Abstract

We study $G/G/\infty$ queues with renewal alternating service interruptions, where the service station experiences up and down periods. The system operates normally in the up periods, and all servers stop functioning while customers continue entering the system during the down periods. The amount of service a customer has received when an interruption occurs will be conserved and the service will resume when the down period ends. We use a two-parameter process to describe the system dynamics: $X^r(t,y)$ tracking the number of customers in the system at time $t$ that have residual service times strictly greater than $y$. The service times are assumed to satisfy either of the two conditions: (i) i.i.d. with a distribution of a finite support, or (ii) a stationary and weakly dependent sequence satisfying the $\phi$-mixing condition and having a continuous marginal distribution function. We consider the system in a heavy-traffic asymptotic regime where the arrival rate gets large and service time distribution is fixed, and the interruption down times are asymptotically negligible while the up times are of the same order as the service times. We show FLLN and FCLT for the process $X^r(t,y)$ in this regime, where the convergence is in the space $D([0,\infty), (D, L_1))$ endowed with the Skorohod $M_1$ topology. The limit processes in the FCLT possess a stochastic decomposition property.

Keywords: $G/G/\infty$ queue; dependent service times; service interruptions; two-parameter stochastic processes; FLLN; FCLT; Skorohod $M_1$ topology

1 Introduction

We study $G/G/\infty$ queues with service interruptions. The service station operates in a renewal alternating environment with “up” and “down” periods. All servers function normally in an “up” period, but break down in a “down” period. Customers enter the system according to some point process, without being affected by the interruptions. They are served immediately upon arrival if
arriving during an “up” period, and wait for service until the next “up” period starts otherwise. Services received when interruptions occur will be conserved, and when interruptions end, services resume from where they are left. Customers may experience multiple interruption “down” periods before completing their services. To describe the system dynamics, we use a two-parameter process \( X^r(t, y) \), representing the numbers of customers in the system at time \( t \) that have residual service times strictly greater than \( y \). The second parameter will help track the amount of services that are interrupted. The total number of customers in the system at time \( t \) is \( X(t) = X^r(t, 0) \).

We consider the system in a heavy-traffic asymptotic regime. The arrival rate gets large while the service time distributions are fixed. The alternating renewal “up” and “down” times are scaled such that the “up” time periods are of the same order as service times, while the “down” time periods are asymptotically negligible compared with service times. Specifically, we let the “down” times be of order \( O(1/n^\gamma) \) for some \( \gamma > 0 \) while the service times are \( O(1) \). We study the system under two different assumptions on the service times: (i) i.i.d. service times with a finite support, and (ii) the sequence of service times is stationary and weakly dependent, satisfying the \( \phi \)-mixing condition and having a continuous marginal distribution function. We show functional law of large numbers (FLLNs) and functional central limit theorems (FCLTs) for the processes \( X^r(t, y) \) and \( X(t) \) in the asymptotic regime under the two assumptions on the service times.

The limit processes in the FLLNs and FCLTs of the process \( X^r(t, y) \) have sample paths in the space \( \mathbb{D} := \mathbb{D}([0, \infty), \mathbb{D}([0, \infty), \mathbb{R})) \). To prove the convergence, due to unmatched jumps of prelimit and limit processes, we employ the Skorohod \( M_1 \) topology in the space \( \mathbb{D} \), where the interior space \( \mathbb{D} := \mathbb{D}([0, \infty), \mathbb{R}) \) is endowed with the \( L_1 \) norm. The fluid limits take the same forms as those in the corresponding \( G/G/\infty \) queues without interruptions [11, 25, 28]. However, the convergence is proved in the Skorohod \( M_1 \) topology. The limits in the FCLT have extra terms to capture the effect of service interruptions, in addition to the same terms as those in the corresponding \( G/G/\infty \) queues without interruptions [11, 25, 28]. We obtain a stochastic decomposition property for the limiting processes in the FCLTs, that is, the variabilities of arrivals, services and interruptions are captured in mutually independent processes. When the arrival limit process is a Brownian motion and the limiting counting process for the number of “up” and “down” periods is Poisson, we characterize the transient and stationary distributions for the limiting processes (Corollaries 3.1 and 4.1).

1.1 Literature and contributions

There has been a large body of literature on queues with service interruptions; see, e.g., [16, 17, 22, 23] and references therein. Infinite-server queues with service interruptions (or in a random environment) are studied in [1, 2, 5, 8, 20, 21, 23]; however, all these studies have focused on either Poisson arrivals and/or exponential service times. Infinite-server queues with general service times and service interruptions (or in a random environment) are studied in [4, 6, 9, 12]. Our work is the first to establish two-parameter heavy-traffic limits for \( G/G/\infty \) queues with service interruptions, under general assumptions on the arrival, service and interruption processes.

