Joint Resource Allocation and Link Association for URLLC Puncturing eMBB Traffic in Multicell Networks

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Introduction

Scenario

- mmWave
 - Multicell¹
- Downlink eMBB and URLLC
 - Puncturing²
 - Hybrid³

¹mitigate path loss and shadowing

²improve spectrum efficiency for sparse URLLC traffic

³accommodate demands for bursty URLLC traffic

Sparse URLLC Traffic

- As the adoption of autonomous vehicles⁴ and drone delivery is still minor, URLLC downlink control traffic is rather dispersed.
- In this regard, having a dedicated channel in URLLC service wastes spectral resources, and such waste scales linearly with the number of base stations in the network.

⁴include private cars and public transits

 Do note that the discussion is applicable to any wireless systems serving two heterogeneous qualities of service (QoS) using short-wavelength technology.

Contributions

- A novel linear model for URLLC puncturing eMBB traffic in multicell networks is introduced, upon which optimality analysis for the multiplexing procedure is conducted.
- The proportional fairness algorithm's asymptotical optimality
 [3] is generalized for eMBB resource allocation in multiconnectivity-based networks.
- The URLLC problem's optimal substructure is proved.
- Two approximation algorithms, nearest association algorithm and tightest association algorithm, that jointly schedule URLLC resources and links in multicell networks are proposed.
- An optimal variant of the two algorithms for single-cell networks is derived.

Related Work

- Current models for URLLC puncturing scheme in the literature are mostly non-linear [2].
- Whilst proposed linear models are either intractable [5] or inappropriate [1] for discrete subchannel allocation with multiple URLLC users.
- Many [2, 5] heuristically optimize the URLLC problem at each time minislot, without analyzing the optimal substructure.

System Model

- Base stations of separated channel bandwidths, downlink, OFDMA, puncturing, eMBB and URLLC users.
- Saturated eMBB traffic [4]: Each eMBB user has infinite amount of data to be transmitted.
- Strict URLLC constraint: Each URLLC user has an amount of data required to be served within a minislot.
- The system aims to maximize eMBB total average rate and fairness while satisfying URLLC demands.

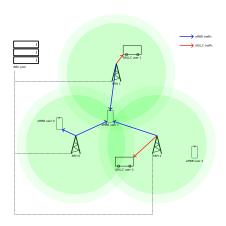


Figure: System model

Stochastic Problem

$$\begin{array}{ll}
\text{maximize} & \sum_{u} \ln \bar{R}_{u} \\
\boldsymbol{\alpha}, \boldsymbol{\beta}, \boldsymbol{\delta}
\end{array}$$

$$\sum_{u} \ln \bar{R}_{u} \tag{1a}$$

subject to

$$\sum_{u} \alpha_{u,n,s,l} \le 1 \qquad \forall n \forall s \forall l, \tag{1b}$$

$$\alpha_{u,n,s,l} \in \{0,1\} \ \forall u \forall n \forall s \forall l,$$
 (1c)

$$\sum_{s} \delta_{v,n,m,s} \le 1 \qquad \forall v \forall n \forall m, \tag{1d}$$

$$\beta_{v,u,n,m,s,l} \le \delta_{v,n,m,s} \forall v \forall u \forall n \forall m \forall s \forall l, \tag{1e}$$

$$\delta_{v,n,m,s} \in \{0,1\} \ \forall v \forall n \forall m \forall s, \tag{1f}$$

$$\sum \beta_{v,u,n,m,s,l} \le \alpha_{u,n,s,l} \,\forall u \forall n \forall m \forall s \forall l, \tag{1g}$$

$$R_{\nu,n,m} \ge R_{\nu,n,m}^{dm} \ \forall \nu \forall n \forall m, \tag{1h}$$

$$\beta_{v,u,n,m,s,l} \in \{0,1\} \ \forall v \forall u \forall n \forall m \forall s \forall l \tag{1i}$$

• The system maximizes eMBB traffic's total average rate and fairness (1a):

$$egin{array}{ll} {
m maximize} & \sum_{u} \ln ar{R}_{u}. \end{array}$$

• For each time slot, the system allocates a subchannel to at most one eMBB user (1b):

$$\forall n \colon \forall s \colon \forall I \colon \sum_{u} \alpha_{u,n,s,I} \leq 1.$$

• For each time slot, the system either un-allocates or allocates a subchannel to an eMBB user (1c):

$$\forall u \colon \forall n \colon \forall s \colon \forall I \colon \alpha_{u,n,s,I} \in \{0,1\}$$
.

