Context-Aware Security for 6G Wireless
The Role of Physical Layer Security

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Abstract—Sixth generation systems are expected to face new security challenges, while opening up new frontiers towards context awareness in the wireless edge. The workhorse behind this projected technological leap will be a whole new set of sensing capabilities predicted for 6G devices, in addition to edge and device embedded intelligence. The combination of these enhanced traits can give rise to a new breed of adaptive and context-aware security protocols, following the quality of security (QoSec) paradigm. In this framework, physical layer security solutions emerge as competitive candidates for low complexity, low-delay and low-footprint, adaptive, flexible and context aware security schemes, leveraging the physical layer and introducing security controls across all layers, for the first time.

I. INTRODUCTION

An intense discussion is currently underway with respect to the resilience and trustworthiness of the 6G radio, pivoting the enhancement of the security of the envisioned sixth generation (6G) wireless access. Notably, some of the recent, increasingly sophisticated attacks on the wireless edge, e.g., jamming or false base stations, can be implemented with a price tag as low as 1k$ using low-cost software defined radios. In addition, we experience an expansion of the attack surface with artificial intelligence (AI) and machine learning (ML) tools. In parallel, as we move gradually away from the standard client-server networking paradigm and enter a new era of truly end-to-end (E2E) quality of service (QoS), service level agreements (SLAs) in the near future will be expected to include guarantees about the quality of security (QoSec) as well. The definition of ingredients of QoSec is currently being investigated: how to identify the security level required and to propose adaptive, dynamic and risk aware security solutions.

Meanwhile, the evolution towards 6G systems is expected to introduce new means of reaching situational awareness by harvesting and interpreting the “context” of the communication, including, network tomography; nodes’ constraints, the age of information, etc. Incorporating context awareness in QoSec is projected to allow handling more efficiently aspects related to identifying the risk or threat level and the required security level, particularly for applications with non-functional security requirements, such as autonomous vehicles, platooning, eHealth, etc. In this framework, incorporating security controls from the palette of physical layer security (PLS) can be particularly attractive due to their low computational complexity (relevant implementations are based on standard encoders) and their inherent ability to adapt to the transmission medium properties. The incorporation of PLS in 6G security requires indeed enhanced context awareness and can be particularly attractive for massive machine type communications (mMTC) and ultra-low latency use cases.

In the rest of this article, we will begin in Section II with a review of open security issues in 5G and research challenges ahead of 6G and move on to presenting a roadmap to address these challenges in Section III. To illustrate some of the proposed ideas we outline viable solutions to address specific security vulnerabilities in 5G and 6G, along with a discussion of possible further directions in Sections IV and V, while conclusions are drawn in Section VI.

II. OPEN 5G SECURITY ISSUES AND SECURITY RESEARCH CHALLENGES AHEAD OF 6G

Despite the strengthening of 5G security protocols with respect to previous generations, there are still open issues that have not yet been fully addressed, e.g., attacks under the generic umbrella of “false base stations”. In parallel, in the path towards the 6G evolution, new security challenges arise as a result of drastic changes in key operation parameters: i) the E2E latency tolerance; ii) the sheer scale of networks in mMTC use cases and very large scale Internet of things (IoT); iii) the long lifespan of deployed IoT devices (notably sensors) that will need to be secured; iv) the wide variety of heterogeneous RF technologies involved; and v) the accelerated steps taken towards bringing quantum computers to life. In the following, we provide a short review of open security issues in 5G and of some of the security challenges in the evolution towards 6G. This discussion provides the motivation for our proposal of context-aware security solutions for future
generations of wireless, which will also be able to leverage the physical layer to provide flexible and adaptive security guarantees.

A. False Base Station Attacks

The expression “false base stations” (FBS) describes impersonation attacks of genuine base stations. The topic is currently studied by the SA3 working group, documented in TR 33.809 [1]. Typically in 5G an FBS is a “man-in-the-middle” (MitM) or a very stealthy jammer. A major vulnerability highlighted by FBSs is that the phases of entry into the network, which precede the enactment of the 5G security protocols, are particularly critical for many of the attacks described in TR 33.809. For example, attacks consisting in replaying modified versions of the broadcast channels can have disastrous consequences on all the terminals of a cell, hindering their connection to the network or forcing them to operate in a degraded mode. As a result, it is necessary to propose methods that allow the user equipment (UE) to determine whether a BS is legitimate, prior to exchanging unauthenticated messages. To this end, PLS could be used by incorporating the BS localization by a UE as a soft authentication factor.