Infinite-server queues with general arrival and service processes have been studied in [10, 11, 18, 35] and more recently in [25, 28, 30]. In [11], when the service time distribution has a finite support, the key idea is to treat that as a mixture of deterministic service time distributions, and split the arrival process into corresponding arrival processes associated with each deterministic service time. The main step relies on the FCLT for the split counting processes. That key idea can be used to show FLLN and FCLT for the total count process \( X(t) \) in the system as well as the two-parameter process \( X^r(t, y) \). We first generalize that approach to \( G/G/\infty \) queues with service interruptions when the service times have finite support in this paper. In [25, 28], Pang and Whitt
have established FLLN and FCLT for the two-parameter process $X^r(t, y)$ when the service times are i.i.d. with continuous distributions and when the service times form a stationary and weakly dependent sequence satisfying certain mixing conditions and have continuous marginal distributions. The approach in [25, 28] relies on the important observation first made in Krichagina and Puhalskii [18] that the queue length process in $G/GI/\infty$ queues can be represented via sequential empirical processes driven by service times. Here we make a novel and important observation that for $G/G/\infty$ queues with service interruptions, the two-parameter process $X^r(t, y)$ can be represented via sequential empirical processes driven by service times together with the cumulative “up” time process. This enables us to establish the limits in the fluid and diffusion scales. Here we assume that the service time distributions are either of finite support or weakly dependent with continuous marginal distributions as in [11, 28]. It is conceivable that the results can be extended to the most general conditions on service times, but that is beyond the scope of this paper. Our approach is extended to study the total count processes for $G/G/N (+G)$ queues in the Halfin-Whitt regime in [19] when the services are i.i.d. with a continuous distribution function. It remains to prove the two-parameter limits for these models. We also conjecture that it can be potentially extended to study other many-server non-Markovian models.

An interesting stochastic decomposition property has been established for $M/M/\infty$ and $M/G/\infty$ queues with service interruptions or vacations in [2, 4, 8, 12, 21]. Namely, the steady-state distribution of the number of customers in the system can be decomposed into two independent components: the steady-state distribution of the number of customers in the system without interruptions and the distribution taking into account the effects of service interruptions. In the heavy-traffic setting, similar stochastic decomposition properties are proved for $G/M/\infty$ queues in [23] and for $G/G/1$ queues in [16, 17]. In this paper we establish a stochastic decomposition property for the two-parameter process $X^r(t, y)$ and the total count process $X(t)$ of $G/G/\infty$ queues with service interruptions under the two assumptions on the service time distributions. As a result, their steady-state distributions also possess a stochastic decomposition property. In fact, as shown in Theorems 3.2 and 4.2, the limiting two-parameter processes and total count processes in the diffusion scale are decomposed into independent processes which capture variabilities in the arrival processes, service processes and service interruptions.

To prove the convergence of the fluid and diffusion scaled processes for queues with service interruptions, the Skorohod $M_1$ topology in the space $\mathbb{D}$ is necessary to take into account the unmatched jumps and discontinuity when the interruption “down” times are scaled properly. For example, single-server queues with service interruptions in the conventional heavy-traffic regime [16, 17] and $G/M/N(+M)$ queues with service interruptions in the many-server heavy-traffic regimes [23, 24] all require the convergence in $(\mathbb{D}, M_1)$. To prove the convergence of the two-parameter process $X^r(t, y)$ of $G/G/\infty$ queues with service interruptions, it requires to use the Skorohod $M_1$ topology in the space $\mathbb{D}_D$. Since the Skorohod $M_1$ topology is only well defined for the space $\mathbb{D}([0, \infty), S)$ when $S$ is a Banach space [29, 32, 36], we endow the interior space $\mathbb{D}$ with the $L_1$ norm. This topology is discussed to study $G/GI/\infty$ queues on page 350 of §10.3.1 in Whitt [36]. We establish some useful continuity properties in the space $\mathbb{D}_D$ with Skorohod $M_1$ topology, which may be of separate interest.

