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Emission Impossible: privacy-preserving carbon emissions claims

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1 INTRODUCTION

Customers of online services may want to take carbon emissions into account when deciding which service to use, but are currently hindered by a lack of reliable emissions data that is comparable across services. Calculating accurate carbon emissions across a cloud computing pipeline involves a number of stakeholders, none of whom are incentivised to accurately report their emissions for competitive reasons. In this paper we explore mechanisms to support verifiable and privacy-preserving emissions reporting across a chain of energy suppliers, cloud data centres, virtual machine hosting services providers and cloud services providers, which are ultimately passed through to APIs used by customers. We believe that adding verifiable and composable emissions transparency to cloud computing architectures enables providers to compete on the basis of sustainability, resulting in demand-side pressure on cloud services to shift to renewable energy sources [6].

Our technique centres around zero-knowledge proofs (ZKPs) [10]. When applying ZKPs to our problem, a stakeholder in a supply chain proves to a verifier (who can be anyone) that the carbon emissions calculations were made accurately, without revealing commercially sensitive data about their business operations. The verifier decides if the claim can be accepted by using only public knowledge and the proof provided by the stakeholder, achieving verification with strong privacy guarantees. We propose the ZKP system to be based on a zero-knowledge Succinct Non-interactive ARguments of Knowledge (zk-SNARK) protocol, which provides "good enough" efficiency for computation of emissions claims and rich enough functionality for zero-knowledge proofs [3, 4].

Figure 1 (overleaf) shows the challenge of tracking carbon emissions from a cloud computing supply chain, and how ZKP can be applied to provide verifiable emissions data without revealing trade secrets.

In this talk we will explore: (*i*) how the conflicting incentives around carbon emissions reporting make existing systems unlikely to succeed; and (*ii*) how we could use ZKPs to allow for the accurate reporting of carbon claims *without* compromising on individual privacy and competitiveness requirements.

2 INCENTIVES TO CHEAT FOR CO2E CLAIMS

It was estimated that 1.8% to 3.9% of the global carbon emissions are attributable to Information and Communication Technology (ICT) [1]. Governments, investors and consumers are therefore watching how cloud computing operators are working towards net zero [2, 9, 16]. However, businesses have strong incentives to make only positive claims, which could involve hiding data or publishing misleading results with dubious evidence, a problem termed "greenwashing" [13]. Moreover, they are keen not to expose any private business data to preserve their competitive advantage. So we end up with the current state that many claims are disclosed to the public but it is difficult to know if any of the claims are true.

Consider three of the biggest data centre providers: Amazon, Google and Microsoft. They have all reported their emissions goals publicly, but have also used creative accounting to hide facts about their carbon emissions [18]. Both Microsoft and Google admitted that their carbon emissions had increased in recent years, despite their climate commitments [12, 15], and Amazon's self-reporting did not include the emissions data of the products sold by thirdparty vendors [7]. The major risks involved in data centre emissions reporting are:

- (1) **Privacy and Trade secrets concerns.** Reports on carbon emissions typically show only high-level aggregated data. Validation of the claims often requires details of carbon activities, which could potentially expose trade secrets through disclosure.
- (2) Untrustworthy claims. Companies make misleading claims based on dubious accounting methodologies to make it look like they are more environmentally friendly than they actually are [13]. Yang et al. studied greenwashing behaviours and impact and found that greenwashing often links with scandals that happen at the supply chain level [19].
- (3) **Missing claims.** Companies can choose not to disclose anything, or report claims that omit some of their emissions-generating activities. Amazon's undercounting on their carbon footprint reports is a good example [7].

We can mitigate the first risk to protect businesses by ensuring that the verification method will not leak secrets. For the second risk we can use verified data to provide trustworthy claims. ZKPs can be used for these mitigations and provide a mechanism to standardise accounting methodologies. For the third risk, we cannot validate data that is missing. We can, however, bind carbon accounting to financial accounting to make it more difficult to cheat. For example, a ZKP can show that the total paid to all suppliers matches the number that is reported on the audited accounts, and that these supplier transactions are also reflected in the emissions calculation. Finally, we can analyse and benchmark the verified data across companies and look for discrepancies and anomalies through manual audits.