B. Security Challenges in Ultra Reliable and Low Latency Communications (URLLC)

Critical ultra reliable low latency communications (URLLC) are typically used for industrial IoT (IIoT), vehicle-to-everything, and other applications requiring low latency and very high reliability. To achieve high reliability, a possible avenue is by increasing diversity, e.g., multiple parallel transmissions can be exploited. However, this consequently increases the “attack surface”, while it might also impose more stringent constraints in terms of the speed of integrity checks. Overly aggressive latency targets could entail a new security architecture altogether. State-of-the-art proposals for fast authentication with use of implicit certificates or certificateless solutions can speed up authentication. Many open challenges for sub-millisecond delay constrained URLLC systems remain, with respect not only to authentication, but as well for the integrity and the confidentiality of both the control and data planes, as documented in [2].

C. Jamming Attacks in mMIMO — RF Resilience

Although multiple input multiple output (MIMO) systems, including massive MIMO (mMIMO), make eavesdropping more difficult thanks to energy focusing, they nevertheless also introduce vulnerability points. Indeed, beamforming in mMIMO systems relies on accurate channel estimation. Pilots are transmitted in order to obtain the channel state information (CSI), which in turn allows precoding. If the CSI is not correctly estimated (e.g., because of interference or due to voluntary contamination by a jammer) the precoder will disperse the power, resulting in potential leakage and poor link quality. The later leads to service unavailability, giving rise to a denial of service (DoS) type of attack, as described in [3]. Similar attacks can also be launched at the medium access control (MAC) by tampering with the CSI reports sent by the devices. As a result, the beam management phase during network entry is vulnerable to RF jamming attacks. It is therefore crucial to have the means to detect, locate and neutralize jammers, or implement mitigation solutions.

D. Privacy

Although 5G incorporates a set of measures to enhance privacy in terms of user identity (subscription) privacy, recent research on user location privacy and user untraceability has shown that there are still many open issues, while the privacy guarantees are rather weak from an end-user perspective. The amount of personal data handled by future mobile networks will substantially increase Governmental agencies as well as adversarial entities have potentially a high interest in such data; future wireless networks have to be designed to ensure privacy without having to place trust in operators.

E. Post-Quantum Resilience

A further challenge comes from quantum computing, which has seen significant progress after massive earlier investments. Since some of the most important cryptographic algorithms used in 5G are not quantum-resistant, the related protocols have to be redesigned involving post-quantum crypto algorithms. The national institute of standardization (NIST) is currently evaluating novel post-quantum crypto algorithms to replace currently used public key encryption schemes. Nevertheless, it is a common concern that quantum resistance will lead, at least in the immediate future, to an increase in terms of the complexity of the new cryptographic systems. For example bigger key sizes might pose a significant problem in practise. This could be especially challenging for URLLC and low-power / low-cost devices, further highlighting conflicting trends in future systems and the interplay between computational based crypto and real-time communication between low-end devices.

F. IoT security

There are numerous security issues arising with the introduction of very large scale, long life, constrained IoT networks. Low-end, SIMless IoT devices, are unlikely to be able to support advanced security mechanisms, due to computing power, memory and – probably most challenging – energy consumption constraints. Although lightweight cryptography could help to address some of the challenges, such algorithms are currently not part of 5G and the development of lightweight post-quantum solutions is a recent field of research. Furthermore, the envisioned huge number (trillions) of very diverse IoT devices connected to the B5G network brings about big challenges in terms of information security management, but also is itself a security risk, as shown by the 2016 Mirai attack with a sever overall impact. In this aspect, decentralised intrusion / anomaly detection becomes important [4].

Another factor at play is that many IoT devices will typically have a very long lifespan (>10 years as opposed to 3 years for
Secret key generation

**Step 1: Advantage distillation:** Alice and Bob exploit the reciprocity of the wireless channel to extract shared randomness.

**Step 2: Information reconciliation:** Alice sends her syndrome to Bob, so he can correct discrepancies of his observation.

**Step 3: Privacy amplification:** Alice and Bob obtain a shared secret key $k$.

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**Figure 1.** Distilling symmetric keys from wireless coefficients $h_{AB}$ in multipath channels, exploiting channel reciprocity during the channel’s coherence time. The procedure comprises three phases, referred to as advantage distillation, information reconciliation and privacy amplification.

