dag Johansen UIT The Arctic University of Norway, Tromsø, Norway

Keywords: Bitcoin, Blockchain as a Service, Longitudinal Study, Performance, Transaction Latency.

Abstract: Public blockchains have emerged as a plausible cloud-like substrate for applications that require resilient communication. However, sending messages over existing public blockchains can be cumbersome and costly as miners require payment to establish consensus on the sequence of messages. In this paper we analyze the network performance of the Bitcoin public ledger when used as a massaging substrate. We present several real-world observations on its characteristics, transaction visibility, and fees paid to miners; and we propose two models for fee-cost estimation. We find that applications to some extent can improve messaging latency by paying transaction fees. We also suggest that spendings should be kept below 300 Satoshi per byte.

1 INTRODUCTION

The Blockchain technology used by popular crypto currencies like Bitcoin¹ and Ethereum, are essentially Peer-to-Peer (P2P) broadcast oriented Group Communication Systems (GCSs) (Cheriton and Zwaenepoel, 1985), where all members see all messages, and in the same order. With a built-in fee system that enables operators to make money for storing and forwarding messages, several decentralized P2P blockchain systems have emerged, providing a common ground for mutually distrusting entities to communicate. Due to their promise of highly resiliency, *blockchain-as-a-service* is currently being touted as a promising permissionless cloud-like building block for critical services in, for instance, the finance and health domains.

Unlike traditional multicast oriented GCSs like Horus (van Renesse et al., 1996) and Totem (Moser et al., 1996), blockchains have the unique property that all broadcast messages are kept and made available to all participants, potentially for the system's lifetime. For blockchains, consensus among participants on the total ordering of messages, and hence also consensus on the resulting data-structure or ledger, is achieved through computational puzzles that randomly grant members a time-limited exclusive right to dictate the next batch of messages to be put on the channel.

As with other permissionless systems such as SecureRing (Kihlstrom et al., 1998) and Fireflies (Jo-

¹The Bitcoin currency is denoted BTC or B.

hansen et al., 2015) that are designed to be highly resilient to intrusions and attacks, providing reliable service by masking Byzantine failures limits scalability. Because blockchains are designed to retain all messages, they are particularly vulnerable to denialof-service through simple flooding attacks. If an attacker can send messages at an unbounded rate, he can quickly swamp the storage and network capacity of the service. Even benign usage might prove problematic. For instance, if Bitcoin would have the same transaction rate as a VISA circuit, with between 2000 to 56000 transactions/sec (Croman et al., 2016), its blockchain structure would grow about 1 MB per 3 seconds.

To throttle its blockchain growth rate, Bitcoin adjusts the difficulty of its cryptographic puzzles to match the aggregate mining capacity of the network, towards a target average block creation time of 10 min per block. In combination with its current block size limit of 1 MB, Bitcoin's transfer capacity is roughly a meager 1.667 kBps, or approximately three transactions per second. This capacity is shared between all concurrent clients. Indeed, scalability and network performances are urgent concerns in existing Blockchain-based systems.

This paper presents key observations from our ongoing longitudinal study on the performance of the Bitcoin blockchain. We provide detailed insights and analysis on several important characteristics of Bitcoin, including paid fees and the size of blocks, and show how the rewards to miners have changed over time from a more recent view point compared to earlier studies (Möser and Böhme, 2015; Rizun, 2015).

Trading Network Performance for Cash in the Bitcoin Blockchain

DOI: 10.5220/0006805906430650

In Proceedings of the 8th International Conference on Cloud Computing and Services Science (CLOSER 2018), pages 643-650 ISBN: 978-989-758-295-0

Copyright © 2019 by SCITEPRESS - Science and Technology Publications, Lda. All rights reserved

Furthermore, we analyze the correlation between the fee paid from a transaction and its *latency*, or the time it takes to become visible in the whole network. We propose two different models to describe how applications should spend money to improve network performance. Although the studies presented in this paper are restricted to the Bitcoin system, we conjecture that our observations are transferable to similar P2P systems that rely on computational expensive proof-of-work for consensus and fee-based incentives.

2 BACKGROUND

This paper considers blockchains as a communication substrate, where a set of *client* or *application* processes communicate by sending and receiving messages. In blockchain systems, a message is often referred to as a transaction, a notation we also adapt in this paper. The blockchain substrate handles all transactions in batches, known as *blocks*. Each block *B* can be in one of two states: *proposed* or *accepted*, and might contain zero or more transactions from zero or more clients. The system in *permissionless* in that there exist no central authority that coordinate or regulate participation or usage.