1.2 Notation

We use $\mathbb{R}^k$ (and $\mathbb{R}_+^k$), $k \geq 1$, to denote real-valued $k$-dimensional (nonnegative) vectors, and write $\mathbb{R}$ and $\mathbb{R}_+$ for $k = 1$. Let $\mathbb{D}^k = \mathbb{D}([0, \infty), \mathbb{R}^k)$ denote the $\mathbb{R}^k$-valued function space of all right continuous functions on $[0, \infty)$ with left limits everywhere in $(0, \infty)$. Denote $\mathbb{D} \equiv \mathbb{D}^1$. 
Let \((\mathbb{D}_k, J_1) = (\mathbb{D}, J_1) \times \cdots \times (\mathbb{D}, J_1)\) be the \(k\)-fold product of \((\mathbb{D}, J_1)\) with the product topology. Similarly, let \((\mathbb{D}_k, M_1) = (\mathbb{D}, M_1) \times \cdots \times (\mathbb{D}, M_1)\) be the \(k\)-fold product of \((\mathbb{D}, M_1)\) with the product topology. We use \(C_x\) and \(D_x\) to denote the space of nondecreasing functions in \(C\) and \(D\), respectively. Denote \(|| \cdot ||\) the uniform norm: for any real-valued function \(x\) and \(T > 0\), \(|| x ||_T = \sup_{t \in [0,T]} |x(t)|\).

Notations \(\rightarrow\) and \(\Rightarrow\) mean convergence of real numbers and convergence in distributions. The abbreviations \(w.p.1\) and \(a.e.\) mean \(\text{with probability 1 and almost everywhere}\), respectively. We use the familiar big-\(O\) and small-\(o\) notations for deterministic functions: for two real-valued functions \(f\) and non-zero \(g\), we write \(f(x) = O(g(x))\) if \(\lim \sup_{x \to \infty} |f(x)/g(x)| < \infty\) and \(f(x) = o(g(x))\) if \(\lim \sup_{x \to \infty} |f(x)/g(x)| = 0\).

All random variables and processes are defined on a common probability space \((\Omega, \mathcal{F}, P)\). For any two complete separable metric spaces \(S_1\) and \(S_2\), we denote \(S_1 \times S_2\) as their product space, endowed with the product topology. Let \(D_D = D((0, \infty), D)\) denote the \(D\)-valued function space of all right continuous functions on \([0, \infty)\) with left limits everywhere in \((0, \infty)\). Let \(C_C = C([0, \infty), C)\) where \(C\) denotes the real-valued function space of all continuous functions on \([0, \infty)\). For \(x(t, s) \in D_D\), we often denote \(x(t) := x(t, \cdot)\). We use \((D_D, J_1)\) to denote the space \(D_D\) with both the interior and exterior \(D\) spaces endowed with the Skorohod \(J_1\) topology. We refer the reader to [33] and [25] for an introduction and some properties of \((D_D, J_1)\). We use \((D_D, M_1)\) to denote the space \(D_D\) with the exterior \(D\) space endowed with the Skorohod \(M_1\) topology, while the interior \(D\) space is endowed with the \(L_1\) norm, that is, for any \(x \in D\), the \(L_1\) norm is defined by \(\|x\|_{L_1} = \int_0^\infty |x(t)|dt\).

See §§10.3 and 11.5 in Whitt [36]. Note that the space \((D, L_1)\) is a Banach space, and thus, the space \(D((0, \infty), (D, L_1))\) with Skorohod \(M_1\) topology is well defined [29, 32, 36]. Let \(((D_D)_k, M_1) = (D_D, M_1) \times \cdots \times (D_D, M_1)\) be the \(k\)-fold product of \((D_D, M_1)\) with the product topology.

### 1.3 Organization of the paper

The rest of the paper is organized as follows. In §2, we describe the model in detail, and the assumptions on the interruptions are given in §2.1. We present the results for service times with finite support in §3 and their proofs in §5. The results for service times that form a stationary and weakly dependent sequence and have continuous marginal distributions are presented in §4 and their proofs are in §6. Some technical proofs in §§5–6 are given in the appendix §7.

### 2 The model

Consider a \(G/G/\infty\) queue subject to service interruptions. The service station is in a renewal alternating environment with \(up\) and \(down\) periods, where all servers function normally during the up period but break down during the down period. During an up period, customers will enter service immediately (no delay or queue). During a down period, customers in service will wait at their associated servers until the end of the down period and resume their services when the next up period starts, and new customers will continue to enter the system and get assigned to some free servers and wait there for service to start until the next up period begins.