 For each time minislot, the system associates at most one base station to a URLLC user (1d):

$$\forall v \colon \forall n \colon \forall m \colon \sum_{s} \delta_{v,n,m,s} \leq 1.$$

• For each time minislot, the system allocates a subchannel to a URLLC user only if it associates the corresponding base station to the user (1e):

$$\forall v \colon \forall u \colon \forall n \colon \forall m \colon \forall s \colon \forall I \colon \beta_{v,u,n,m,s,I} \leq \delta_{v,n,m,s}.$$

• For each time minislot, the system either un-associates or associates a base station to a URLLC user (1f):

$$\forall v \colon \forall n \colon \forall m \colon \forall s \colon \delta_{v,n,m,s} \in \{0,1\}$$
.

 For each time minislot, the system allocates a subchannel to at most one URLLC user, and punctures the subchannel for a URLLC user only if it allocates the subchannel to the corresponding eMBB user (1g)⁵:

$$\forall u \colon \forall n \colon \forall s \colon \forall l \colon \sum_{v} \beta_{v,u,n,m,s,l} \leq \alpha_{u,n,s,l}.$$

 For each time minislot, the system serves demands of URLLC users without delay (1h):

$$\forall v \colon \forall n \colon \forall m \colon R_{v,n,m} \geq R_{v,n,m}^{dm}.$$

 For each time minislot, the system employs URLLC puncturing scheme (1i):

$$\forall v \colon \forall u \colon \forall n \colon \forall m \colon \forall s \colon \forall I \colon \beta_{v,u,n,m,s,I} \in \{0,1\}$$
.

⁵Proof in supplementary

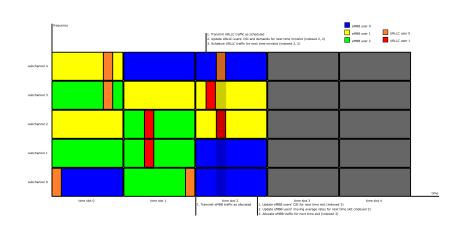


Figure: Multiplexing procedure

eMBB Problem

$$\underset{\boldsymbol{\alpha}}{\text{maximize}} \quad \sum_{u} \ln \bar{R}_{u} \tag{2a}$$

subject to
$$\sum_{u} \alpha_{u,n,s,l} \leq 1 \quad \forall n \forall s \forall l,$$
 (2b)

$$\alpha_{u,n,s,l} \in \{0,1\} \forall u \forall n \forall s \forall l \tag{2c}$$

Relaxed eMBB Problem

$$\begin{array}{cc}
\text{maximize} & \sum_{u} \ln \bar{R}'_{u} \\
\alpha'
\end{array} \tag{3a}$$

subject to
$$\sum_{u} \alpha'_{u,n,s,l} \le 1 \forall n \forall s \forall l,$$
 (3b)

$$\alpha'_{u,n,s,l} \ge 0 \forall u \forall n \forall s \forall l \tag{3c}$$

Gradient Problem

$$\begin{array}{ccc}
\text{maximize} & \sum_{u} \frac{r'_{u,n_0}}{\tilde{r}'_{u,n_0}} & (4a)
\end{array}$$

subject to
$$\sum_{u} \alpha'_{u,n_0,s,l} \leq 1 \forall s \forall l,$$
 (4b)
$$\alpha'_{u,n_0,s,l} \geq 0 \forall u \forall s \forall l$$
 (4c)

$$\alpha'_{u,n_0,s,l} \ge 0 \forall u \forall s \forall l$$
 (4c)