A blockchain has one or more *miner* processes that act as the ingress points for transactions submitted by the clients. The set of all miners collaborate to decide on which transactions to admit and their ordering. Every miner has a *mempool* containing the new and unapproved transactions. Applications are free to submit transactions to any miner's mempool, and miners are free to choose which transactions to include in their blocks. Most blockchain systems, including Bitcoin, enforce a strict upper bound Q on the block size, which also limits the number of transactions each block can contain.

The blockchain data structure, often referred to to as a ledger, is maintained by a P2P overlay network where members cooperate to verify and distribute blocks such that each member process has a full replica of the data structure in local storage. The integrity of the data structure is dependent on consensus among the set of correct member processes on which blocks are in the blockchain and their total order. For this, existing blockchain systems, like Bitcoin, use the Nakamoto consensus protocol. This protocol relies on members solving computationally complex cryptographic puzzles as proof-of-work for admitting new blocks, commonly referred to as mining. Once a miner has solved a puzzle, it broadcasts the block along with a solution to the puzzle so that all other nodes can check its correctness. The block is then tentatively recorded in the blockchain.

In the early days of Bitcoin, it was possible to mine productively with commodity desktop or laptop computers. Nowadays, successful miners use highly specialized hardware called Application Specific Integrated Circuitss (ASICs) (Taylor, 2017), which typically offer up to 100 times improved performance over commodity CPUs and GPUs.

The cost associated with mining was defined by Rizun (2015) to be:

$$\langle C \rangle = \eta h \mathcal{T} \tag{1}$$

Here η is the cost per generated hash, *h* is the miner's individual hash rate and \mathcal{T} is the block creation time. Hence, the block creation time is directly proportional to the hashing cost. The underlying assumption of proof-of-work consensus is that the high η value, due to the energy cost of mining, will discourage and limit malicious behavior. At the same time, benign participation is promoted by means of incentives: the nodes receive payment as a reward for solving puzzles.

As miners compete to solve the latest puzzle, it may happen that two or more nodes succeed at approximately the same time. This may result in different blocks with different transactions being proposed for the blockchain. Thus, proposing further blocks may result in different chains, often referred to as a fork. To break such ties, Bitcoin adopts the simple rule that the longest consecutive chain of blocks wins. Therefore, a tentative block of transactions may be discarded, which is known as *orphaning*. The recording of a block is only considered permanently accepted after five additional blocks have been added, approximately after 1 hour. Thus, eventually the nodes reach consensus on the ordering of all the blocks on the blockchain.

Given a transaction t from some client c_1 to some other client c_2 , we have the following:

Definition 1. The commit latency of a transaction t is the time from when c_1 first proposes a transaction t to a mining pool, to when some block B including t is first mined and permanently accepted.

Note that the total *end-to-end latency* of t also includes the time it takes for B to be delivered to c_2 . However, as blockchain clients must pull the system for updates, the last-hop delivery time will depend to the application specific pull interval. For generality, this paper will therefore only consider the commit latency.

In existing blockchain systems, miners can freely choose which transactions to include when proposing a new block. A client must therefore negotiate with miners to have its transactions included. To entice miners, each transaction t can include a transaction fee t_f paid by the sender to be claimed by the

miner whom first successfully include t in an accepted block. Due to the cost of mining, most miners are assumed to behave rationally (Aiyer et al., 2005): following the blockchain protocol, but such that their mining rewards are optimized. Hence, we conjecture that it is possible to use the transaction fee mechanism to improve messaging performance, which will be the focus of the remainder of this paper.

3 OBSERVATIONS

In this section we describe key observations and insights from our studies on the public Bitcoin ledger. These form the basis of our latency-fee models in Section 4. We start this section, however, by describing our method for collecting observational data.

3.1 Methodology

There are several methods that can be used to study Bitcoin. *Real-time analysis* requires setting up and operating one or more full Bitcoin nodes that connect to the P2P network and record traffic. The advantages of this approach is that some of the inner-node communication can sampled, including block propagation time and orphaning rate. However, to obtain usable coverage, multiple geographically dispersed nodes must be set up and injected into the network. Each one must download and store the full ledger and participate in the forwarding of new blocks. At time of writing, the full ledger of data requires 250 GB of storage. This requires significant up-front hardware investments and might potentially disrupt some of the system's characteristics that are under study.