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a laptop) and can be distributed in large geographical areas. It is difficult to guarantee that mass-produced, computationally and power constrained IoT devices will have a hardware capable of being updated with the necessary patches to resist all the threats that will arise in their lifetimes (e.g., post-quantum resistance).

III. 6G AS AN ENABLER TO CONTEXT AWARE QoSEC LEVERAGING PLS

Even though 6G is still some years away from standardization, consensus is growing on its likely evolution path, briefly outlined in the following:

- **Higher frequencies and bandwidth:** Continuing the evolution seen in the previous generations, 6G will make use of ever higher carrier frequencies and bandwidth, moving towards frequencies above 100 GHz, which allows the allocation of bandwidths larger than 1 GHz. The large bandwidth may increase the observable channel entropy in the frequency domain, which can potentially be exploited in PLS secret key generation (SKG) from wireless coefficients [5], whose principal mechanisms are depicted in Fig. 1.

  Additionally, in mmWave and SubTHz systems beamforming with pencil-sharp beams becomes both a possibility, because of the smaller area occupied by antenna arrays, and, a necessity, because of the need to compensate for the higher channel attenuation. Highly directive beamforming can then reduce eavesdropping opportunities, as depicted in Fig. 2, while similar opportunities exist for visible light communications (VLC).

  Thus, in 6G, massive MIMO (mMIMO) could offer a viable application scenario for the wiretap channel.

- **Integrated sensing and communications:** In addition to high-resolution image, video and sound, among other
possible sensing data, which can be transmitted through mobile communication networks, radar sensing is likely to be an integral part of future wireless systems [6], reusing the same spectrum and waveform as communications. These new capabilities along with centimetre-level localization precision will allow the network to have a better understanding of the surroundings and gain situational awareness, i.e., understanding of the context of communication. On the other hand, this raises other security issues, as the sensing data themselves may be subject to tampering by attackers. Their integrity must be assured. As a result, trustworthiness of sensing and communications is expected to be a key performance indicator of 6G systems.

- **Learning at the wireless edge and native AI**: Centralized machine learning, which processes data centrally using cloud-based computing, is subject to critical security challenges, e.g., a single point of failure and the vulnerability of data during backhaul. Moreover, it might not be suitable for real-time applications, due to the capacity and latency requirements resulting from centralized data aggregation and processing. Thus, decentralized ML solutions are becoming increasingly important, e.g., federated learning, in which data are in principle processed locally at end-user devices where they are collected. While such distributed ML solutions can serve as enabling technologies for 6G mobile edge networks, they also introduce vulnerabilities such as the leakage of private information through learned model parameters, exposure to malicious end-user devices and adversarial training examples.

These anticipated 6G features provide novel opportunities to address the security and privacy challenges outlined in Section [1], allowing for the security architecture of 6G networks to be built around automation. Following the principles of multilateral security, the system should understand the security goals of the entities involved and should adapt the security controls accordingly based on contextual information, harvested from the novel 6G features. To this end, we need a set of building blocks:

i) Quantify security in the QoSec framework, i.e., the ability to express the desired and actual “level of security”;  
ii) Context awareness at the wireless edge with the aid of sensing and AI;  
iii) New, adaptive security controls, incorporating PLS;  
iv) Automation in the form of a ML/AI based security orchestrator.

In the following subsections we discuss in further detail some of these necessary building blocks.

### A. Quantifying Security: Quality of Security (QoSec)

Similar to QoS definitions (e.g., [ITU-T E.800]), QoSec is the totality of characteristics of a service that bear on its ability to satisfy stated and implied security needs of the user. QoSec is able to provide different security guarantees, in response to the security needs of different use cases and related slices of the network, reflecting on the DiffServ QoS paradigm. A central aspect related to QoSec is to identify how to make the security level and its implementation adaptive: how to automatically identify the right QoSec and the right combination of crypto schemes (encryption, integrity, authentication primitives), as well as how to incorporate these flexibly in security protocols.

Thereby, adaptivity can happen at different levels: for a fixed cryptographic strength (e.g., 256-bit symmetric block ciphers considering quantum-resistant) and a fixed attacker model (e.g., “zero trust”, i.e., minimal (trust) assumptions regarding all involved entities) we can adapt the specific cryptographic algorithms and protocols that are used [7]. On the other hand, we could also adapt the desired cryptographic strength or the considered attacker model based on contextual information. In future security protocols varying levels of trustworthiness (e.g., as defined by NIST in SP800-53 Rev. 4) are envisioned through the use of security control baselines. Note that these are developed based on a number of general assumptions, including common environmental, operational and functional considerations, giving rise to the question of context awareness in security. In Section IV we discuss in detail how PLS can be leveraged to develop adaptive security controls.