 The relaxed moving average rate of eMBB user is defined based on exponential moving average (EMA) as

$$\forall u \colon \tilde{r}'_{u,n} = \begin{cases} \frac{1}{n} \sum_{s,l} \frac{1}{u} \mathfrak{r}_{u,n,s,l} & n = 0\\ (1 - \epsilon) \tilde{r}'_{u,n-1} + \epsilon r'_{u,n-1} & n = 1, \dots, n-1 \end{cases} \begin{bmatrix} \frac{bits}{slot} \end{bmatrix}.$$

$$(5)$$

 $oldsymbol{lpha}$ The initial value of which is defined by the feasible policy $oldsymbol{\hat{lpha}}'$ for the relaxed eMBB problem where 6

$$\forall u \colon \forall s \colon \forall I \colon \hat{\alpha}'_{u,n,s,I} = \begin{cases} \frac{1}{\mathfrak{u}} & n = 0\\ 0 & n = 1, \dots, \mathfrak{n} - 1 \end{cases}$$
 (6)

⁶Proof in supplementary

- To solve the gradient problem, linear programming techniques might be considered.
- Although ellipsoid methods solve by polynomial time in theory, and interior-point methods converge remarkably fast in practice, such solvers do not guarantee a binary solution.
- Fortunately, the gradient problem's totally unimodular property makes simplex methods viable.

• Given a policy $\hat{\alpha}$ where for n = 0, ..., n - 1, $\hat{\alpha}_n$ is a basic optimal point for the n^{th} gradient problem, then $\hat{\alpha}$ is an asymptotically optimal policy for the eMBB problem⁷.

⁷Proof in supplementary

- However, simplex methods run in exponential time in worst-case scenarios, which could not meet the multiplexing procedure's time requirement for allocating eMBB resources.
- By further observations, it is found that the proportional fairness algorithm solves the gradient problem optimally in linear time.

 The following policy is asymptotically optimal with respect to the eMBB problem⁸:

$$\forall n \colon \forall s \colon \forall I \colon \hat{\alpha}_{u,n,s,I} = \begin{cases} 1 & u = \min_{\hat{u}} \arg \max_{\hat{u}} \frac{\mathfrak{r}_{\hat{u},n,s,I}}{\tilde{r}_{\hat{u},n}'} \\ 0 & \text{otherwise} \end{cases} . \tag{7}$$

⁸Proof in supplementary

URLLC Problem

$$oldsymbol{eta}_{n_0}, oldsymbol{\delta}_{n_0}$$

$$\begin{array}{ll}
\text{maximize} & \sum_{u} \frac{r'_{u,n_0}}{\tilde{r}'_{u,n_0}} \\
\beta_{n_0}, \delta_{n_0} & \tilde{r}'_{u,n_0}
\end{array}$$
(8a)

$$\sum_{s} \delta_{\nu, n_0, m, s} \le 1 \qquad \forall \nu \forall m, \tag{8b}$$

$$\beta_{v,u,n_0,m,s,l} \le \delta_{v,n_0,m,s} \forall v \forall u \forall m \forall s \forall l, \tag{8c}$$

$$\delta_{v,n_0,m,s} \in \{0,1\} \quad \forall v \forall m \forall s,$$
 (8d)

$$\sum \beta_{v,u,n_0,m,s,l} \le \hat{\alpha}_{u,n_0,s,l} \,\forall u \forall m \forall s \forall l, \tag{8e}$$

$$R_{\nu,n_0,m} \ge R_{\nu,n_0,m}^{dm} \ \forall \nu \forall m, \tag{8f}$$

$$\beta_{v,u,n_0,m,s,l} \in \{0,1\} \quad \forall v \forall u \forall m \forall s \forall l$$
 (8g)

Puncturing Problem

$$\overset{\text{minimize}}{\beta_{n_0,m_0},\delta_{n_0,m_0}}$$

subject to

$$\underset{\boldsymbol{\beta}_{n_0,m_0},\boldsymbol{\delta}_{n_0,m_0}}{\operatorname{minimize}} \quad \sum_{v,u,s,l} \frac{\mathfrak{r}_{u,n_0,s,l}}{\tilde{r}'_{u,n_0}} \beta_{v,u,n_0,m_0,s,l} \tag{9a}$$

$$\sum_{s} \delta_{\nu,n_0,m_0,s} \le 1 \qquad \forall \nu, \tag{9b}$$

$$\beta_{v,u,n_0,m_0,s,l} \le \delta_{v,n_0,m_0,s} \forall v \forall u \forall s \forall l,$$
 (9c)