Another approach is to use the *Bitcoin Core application* (van der Laan et al., 2017), which downloads the full ledger into local storage, but without having to run the full P2P protocol. Retrieved data does, however, not include the block propagation time or information from miners, which we require for our studies. Downloading the full ledger can also take significant time (in our case it took four days), and requires significant available disk capacity.

In our studies, we instead adapted a similar methodology to the one used by Möser and Böhme (2015), collecting data from some of the many *online third-party APIs*, made available by various organizations that are already monitoring the Bitcoin system, including tradeblock.com and blockchain.info. Websites like coinbase.com also provide useful information about the money exchange price, and provide an API along with libraries, like *forex-python*,² which we



Figure 1: Observed transaction fee (t_f) distribution from 2013 to 2017.

used.

Data was retrieved from these public APIs, and collected as JavaScript Object Notation (JSON) objects stored locally in Pandas data frames (Augspurger et al., 2012). Some data not available in these APIs directly, was instead scraped from the sites' HTML pages. The data was processed and visualized with *Matplotlib* (Hunter et al., 2017) and *Seaborn* (Waskom et al., 2016). This approach enabled us to analyze a considerable part of the blockchain with little up-front investment in computational resources.

For this paper, we elected to study data in the range from April 2013 to September 2017, sampling more than 120 million transactions and 100000 blocks. Several studies have already been conducted on Bitcoin data before 2013 (Croman et al., 2016; Houy, 2014; Möser and Böhme, 2015; Rizun, 2015), and the popularity of the system before 2011 was low. Interpreting data outside our selected date range would probably not generate new insight. Table 1 names and summarizes the exact segments retrieved and used in this paper.

3.2 Transaction Fees

In this section, we present our observations on how the transaction fees t_f , paid by the clients to the miners, have changed over time. For each transaction t, The fee t_f is the difference between the sum of all input values t_{in} and the sum of every output values t_{ou} . If n is the number of inputs and m is the number of outputs, then we have:³

²https://pypi.python.org/pypi/forex-python

³The unit of Equation 2 is B.

Name	Start Date	End Date	Block Height Range
2009	09-01-2009 03:54:25	08-03-2009 06:31:22	1 - 6710
2011	01-04-2011 19:58:59	09-05-2011 12:58:13	116167 - 122876
2013	21-04-2013 03:03:51	01-06-2013 12:37:51	232333 - 239042
2015	21-03-2015 04:01:39	06-05-2015 14:37:12	348499 - 355208
2017	15-12-2016 18:17:45	19-06-2017 12:04:23	443 599 - 471 951

Table 1: Bitcoin blockchain regions analyzed.

$$t_f = \sum_{i=1}^n t_{in_i} - \sum_{i=1}^m t_{ou_i}$$
(2)

Figure 1 plots the calculated values for t_f in blocks from 2013 to late 2017, categorized into six payment classes ranging from 0 to 0.01 Å. As we can see in the figure, in the first half of 2016 fees between 0 and 0.0001 Å almost disappeared. Considering that the Bitcoin price raised from less than 1000 USD to more than 5000 USD between mid 2016 and second half of 2017, this is indicative of a huge increment in the monetary value collected by miners. If we compare the Bitcoin price and the fee paid in USD, we see a substantial co-movement, which indicates that Å is the dominant unit to consider when deciding about what fee to offer.

Because the number of bytes per transaction can vary in all major blockchain systems today, including Bitcoin, an interesting metric to study is how many B per bytes a transaction t has to offer in payment to miners. This metric is known as the *fee density* t_p (Rizun, 2015). For some transaction t, with associated fee t_f and having a payload of t_q bytes, the fee density is defined as:

$$t_{\rm p} = \frac{t_f}{t_q}.\tag{3}$$

Figure 2 plots the observed fee density, similarly to the fee plot in Figure 1. The observed average transaction size t_q is 500 B. At the end of 2017, we see some transactions offering less than 0.0001 B in payment. We observe almost no transactions with fee density $t_{\rho} = 0$. This indicates that density has become an important metric for miners when deciding whether or not to include a transaction in their next block.

