

Distributed approaches for signal processing and machine learning tasks in 5G/6G wireless communications

Research group of the Signal Processing and Communications Laboratory (SPCLab)

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Outline

- Introduction of SPCLab
- Activities at the PHY/Access of wireless systems
- Activities for SP&ML learning tasks
- Concluding remarks

Introduction of SPCLab (1/3)

- The Signal Processing and Communications Lab
 - Established in 1998 at CEID of UPatras
 - Research and education activities covering the general areas of Signal Processing, Wireless Communications and Machine Learning
- The team of SPCLab consists of
 - 4 Faculty members
 - 4 Collaborating faculty members
 - 4 Research fellows
 - 12 PhD candidates
 - About 15 post-graduate students per year
 - About 30 under-graduate students per year

Introduction of SPCLab (2/3)

- SPCLab has been involved in numerous national, European and bilateral R&D projects
 - In recent years, SPCLab handled over 10MEuro via multiple roles (including coordination and WP/Task leading)
- Research areas
 - <u>Signal processing and learning in Communications</u> (e.g., interference management, MIMO, channel estimation and equalization, multicarrier systems -OFDM, FBMC-, adaptive beamforming, cooperative comm, cognitive radio networks)
 - <u>Signal processing and learning over WSNs</u> (e.g., localization, distributed source coding, prediction mechanisms, node placement, distributed (adaptive) detection, estimation and statistical learning)

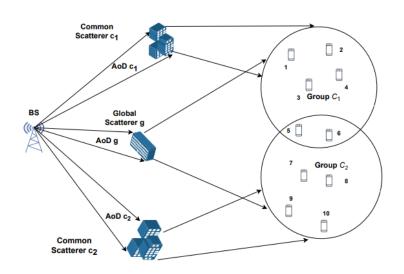
Introduction of SPCLab (2/3)

- Research areas (continued)
 - <u>Signal Processing & Learning: Algorithms & Theoretical</u>
 <u>Studies</u> (e.g., Array signal processing / DOA / BSS,
 Compressed sensing techniques, Distributed (adaptive)
 detection and estimation, dictionary learning)
 - <u>Machine Learning / Computer Vision / Image Analysis</u> (e.g., registration, super-resolution, motion estimation, object reconstruction and detection, hyperspectral imaging)
 - <u>Biomedical Signal Processing</u> (e.g., EEG-based biometrics, EEG-based emotion recognition, automated diagnosis)
- Application areas
 - PHY / Access COM layers, smart-grid, forestry safety, autonomous vehicles, robotics, non-distractive health monitoring, sign language translation, et al

Current activities concerning the PHY/Access layers of modern wireless communications

Fully distributed channel estimation

- In machine-type communications, channel estimation raises training overheads (due to massive MIMO in mmWaves) and scalability issues (high device density per area)
- An efficient, fully distributed approach has been devised when many devices and multiple antennas are involved



 An adaptive voting mechanism allows the devices to collaborate and estimate their channels by exploiting underlying (even partially) overlapping sparsity patterns manifesting in mmWaves

¹M. Trigka, C. Mavrokefalidis, K. Berberidids, "Efficient Distributed Multi-task Schemes for mmWave MIMO Channel Estimation", submitted to IEEE Access

²-----, "A Distributed Sparse Channel Estimation Technique for mm Wave Massive MIMO Systems", in proc. of EUSIPCO'21

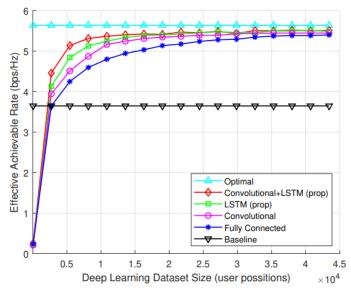
UAVs comm: Channel estimation in view of the Beam Squint effect

- Using large bandwidths in mmWave/THz bands and massive MIMO lead to the so-called beam squint effect
 - There are measurable propagation delays manifesting along the large antenna arrays
 - The steering vectors become frequency dependent, thus, current channel estimation algorithms are not applicable
- We have considered the problem of channel estimation for UAV-based mmWave massive MIMO communications with beam squint
- An efficient (computationally- and performance-wise) algorithm has been devised by employing the alternating method of multipliers
 - The algorithm estimates not only the channel gains and AoA but also the doppler shifts due to the UAV mobility

¹E. Vlachos, C. Mavrokefalidis, K. Berberidids, "Channel Estimation for UAV-based mmWave Massive MIMO Communications with Beam Squint", to be presented in EUSIPCO'22

Deep-learning-aided coordinated beamforming

- The multi-cell, coordinated beamforming problem towards a user has been simplified recently via deep learning
- Once a model has been trained, beamforming vectors can be selected very fast, but training requires many training data

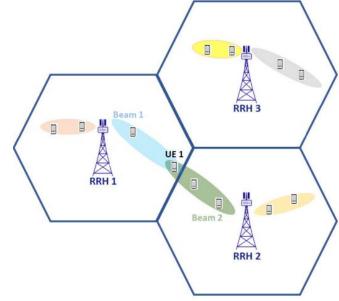


Aiming for fast training phases (with less required training data), novel CNN and LSTM-based models have been designed that exploit inherent time and frequency correlations of the transmitted OFDM signals (96% of peak performance achieved with only 18% of training data for the best model - red line)