$$\delta_{v,n_0,m_0,s} \in \{0,1\} \quad \forall v \forall s, \tag{9d}$$

$$\sum_{l} \beta_{v,u,n_0,m_0,s,l} \le \hat{\alpha}_{u,n_0,s,l} \ \forall u \forall s \forall l, \tag{9e}$$

$$r_{\nu,n_0,m_0} \ge \mathfrak{r}_{\nu,n_0,m_0}^{dm} \quad \forall \nu, \tag{9f}$$

$$\beta_{v,u,n_0,m_0,s,l} \in \{0,1\} \quad \forall v \forall u \forall s \forall l$$
 (9g)

URLLC Problem's Optimal Substructure

• Given a time slot policy $(\hat{\beta}_{n_0}, \hat{\delta}_{n_0})$ where for $m = 0, \dots, \mathfrak{m} - 1$, $(\hat{\beta}_{n_0,m}, \hat{\delta}_{n_0,m})$ solves the m^{th} puncturing problem, then $(\hat{\beta}_{n_0}, \hat{\delta}_{n_0})$ is an optimal point for the URLLC problem⁹.

⁹Proof in supplementary

Scheduling Problem

minimize
$$\gamma_{n_0,m_0}, \delta_{n_0,m_0} \quad \sum_{v,s,l} c_{n_0,s,l} \gamma_{v,n_0,m_0,s,l}$$
(10a)
subject to
$$\sum_{s} \delta_{v,n_0,m_0,s} \leq 1 \quad \forall v,$$
(10b)
$$\gamma_{v,n_0,m_0,s,l} \leq \delta_{v,n_0,m_0,s} \forall v \forall s \forall l,$$
(10c)
$$\delta_{v,n_0,m_0,s} \in \{0,1\} \quad \forall v \forall s,$$
(10d)
$$\sum_{v} \gamma_{v,n_0,m_0,s,l} \leq 1 \quad \forall s \forall l,$$
(10e)
$$\sum_{s,l} \gamma_{v,n_0,m_0,s,l} \mathfrak{r}_{v,n_0,m_0,s} \geq \mathfrak{r}_{v,n_0,m_0}^{dm} \quad \forall v,$$
(10f)

(10g)

 $\gamma_{v,n_0,m_0,s,l} \in \{0,1\} \quad \forall v \forall s \forall l$

• The puncturing cost of subchannel is defined as

$$\forall n \colon \forall s \colon \forall l \colon c_{n,s,l} = \max_{u} \frac{\mathfrak{r}_{u,n,s,l}}{\tilde{r}'_{u,n}}. \tag{11}$$

 The scheduling problem is equivalent to the puncturing problem¹⁰.

¹⁰Proof in supplementary

ullet The scheduling problem is independent from lpha.

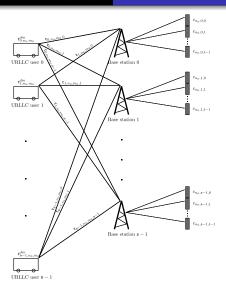


Figure: Scheduling Problem

Homogeneous Problem

minimize
$$\gamma_{n_0,m_0}, \delta_{n_0,m_0} \qquad c_{n_0}^{mn} \sum_{v,s,l} \gamma_{v,n_0,m_0,s,l} \qquad (12a)$$
subject to
$$\sum_{s} \delta_{v,n_0,m_0,s} \leq 1 \qquad \forall v, \qquad (12b)$$

$$\gamma_{v,n_0,m_0,s,l} \leq \delta_{v,n_0,m_0,s} \forall v \forall s \forall l, \qquad (12c)$$

$$\delta_{v,n_0,m_0,s} \in \{0,1\} \quad \forall v \forall s, \qquad (12d)$$

$$\sum_{v} \gamma_{v,n_0,m_0,s,l} \leq 1 \qquad \forall s \forall l, \qquad (12e)$$

$$\sum_{s,l} \gamma_{v,n_0,m_0,s,l} \geq \frac{\mathfrak{r}_{v,n_0,m_0}^{dm}}{\mathfrak{r}_{v,n_0,m_0}^{mx}} \ \forall v, \qquad (12f)$$

$$\gamma_{v,n_0,m_0,s,l} \in \{0,1\} \quad \forall v \forall s \forall l \qquad (12g)$$