### B. Context Awareness at the Wireless Edge: The Role of Sensing and AI

The opening of the THz spectrum will provide new “sensing” capabilities to 6G devices, such as high-definition imaging and frequency spectroscopy. Unique opportunities arise for reaching context awareness through the processing of sensing information with both centralized and edge AI; in turn, context awareness is key for trust building and for predicting reliability, i.e., QoS can be driven by context awareness. Incorporating
context awareness in security controls amounts to being able to provide answers – with the aid of AI – to the following open-ended questions:

1 How to extrapolate the threat level from context: PHY layer inputs, particularly in the form of sensing information including the location of a node, the time of communication, the ambient temperature, etc., carry important contextual information, directly related to semantics. We can envision AI multi-modal fusion of sensing information to obtain an enhanced evaluation of the threat level. In very demanding scenarios such as platooning, this approach might help provide a viable route to develop anomaly detection solutions for highly dynamic, seemingly chaotic, networks.

2 How to use context to identify the security level required: We need to take steps towards defining new metrics describing the criticality of the particular data exchanged and furthermore, how valuable they are considered from an adversarial point of view. This can be thought of as the analogous of defining the priority level in QoS.

3 How to match security levels to security schemes: After defining the security level with rapport to the context of communication, the next question is how to map this to an actual set of algorithms and security schemes. Two approaches emerge that can possibly be used jointly: i) crypto based approaches, in which the strength of crypto systems is, roughly speaking, related to the lengths of the keys (after the right transformations are accounted for); ii) PLS approaches, in which the wireless channel and the hardware are used as sources of uniqueness (for authentication) and / or entropy for confidentiality purposes (e.g., for SKG) [7]. Next, we delve into the potential use of PLS in 6G and discuss how PLS is inherently adaptive and can be enabled by context awareness.

IV. QoSec Adaptive Security Controls: The Role of Physical Layer Security in 6G

In the past years, PLS [6, 9] has been studied and indicated as a possible way to emancipate networks from classic, complexity based, security approaches [10]. PLS is based on the premise that we can complement some of the core security functions, exploiting both the communication radio channel and the hardware as sources of uniqueness or of entropy.

It is usually this latter aspect of PLS that is considered in the literature, around the concept of the secrecy capacity and of the SKG capacity [11]. In this framework, PLS leverages the physical properties of the radio channel, namely diffusion, superposition and reciprocity, to create opportunities for secure data transmission in the presence of eavesdroppers in the channel. These properties can be exploited in a variety of ways, including taking advantage of independent fading between legitimate users and eavesdroppers, the use of multiple-antennas or relays and the injection of artificial noise to create secure degrees of freedom.

In the celebrated wiretap channel model introduced by Wyner in 1975, the adversarial link is degraded with respect to the main link, i.e., legitimate users do not share a secret but have a link quality advantage; whenever this can be substantiated, the existence of wiretap codes that can ensure asymptotically both reliability in the reception of a confidential message by a legitimate receiver and negligible information leakage to an eavesdropper has been demonstrated. Furthermore, by adjusting network / system parameters, different secrecy outage probabilities – potentially corresponding to different QoSec levels – can be attained. We illustrate the underlying ideas in uses cases in which the wiretap channel is used to convey securely symmetric secret keys in hybrid PLS-crypto systems. In this case, very low secrecy rates can be targeted as a single key of 256 bits can be used to encrypt up to Gigabytes of data, e.g., when wiretap coding is used jointly with modern ciphers such as the advanced encryption algorithm (AES) in Galois counter mode (GCM), then negligible secrecy rates in the order of $10^{-7}$ could be sufficient. Under this assumption, we illustrate system design parameters to achieve positive secrecy rates in two scenarios: i) first we evaluate the minimum number of antennas at a BS in Table I for MISO channels using the results in [12]; ii) secondly, in Table II the maximum eavesdropper density is evaluated for unmanned aerial vehicle (UAV) networks based on [13].