3.3 Transaction Latency

As shown above, a blockchain-based application may attempt to offer various mining fees to improve the commit latency of its messages. In the following we will investigate to what extent we are able to do so in practice.

The experienced commit latency t_l of most transactions can relatively easily be observed in the avaiable data. All transaction are timestamped when



Figure 2: Observed fee density (ρ) distribution from 2013 to 2017.



Figure 3: Relation between t_l and t_f grouped by year.

added to a mining pool, and blocks are similarly timestamped when mined.

Let B_{epoch} be the timestamp of some block *B* containing the transaction *t*, and let t_{epoch} be the timestamp of when that transaction was first added. Then the commit latency of observed transactions *t* can simply be calculated from:

$$t_l = B_{epoch} - t_{epoch} \tag{4}$$

Figure 3 plots the observed transaction latency (in hours) against the five fee density classes for each



Figure 4: Daily miners revenue divided in block reward R and the sum of transaction fees M.

year included in our study. In all cases, we observe that paying transaction fees gave a significant boost to latency, and that in 2017 doing so became more important than previously. We also observe the existence of a threshold from where increasing payment has little effect. For the years 2013–2016, the threshold was around 0.0002 B, while in 2017 it increased to 0.0006 B.

In addition to the total mining reward (M) from all transaction fees in a block, miners also receive a block reward (R) for each mined block. The block reward has historically been an important incentive for miners to produce blocks, regardless of the transaction fees offered by clients. However, the reward mechanism in Bitcoin is designed to halve the size of R every 210000 blocks. As can be seen in Figure 4, in the period from 2009 to 2013 miners had little considerations for the transaction fees, and relied more on the reward R. In mid 2016, when the block reward was last halved, we observe a clear shift in how the miners profit from their efforts, becoming more influenced by the transaction fees.

This observation is not surprising as we expect most miners to be rational (Aiyer et al., 2005), trying to optimize their profit. With less block reward, rational miners will need to prioritize transactions with a higher fee density over lower ones until the max block size is reached. If the total fee of the included transactions is less than the expected monetary cost of mining the block $\langle C \rangle$, the miner may even opt to wait until a higher density transaction arrives. This can significantly increase experienced commit latency and jitter. We expect *M* to overcome *R* by 2020 when the reward is halved again to 6.25 β . Hence, for applications that intend to use blockchain as a service for communication, there is a clear potential and need for clever usage of the transaction fee mechanism to optimize



Figure 5: Fee-latency interpolation F with a 2 and 39 degrees polynomials for Bitcoin transactions analyzed in 2017.

the commit latency and the monetary cost of sending messages.

4 MODELS

With data from the 2017 transactions, we generated two models that applications can use when deciding what transaction fees to offer.

4.1 Fee-latency

The first model F describes the expected latency of some transaction t given some transaction fee t_f . We compute two variants of F using polynomial regression: one of degree 2 (F^2) and one of degree 39 (F^{39}). The lower degree regression is used to show the general trend, while the higher degree one is used to show the utility threshold. The resulting regressions are shown in Figure 5. Measured Mean Absolute Error (MAE) was 1.755 for F^2 and 1.7476 for F^{39} . For some given transaction fee x, the function F^2 is given by

$$F^2(x) = 6248x^2 - 555.8x + 1.42 \tag{5}$$

From the plot of F^2 , we we see a clear linear trend that transactions offering higher fees experience lower commit latency, which is what we expected. From F^{39} in Figure 5, we also see a clear threshold at about 0.007 β when the benefits of adding extra fee starts declining.

4.2 Fee Density-latency

As argued in Section 3.2, miners nowadays tend to select transactions based on fee density, rather than solely on the fee amount. We therefore generate a



Figure 6: Fee density vs latency interpolation with a 2 and 39 degrees polynomial for Bitcoin transactions analyzed in 2017.

model *D* that provides the expected latency t_l as a function of the fee density $t_p = t_f/t_q$.

Similar to *F*, we compute two variants of *D* using polynomial regression: one of degree 2 (D^2) and one of degree 39 (D^{39}). The resulting regressions are shown if Figure 6. Measured MAE was 1.749 for D^2 and 1.837 for D^{39} .