• The minimum puncturing cost is defined as

$$\forall n \colon c_n^{mn} = \min_{s,l} c_{n,s,l}. \tag{13}$$

The maximum rate of URLLC user is defined as

$$\forall n \colon \forall m \colon \forall v \colon \mathfrak{r}_{v,n,m}^{mx} = \max_{s} \mathfrak{r}_{v,n,m,s}. \tag{14}$$

•
$$c_{n_0}^{mn} \sum_{v} \left[\frac{v_{v,n_0,m_0}^{dm}}{v_{v,n_0,m_0}^{mx}} \right]$$
 is the optimal value for the homogeneous problem.

• The nearest association algorithm defines the following point:

$$\forall v \colon \hat{\delta}_{v,n_0,m_0,s} = \begin{cases} 1 & s = \min_{\hat{s}} \arg\max_{\hat{s}} \mathfrak{r}_{v,n_0,m_0,\hat{s}} \\ 0 & \text{otherwise} \end{cases},$$

$$\forall s \colon \forall v \colon \hat{\gamma}_{v,n_0,m_0,s,l} = \begin{cases} \sum_{\hat{v}=0}^{v-1} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0}^{m}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s} \\ \leq \left(\sum_{\hat{v}=0}^{v} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0}^{m}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s} \right) - 1 \\ 0 & \text{otherwise} \end{cases}.$$

• The nearest association algorithm is a $\frac{c_{n_0}^{mx}}{c_{n_0}^{mn}}$ -approximation algorithm for the scheduling problem¹¹.

¹¹Proof in supplementary

• The maximum puncturing cost is defined as

$$\forall n \colon c_n^{mx} = \max_{s,l} c_{n,s,l}. \tag{15}$$

 The opportunistic nearest association algorithm defines the following point:

$$\forall \mathbf{v} \colon \hat{\delta}_{\mathbf{v},n_0,m_0,s} = \begin{cases} 1 & s = \min_{\hat{s}} \arg\max_{\hat{s}} \mathfrak{r}_{\mathbf{v},n_0,m_0,\hat{s}}, \\ 0 & \text{otherwise} \end{cases},$$

$$\forall \mathbf{s} \colon \forall \mathbf{v} \colon \hat{\gamma}_{\mathbf{v},n_0,m_0,s,l} = \begin{cases} & \left(\sum_{\hat{v}=0}^{v-1} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0}^{m}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s}\right) + 1 \\ 1 & \leq \mathbf{card} \left\{ \hat{\boldsymbol{f}} \middle| c_{n_0,s,l} > c_{n_0,s,\hat{l}} \lor (c_{n_0,s,l} = c_{n_0,s,\hat{l}} \land l > \hat{\boldsymbol{l}}) \right\} \\ & \leq \sum_{\hat{v}=0}^{v} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0}^{m}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s} \\ 0 & \text{otherwise} \end{cases}$$

 The opportunistic nearest association algorithm is an optimal algorithm for the scheduling problem in single-cell networks¹².

¹²Proof in supplementary

The tightest association algorithm defines the following point:

$$\forall v \colon \hat{\delta}_{v,n_0,m_0,s} = \begin{cases} 1 & s = \min_{\hat{s}} \arg\min_{\hat{s}} c_{n_0,\hat{s}}^{mn} \left\lceil \frac{\mathfrak{r}_{v,n_0,m_0}^{dm}}{\mathfrak{r}_{v,n_0,m_0,\hat{s}}} \right\rceil, \\ 0 & \text{otherwise} \end{cases}$$

$$\forall s \colon \forall v \colon \hat{\gamma}_{v,n_0,m_0,s,l} = \begin{cases} & \sum_{\hat{v}=0}^{v-1} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0,s}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s} \\ & \leq \left(\sum_{\hat{v}=0}^{v} \left\lceil \frac{\mathfrak{r}_{\hat{v},n_0,m_0}^{dm}}{\mathfrak{r}_{\hat{v},n_0,m_0,s}} \right\rceil \hat{\delta}_{\hat{v},n_0,m_0,s} \right) - 1 \\ & 0 & \text{otherwise} \end{cases}.$$

• The tightest association algorithm is a ρ -approximation algorithm for the scheduling problem¹³.