In Table I, notice the critical role of the relative quality of the adversarial versus the legitimate link, captured in ratio of the corresponding large scale fading coefficients, denoted by $\alpha_e$ and $\alpha_s$ respectively. When the legitimate user is much closer to the BS than the adversary, a secrecy outage probability as low as $10^{-10}$ can be attained with the use of only 21 antennas. On the other hand, due to line-of-sight in UAV communications, the maximum eavesdropper density for the same secrecy outage probability is $\lambda_e = 10^{-8}$ when the legitimate node density is $\lambda_l = 10^{-3}$ and the UAV is 10 m above ground. Only by reducing the secrecy outage probability, i.e., by reducing the target QoSec level, can the maximum eavesdropper density be increased. These two examples demonstrate that context awareness is necessary for the correct employment of PLS; in the MISO setting, proximity to the access point is critical, in the UAV example node density plays a major role. These two examples further show that in a given context, PLS can be used to achieve potentially a subset of QoSec levels, articulated

| $P_{e|o_0}$ | $P_{e|o_1}$ | $P_{o_0}$ | $P_{e|o_0}$ | $P_{e|o_1}$ |
|-------------|-------------|-----------|-------------|-------------|
| $\alpha_e/\alpha_s = 1/4$ | 1 | 10 | 7 | 21 |
| $\alpha_e/\alpha_s = 1/2$ | 2 | 10 | 7 | 21 |
| $\alpha_e/\alpha_s = 1$ | 9 | 952 | 7 | 21 |

Table II. Maximum eavesdropper density $\lambda_e$ to achieve a target secrecy outage probability $P_{e|o_0}$ in an UAV network using eq. (17) in [13]. $\lambda_l$ denotes the density of legitimate nodes, UAV at a height of $H = 10 \text{ m}$.

| $P_{e|o_0}$ | $P_{e|o_1}$ | $P_{o_0}$ | $P_{e|o_0}$ | $P_{e|o_1}$ |
|-------------|-------------|-----------|-------------|-------------|
| $\lambda_e = 10^{-8}$ | 1 | $10^{-8}$ | $10^{-8}$ | $10^{-8}$ |
| $\lambda_e = 10^{-2}$ | 1 | $10^{-2}$ | $10^{-2}$ | $10^{-2}$ |
The ability of ML techniques to learn and capture statistics of complex features, we can achieve low-cost, continuous, highly reliable, model-independent, and context-aware authentication, e.g., leveraging localization and RF fingerprinting. To enhance the reliability of such authentication mechanisms, the trustworthiness of the observed and estimated attributes needs to be monitored, accounting for context.

Finally, in terms of device authentication, it is further possible to leverage “hardware fingerprints” in the form of physical unclonable functions (PUFs), as an authentication factor in multi-factor authentication protocols. PUFs rely on the use of Wyner-Ziv reconciliation approaches to offer measurable re-usability of the hardware fingerprint. Combining various PLS technologies, hybrid PLS-crypto systems can be built around the ideas of zero-round trip time (0-RTT) protocols and / or authenticated encryption [13], offering further tools to develop fast authentication schemes at PHY, potentially exploiting multiple authentication factors.

V. DISCUSSION AND PROPOSED ROADMAP
Looking at the broader picture, down the path towards 6G, novel security challenges and opportunities arise. Among the challenges, noteworthy are issues related to vulnerabilities in the initial entry phases of a node in a network (before the enactment of the 5G security protocols), the massive number of low-end and heterogeneous IoT devices, sub-millisecond delay constraints for critical IoT use cases, etc., while offering post-quantum security guarantees and addressing issues of privacy. On the other hand, 6G is expected to be the first generation of wireless to offer edge- and device-level intelligence, leveraging novel sensing capabilities and the extensive use of ML. The incorporation of context awareness in 6G security protocols can propel the introduction of disruptive new technologies to provide flexible and adaptive security guarantees, based on an on-line evaluation of the security threat level.

It is in this context that PLS technologies can be truly exploited; PLS can be realised only with provably trustworthy monitoring and understanding of the communication environment and communication medium in 6G. In applications such as the IoT, PLS emerges as a very competitive candidate to be used in context-aware, flexible and adaptive security controls, both for authentication as well as for confidentiality schemes. While PLS might not, at least in the near future, be incorporated in zero-trust security protocols, it does provide a viable alternative to securing massive and ultra-low latency networks with relaxed security guarantees, as a competitive candidate for emerging QoSec approaches that will cut across all layers of the network stack.