¹I. Nikas, C. Mavrokefalidis, K. Berberidids, "Efficient Deep Model Training for Coordinated Beam-Forming in mmWave Communications", to be presented in IWSSIP'22

Cell-edge throughput improvement via JT-CoMP in small cells

- To support many users in modern wireless systems, existing and new technologies need to coexist and share limited resources
- Assuming a C-RAN with small cells, clustering of BSs in JT-COMP employing NOMA and beamforming is considered



 A coalition formation game is formulated for clustering the BSs and the solution, having linear complexity, aims at improving cell-edge user throughput without negatively impacting the remaining users

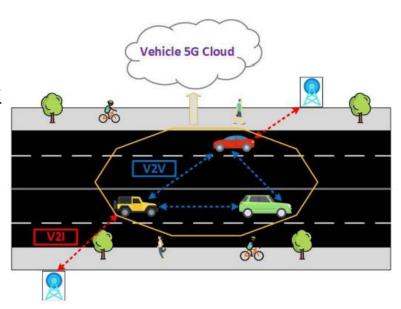
¹P Georgakopoulos et al., "Coalition Formation Games for Improved Cell-Edge User Service in Downlink NOMA and MU-MIMO Small Cell Systems", IEEE Access, 2021

²T Akhtar et al., "Efficient radio resource management with coalition games using NOMA in small cell networks", in proc. of IEEE Globecom'20

Current activities concerning SP&ML learning tasks

Distributed Localization in CAVs

- Localization in CAVs, i.e., accurate knowledge of self and other's road users/vehicles locations, is a vital task
- The accuracy of GPS sensors is highly degraded in dense urban canyons or tunnels
- Cooperative localization has received extensive interest for coping with GPS limitations

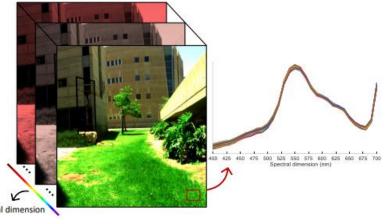


- Novel distributed multi-modal localization approaches for CAVs employing diffusion approaches have been devised
- Each agent estimates accurately its position and the position of the neighboring vehicles enabling more accurate 4D situational awareness

¹N. Piperigkos, A. S. Lalos, K. Berberidis, "Graph Laplacian Diffusion Localization of Connected and Automated Vehicles", IEEE Transactions on Intelligent Transportation Systems, 2021

Multi-dimensional data modeling and processing

Effective modeling of multidimensional visual data (e.g., hyper-spectral, SAR imaging, FMRI, geophysical signals) enables denoising, super-resolution, unmixing, deconvolution



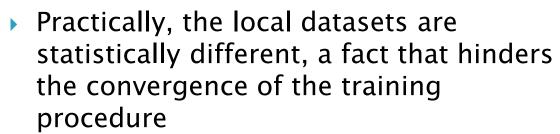
- Novel sparse coding schemes have been proposed that utilize the idea of CNN-based regularizers for imposing learnable structure priors with an application to hyper-spectral denoising
- An efficient distributed unmixing scheme has been devised by employing a sparse and low rank non-negative factorization

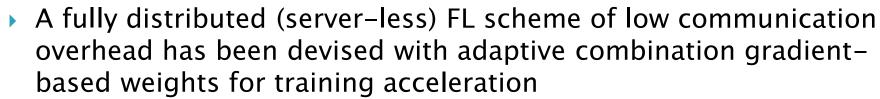
¹A. Gkillas, D. Ampeliotis, K. Berberidis, "Connections between Deep Equilibrium and Sparse Representation models with Application to Hyperspectral Imaging", submitted to IEEE Transactions on Image Processing

²C.G. Tsinos, A.A. Rontogiannis and Kostas Berberidis, "Distributed Blind Hyperspectral Unmixing via Joint Sparsity and Low-Rank Constrained Non-Negative Matrix Factorization", IEEE Transactions on Computational Imaging, 2017.

Federated learning over non-iid datasets

 In FL, collaborating agents employ local datasets and exchange only parameter information for training purposes





 FL dictionary learning has been considered by introducing the concept of "atom probabilities" for appropriate parameter aggregation

¹A. Gkillas, D. Ampeliotis, K. Berberidis, "Federated Dictionary Learning from Non-IID Data", to be presented in IVMSP'22

²E. Georgatos, C. Mavrokefalidis, K. Berberidis, "Efficient Fully Distributed Federated Learning with Adaptive Local Links", arXiv:2203.12281, 2022

Concluding remarks

- Modern wireless communication systems will support (massive) machine-type transmissions with a large number of participating devices
- This scenario raises several challenges like communication overheads, scalability and latency issues as well as privacy considerations
- Signal processing and machine learning (SPML) is "moving" towards the edge of the network (EdgeSPML) in distributed architectures to address these challenges
- Two main directions can be identified
 - SPML for communications and communications for SPML
- In HE, several related topics (e.g., in SNS-JU, CL4&5)
- A combined expertise at the intersection of signal processing, machine learning and wireless communications, enables to effectively address many of the challenges in EdgeSPML

Thank you for your attention!

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