3 A ZKP EMISSIONS DISCLOSURE SCHEME

A zk-SNARK allows a prover to convince a verifier that the prover knows values (a *witness*) that satisfy a given set of equations (a *circuit*) without revealing any information about the witness [4, 5, 14]. Similarly to a digital signature, SNARKs do not require any interaction between the prover and verifier besides sending the proof. Proofs are short in length and can be verified quickly – in some cases, in constant time. SNARKs are actively used in real world applications such as the Zcash cryptocurrency, where transactions are executed without any sensitive information such as the origin or amount of the transaction being revealed [11].

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Figure 1: Unverified emissions claims can affect sustainability goals negatively (*left*) but ZKP claims can be verified directly by consumers to make informed choices (*right*) via transparent emissions declarations across the cloud stakeholder chain.

To illustrate, consider a scenario where a user wants to compare the carbon emissions of AI chatbots such as ChatGPT (OpenAI), Gemini (Google) and Claude (Anthropic). To ensure a fair comparison, they must be computed according to the same methodology, and each emissions claim must be verified end-to-end. One aspect of such a claim would be to verify that the disclosed electricity consumption figures come from a trusted source.

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First we assume that the data centre operator uses smart meters to track power consumption, and that the smart meters have a secure hardware element that signs meter readings with a private key configured by the meter manufacturer. Then we apply a zk-SNARK in three stages:

Stage 1: Generate circuit C. The circuit encodes the accounting 149 methodology used to calculate the emissions. The constraints and 150 relationships among the input parameters are reduced to a set of 151 polynomial equations. Cryptography is applied such that it is not feasible for anyone to calculate the input data by knowing the 153 output proof generated by the circuit, hence protecting the secrets. 154 There are many ways to calculate emissions and the accounting 155 is non-trivial because of many variables. For example, to calculate 156 the power consumption of a ChatGPT session, we cannot simply 157 divide the total consumption by the number of users, as different 158 users may have very different usage levels. We also need to account 159 for fixed costs (such as model training) and waste (such as models 160 161 that were trained but never used because they performed poorly). There are emerging carbon accounting methods (e.g. PACT [8]) and 162 methodologies for attributing load to individual customers within 163 a data centre [17], which could inform circuit construction. 164

Stage 2: Prove. The prover proves that the emissions were cal-166 culated using only trusted and accurate data, such as emissions 167 figures, smart meter readings, carbon intensity, GPU power con-168 sumption per tenant, AI token metrics and many more. Some of 169 these parameters are only known to the prover. To prove that the 170 smart meter readings came from trusted smart meters, the public 171 172 keys used by the smart meters can be revealed as public data used 173 to generate the proofs.

Stage 3: Verify. The verifier uses the proof and public data to determine if they can believe that the prover has the knowledge of the private data, such that when both the private and public data are used along with the emissions figures, they satisfy all the equations encoded in the circuit. That is,

\exists private witness. $C(\{public, private\} witness) = True$

In our example, the verifier would only believe a smart meter reading came from a trusted smart meter if the reading has been signed by a private key and the signature can be verified using the public key available to them, without knowing the private key.

4 UNLOCKING EMISSIONS IMPOSSIBLE

It seems almost impossible to balance privacy and competitiveness needs with our urgent sustainability goals to reduce emissions, particularly in ICT sector. The use case outlined in this paper is an example of how we can achieve privacy-preserving and trustworthy carbon emissions claims for data centres. The approach can be used in other industries as well. To adopt the ZKP system companies can apply carbon accounting alongside their financial accounting, which already needs to attribute use of computing resources to individual customers. A benefit of using a SNARK is that the proofs are small enough to be bundled along with emissions reporting.

We believe that this proposal is a step forward for carbon emissions accounting to be public and explicit, making emissions tracking more accurate and comparable across companies. We are exploring how this could be exposed directly to end users via browser plugins, providing an end-to-end verifiable CO2e cost alongside conventional costs used by users to make their buying decisions (such as price, delivery time, or distance to the service). Our overall aim is to drive demand-side pressure to reduce unnecessary emissions from data centre use by informing consumers about the environmental cost of their actions online.

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