PLS offers notable advantages. Firstly, it is inherently adaptive: by adjusting the target secrecy rate or secret key rate, one can adapt related secrecy outage probabilities, offering a flexible framework with respect to adaptive security controls. Furthermore, PLS can provide information-theoretic security guarantees using lightweight mechanisms (e.g., using Polar or low density parity check (LDPC) encoders) as opposed to computationally expensive cryptographic schemes. Thus, such approaches are suitable for low complexity IoT devices and
Table III. **ROADMAP OF SOLUTIONS FOR 5G / 6G SECURITY CHALLENGES**

| Security Challenge / Scenario                  | Recommended techniques (with * we denote PLS / PHY solutions)                                                                 |
|------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------|
| False Base Station Attacks                     | * Intelligent PHY authentication using RF fingerprinting and localization of BS from UE (inverse localization)                 |
|                                                | * Pre-shared keys established / distributed with SKG                                                                       |
| Low Latency Communications                     | * Fast authentication using PUFs and RF fingerprinting as early authentication factors                                         |
|                                                | * Short packet secrecy encoding                                                                                             |
|                                                | * Short blocklength Slepian Wolf and Wyner Ziv reconciliation decoders (for SKG and PUFs)                                    |
| Jamming Attacks in mMIMO — RF Resilience       | * Spectrum sensing, channel charting, channel learning                                                                     |
|                                                | * Advanced modulation and coding                                                                                            |
|                                                | * Intrusion detection at PHY                                                                                               |
|                                                | * Covert communications / low probability of detection                                                                   |
| Privacy                                        | - Context aware choice of pseudonymity, partial identities                                                               |
|                                                | - Contextual aware integrity to detect and mitigate violations                                                             |
|                                                | - Context aware appropriateness and distribution                                                                           |
| Post-Quantum Resilience                        | * PLS is information theoretic secure                                                                                      |
|                                                | * Long symmetric encryption keys using channel-based key generation                                                        |
|                                                | * Hybrid crypto-PLS schemes                                                                                               |
| Low-cost IoT devices                           | * PLS is lightweight, secrecy encoders, SKG, PUFs, etc.                                                                     |
|                                                | Awareness of low-cost / low-security IoT devices for appropriate isolation in a dedicated network slice                    |
| Huge Number of IoT devices                     | - Contextual understanding to automatically select appropriated QoSec                                                        |
|                                                | - Adaptive and automatic security controls removing the burden to manually configure and monitor all the IoT devices       |
|                                                | * PLS as a scalable technique for key management and distribution                                                          |
|                                                | * PLS as adaptive security scheme                                                                                        |
| Long-term IoT security                         | - Awareness of a decrease over time in QoSec and trustworthiness                                                           |
|                                                | - Automatic adoption of the overall security controls and policies                                                         |
|                                                | - Context aware access control, e.g., excluding untrustworthy devices from the network or reduction of (access) rights       |

for networks with light or no infrastructure, either as stand-alone best-effort security mechanisms or as complements to more traditional methods.

To exemplify some of the points made previously, in Table III we present a roadmap on how to address the security challenges listed in Section II and how PLS fits into this picture. We want to emphasize that the presented ideas are still just parts of the puzzle and have to be embedded in a much more holistic approach, which, besides additional technical means, has to incorporate organisational, regulatory, economical – and not to forget, standardisation – aspects.

VI. **CONCLUSIONS**

Unarguably, 5G security enhancements present a big improvement with respect to LTE. However, as the complexity of the application scenarios increases with the introduction of novel use cases, notably URLLC and mMTC, novel security challenges arise that might be difficult to address using the standard paradigm of complexity based classical cryptographic solutions. At the same time, in the longer 10-year horizon novel security concepts based on “trust models” and risk-based, adaptive identity management and access control will come to life, enabled to a large extent by AI. To allow for flexible QoSec, the development and integration of security controls at all layers of the communications system is envisioned.

In this framework, PLS is being considered as a possible way to emancipate networks from classical, complexity based, security approaches. With respect to authentication, PUFs, wireless fingerprinting / localization, combined with more classical approaches, could also enhance authentication and key agreement (AKA) in demanding scenarios. In parallel, THz communications will rely upon setting up highly directional beams, potentially providing a concrete scenario for the wiretap channel. Furthermore, with the opening up of higher frequency bands in 6G, the opportunity to harness entropy in the frequency domain can be exploited in SKG protocols. As a general direction, context awareness, enabled by enhanced sensing and AI capabilities anticipated in 6G, can allow introducing disruptive tools for providing adaptive QoSec based security guarantees, tailored to the context of the communication, incorporating PLS security controls.

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