¹³Proof in supplementary

• The opportunistic tightest association algorithm is an optimal algorithm for the scheduling problem in single-cell networks.

Simulation

- Compare performance with dedicated URLLC channel systems?
- Compare performance with [1] in single-cell networks with one URLLC user.
- Compare performance with [5] in single-cell networks with multiple URLLC users?
- Compare performance of nearest association algorithm and opportunistic nearest association algorithm in multicell networks.

Parameters

- 3 base stations
- 51 subchannels per base station
- 100 eMBB users
- 100 time slots
- 7 time minislots per time slot
- 3 algorithms
 - Opportunistic Nearest Association Algorithm (ONAA)
 - Opportunistic Tightest Association Algorithm (OTAA)
 - Exhaustive Algorithm (EA)

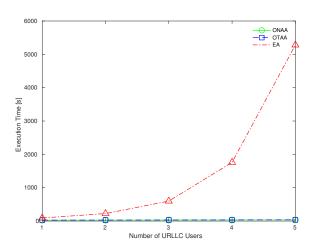


Figure: Optimality execution time

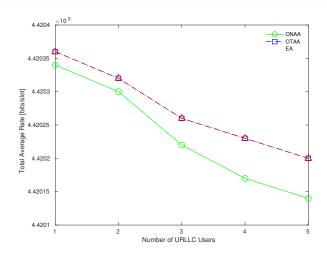


Figure: Optimality total average rates

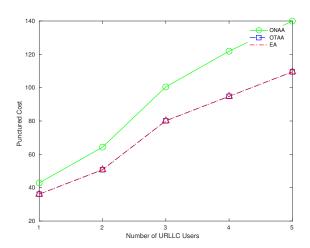


Figure: Optimality punctured costs

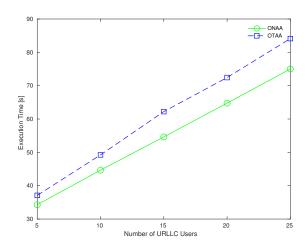


Figure: Heuristics execution time

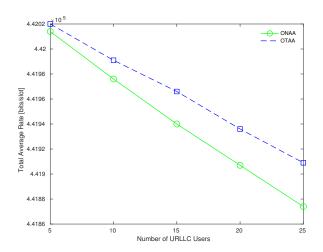


Figure: Heuristics total average rates

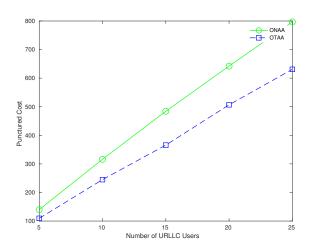


Figure: Heuristics punctured costs

Introduction System Model Simulation Conclusion

Conclusion

References

- Arjun Anand, Gustavo de Veciana, and Sanjay Shakkottai. "Joint Scheduling of URLLC and eMBB Traffic in 5G Wireless Networks". In: IEEE/ACM Transactions on Networking 28.2 (2020), pp. 477–490. DOI: 10.1109/TNET.2020.2968373.
- [2] Anupam Kumar Bairagi et al. "Coexistence Mechanism Between eMBB and URLLC in 5G Wireless Networks". In: IEEE Transactions on Communications 69.3 (2021), pp. 1736–1749. DOI: 10.1109/TC0MM.2020.3040307.
- [3] Harold J. Kushner and Philip A. Whiting. "Asymptotic Properties of Proportional-Fair Sharing Algorithms". In: 40th Annual Allerton Conference on Communication, Control, and Computing. 2002, pp. 1051–1059.
- [4] Alexander L. Stolyar. "On the Asymptotic Optimality of the Gradient Scheduling Algorithm for Multiuser Throughput Allocation". In: Operations Research 53.1 (2005), pp. 12–25. DOI: 10.1287/opre.1040.0156.
- [5] Hao Yin, Lyutianyang Zhang, and Sumit Roy. "Multiplexing URLLC Traffic Within eMBB Services in 5G NR: Fair Scheduling". In: IEEE Transactions on Communications 69.2 (2021), pp. 1080–1093. DOI: 10.1109/T00MW.2020.3035852.