Similarly to *F*, we also observe in *D* a clear trend that offering higher fee-density transactions improves commit latency. The threshold for diminishing returns is about 300 Satoshi per byte.⁴ Paying a higher fee per byte gives little improvements The calculated polynomial for D^2 is given by:

$$D^{2}(x) = \frac{5.416}{10^{8}}x^{2} - \frac{2.215}{10^{3}}x + 1.598$$
(6)

5 DISCUSSIONS

When Bitcoin was first released, one of its strengths was its decentralized P2P architecture. Miners could join the network all over the world, and more miners meant a more robust service.

Over the years though, the opportunities to exchange block and transaction rewards into hard cash enticed more and more people around the world to join the mining effort and make money. This increased the total hashing power of the system, but also increased the difficulty of the proof-of-work puzzle, as Bitcoin is designed to do. Eventually the puzzles became too difficult for most individual casual miners to solve, and people started teaming up into mining pools to share both computational power and profit.

Today only a few large mining pools remain, and the ability of the system to make progress has to a large extent been centralized. Still, the mechanisms controlling mining are governed by marked forces that remain to be exactly described. This may be a difficult task as most large mining pools withhold information about their number of miners, the hardware they use, their profit, their transaction selection criteria, etc. Observational studies, like the one described here, might be the only plausible method for understanding the mechanisms governing these systems.

Towards that end, Table 2 summarizes our findings by listing the effect of changing two key design parameters in Bitcoin: the block size Q and the block creation time T. Decreasing Q might first appear as a good solution for the number of advantages it has. However, its few disadvantages are critical for both performance and scalability. We therefore deem that decreasing Q is ill judged. It is likely much easier to deal with the orphan rate amplification resulting from increasing Q.

For the block creation time \mathcal{T} , we have the opposite scenario. An increment in \mathcal{T} will reduce performance as miners will make less profit and thus have less incentive to mine. The only advantage is a lower orphaning rate since blocks will have more time to propagate.

As we can observe from Table 2, the throughput γ increase when either the block size Q is raised or the creation time \mathcal{T} is lowered. According to Croman et al. (2016), the block size should not exceed 4 MB given $\mathcal{T} = 10$ min. A good compromise could be to increase the block size limit to 1.5 MB and lower the creation time to 8 min. In that way, the system have a potential throughput capacity of 3.20 kBps and 10 transactions per second.

We cannot find a clear and general relation between Q and t_f in our studies. They seem to relate only when the drastic change in the block size occurs. We do, however, find that from 2013 to 2017 the relation between t_f and t_l became more noticeable day-by-day, having almost an inverse proportionality in the latest data from 2017, as seen in Figure 3. In 2017, zero-fee transactions almost disappeared from the system. This is probably due to the incredibly high commit latency many clients were experiencing at that time, with zero-fee transactions taking up to an average of 33 h to get committed into blocks. At the same time, clients that paid only modest transaction fees, less than 0.0002 B, experienced commit latencies of only 5 h, while the ones that paid between 0.0008 B and 0.0010 B expected latencies of less than 1 h. Hence, applications that intend to use blockchains as a service for communication will benefit from having a dynamic fee-latency prediction model, like the one described here, to optimize performance.

 $^{^{4}1}$ Satoshi = 0.00000001B.

	Higher ↑	Lower ↓
Q	 + improved scalability with more transactions accepted per day + improved commit latency t_l ± lower fees (good for clients, bad for miners) - orphan rate amplification - increased centralization - congestion concern solved with transaction eviction by miners - no permanent effect 	 + no transaction spam + no 0-fee transactions + less mining cost + less propagation time + less chance of orphaning ± higher fees (good for miners bad for clients) - less throughput - higher commit latency t_l
T	 + lower orphaning rate + no physical changed needed to support faster inner node communication - lower throughput γ (unless <i>Q</i> is increased) - less scalable (unless <i>Q</i> is increased) - mining profit is confined 	 + higher throughput γ + system is more scalable + increased mining profit - require faster inner-node communication - exponential increment of orphaning rate

Table 2: Scalability and performance tradeoffs when changing block size Q and block creation time T.

6 CONCLUSION

The Bitcoin blockchain has undoubtedly emerged as a notable substrate and service for communication. The built in mechanism for gaining monetary value by doing useful work for others has clearly moved the underlying architecture from its initial P2P model, to a more centralized model resembling a public cloudlike service. Individual incentives for providing the service is no longer motivated by own needs for it, but rather motivated by the prosperity of earning money.

Although the service provided by the Bitcoin substrate is highly robust, it is also painstakingly slow. Commit latencies of transactions are often measured in hours. Applications that intend to use Bitcoin, or one of its derivatives, as a public cloud-like service for communication, can still improve their messaging performance by adjusting the offered transaction fee to the number of bytes sent. There are, however, clear limits to what can be achieved. For Bitcoin, spending more than 300 Satoshi per byte seems to be ineffective. As the incentives to mine new blocks shift focus from block fees to transaction fees in the years to come, we expect that new schemes for optimizing messaging performance needs to be developed.

ACKNOWLEDGMENT

This work was supported in part by the Norwegian Research Council project number 263248. We would like to thank the anonymous reviewers for their useful insights and comments.

REFERENCES

- Aiyer, A. S., Alvisi, L., Clement, A., Dahlin, M., Martin, J.-P., and Porth, C. (2005). Bar fault tolerance for cooperative services. In *Proceedings of the Twentieth ACM Symposium on Operating Systems Principles*, SOSP '05, pages 45–58, New York, NY, USA. ACM.
- Augspurger, T., Bartak, C., Cloud, P., Hayden, A., Hoyer, S., McKinney, W., Reback, J., She, C., Horikoshi, M., den Bossche, J. V., et al. (2012). Pandas: Python Data Analysis Library. software v0.21.0, Pandas community.
- Cheriton, D. R. and Zwaenepoel, W. (1985). Distributed process groups in the V kernel. *ACM Transactions on Computer Systems*, 3(2):77–107.
- Croman, K., Decker, C., Eyal, I., Gencer, A. E., Juels, A., Kosba, A., Miller, A., Saxena, P., Shi, E., Gün Sirer, E., Song, D., and Wattenhofer, R. (2016). On scaling decentralized Blockchains. In *Financial Cryptography and Data Security. FC 2016.*, volume 9604 of *LNCS*, pages 106–125, Berlin, Heidelberg. Springer Berlin Heidelberg.

- Houy, N. (2014). The economics of Bitcoin transaction fees. Working Papers 1407, Groupe d'Analyse et de Théorie Economique (GATE), Université Lyon 2.
- Hunter, J., Dale, D., Firing, E., Droettboom, M., et al. (2017). Matplotlib for data plotting. software v2.1.1.
- Johansen, H. D., van Renesse, R., Vigfusson, Y., and Johansen, D. (2015). Fireflies: A secure and scalable membership and gossip service. ACM Transactions on Computer Systems (TOCS), 33(2):5:1–5:32.
- Kihlstrom, K. P., Moser, L. E., and Melliar-Smith, P. M. (1998). The SecureRing protocols for securing group communication. In *Proc. of the 31st Annual Hawaii International Conference on System Sciences*, pages 317–326. IEEE.
- Moser, L. E., Melliar-Smith, P. M., Agarwal, D. A., Budhia, R. K., and Lingley-Papadopoulos, C. A. (1996). Totem: a fault-tolerant multicast group communication system. *Communications of the ACM*, 39(4):54– 63.
- Möser, M. and Böhme, R. (2015). Trends, tips, tolls: A longitudinal study of Bitcoin transaction fees. In *Financial Cryptography and Data Security: FC 2015.*, number 8976 in LNCS, pages 19–33, Berlin, Heidelberg. Springer Berlin Heidelberg.
- Rizun, P. R. (2015). A transaction fee market exists without a block size limit. Technical report.
- Taylor, M. B. (2017). The evolution of bitcoin hardware. *IEEE Computer*, 50(9):58–66.
- van der Laan, W. J., Wuille, P., Andresen, G., et al. (2017). Bitcoin client application. software v0.15.1.
- van Renesse, R., Birman, K. P., and Maffeis, S. (1996). Horus: a flexible group communication system. *Commu*nications of the ACM, 39(4):76–83.
- Waskom, M., Botvinnik, O., drewokane, Hobson, P., David, Halchenko, Y., Lukauskas, S., Cole, J. B., Warmenhoven, J., de Ruiter, J., Hoyer, S., Vanderplas, J., Villalba, S., Kunter, G., Quintero, E., Martin, M., Miles, A., Meyer, K., Augspurger, T., Yarkoni, T., Bachant, P., Williams, M., Evans, C., Fitzgerald, C., Brian, Wehner, D., Hitz, G., Ziegler, E., Qalieh, A., and Lee, A. (2016). Seaborn. Software v0.7.1.