Assume that customers arrive to the system according to a general arrival process \(A = \{A(t) : t \geq 0\}\) with \(\tau_i\) representing the arrival time of the \(i^{th}\) customer, i.e., \(A(t) = \max\{i \geq 1 : \tau_i \leq t\}\) for each \(t > 0\) and \(A(0) = 0\). Let \(\{\eta_i : i \geq 1\}\) be a sequence of random variables, with \(\eta_i\) representing the service times of the \(i^{th}\) customer. We will consider two assumptions on the sequence \(\{\eta_i : i \geq 1\}\): (i) i.i.d. with a finite support, and (ii) stationary and weakly dependent, satisfying the \(\phi\)-mixing condition and having a general continuous marginal distribution function (see Assumption 4.2
for the precise statement). This second assumption includes i.i.d. service times with a general continuous distribution function as a special case. Let \( \{(u_i, d_i) : i \geq 1\} \) be a sequence of i.i.d. random vectors with \( u_i \) and \( d_i \) representing the up and down times in the \( i^{th} \) up-down cycle of the underlying renewal process. We assume that the arrival, service and service-interruption processes are mutually independent.

Let \( X^r(t, y) \) be the number of customers in the system at time \( t \) that have a remaining amount of service time strictly greater than \( y \). Let \( X(t) \) be the total number of customers in the system at time \( t \). Then \( X(t) = X^r(t, 0) \) for each \( t \geq 0 \). We assume that the system starts empty at the starting epoch of an up period.

To describe the dynamics of \( X^r(t, y) \), we first define some auxiliary processes associated with the underlying renewal process \( \{(u_i, d_i) : i \geq 1\} \). Let the sequence \( \{T_i : i \geq 1\} \) be the renewal times, defined by \( T_i = \sum_{k=1}^{i} (u_k + d_k) \), for \( i \geq 1 \) and \( T_0 = 0 \). Let \( N = \{N(t) : t \geq 0\} \) be the associated renewal counting process, defined by \( N(t) = \max\{i \geq 0 : T_i \leq t\} \), \( t \geq 0 \). Let \( \xi = \{\xi(t) : t \geq 0\} \) be the service-availability process, defined by \( \xi(t) = 1 \) when \( T_i \leq t \leq T_i + u_{i+1} \) and \( \xi(t) = 0 \) when \( T_i + u_i < t < T_{i+1} \) for \( i \geq 0 \). The cumulative up-time process \( U = \{U(t) : t \geq 0\} \) is defined by \( U(t) = \int_0^t \xi(s) ds \), \( t \geq 0 \). The cumulative down-time process \( D = \{D(t) : t \geq 0\} \) is defined by \( D(t) = t - U(t) \) for each \( t \geq 0 \).

We now give a representation of the processes \( X^r(t, y) \) and \( X(t) \):

\[
X^r(t, y) = \sum_{i=1}^{A(t)} 1(\eta_i > U(t) - U(\tau_i) + y), \quad t, y \geq 0, \tag{2.1}
\]
\[
X(t) = X^r(t, 0) = \sum_{i=1}^{A(t)} 1(\eta_i > U(t) - U(\tau_i)), \quad t \geq 0. \tag{2.2}
\]

We give an intuitive explanation of representation in (2.1). For the process \( X^r(t, y) \), we count the arrivals up to time \( t \) that have residual service times strictly greater than \( y \). The amount of service that the \( i^{th} \) customer has received by time \( t \), if she is still in the system, is equal to \( U(t) - U(\tau_i) \), the cumulative up time from her arrival time \( \tau_i \) to time \( t \). Thus, for the \( i^{th} \) customer to be counted at time \( t \), the amount of received service \( U(t) - U(\tau_i) \) must be strictly less than the service requirement \( \eta_i \) minus \( y \), i.e., \( \eta_i > U(t) - U(\tau_i) + y \).

We will consider a sequence of systems indexed by \( n \) and let \( n \to \infty \), where the arrival processes and the underlying interruption processes are scaled while the service time distribution does not change with \( n \). We let the arrival rate increase to infinity as \( n \to \infty \) and the underlying interruption processes have the up times \( O(1) \) and the down times \( O(1/n^\gamma) \), for some positive constant \( \gamma > 0 \) (see §2.1).

### 2.1 Assumptions on service interruptions

We consider a scaling regime for the underlying service interruption process, where the down times are asymptotically negligible.

**Assumption 2.1.** The sequence of up and down times \( \{(u_i^n, d_i^n) : i \geq 1\} \) satisfies

\[
\{(u_i^n, n^\gamma d_i^n) : i \geq 1\} \Rightarrow \{(u_i, d_i) : i \geq 1\} \quad \text{in} \quad (\mathbb{R}^2)^\infty \quad \text{as} \quad n \to \infty, \tag{2.3}
\]

for some \( \gamma > 0 \), where \( u_i, d_i > 0 \) for each \( i \) w.p.1.
Let \( \{ \bar{T}_i : i \geq 0 \} \) be the associated sequence of cycle renewal times in the limit, defined by \( \bar{T}_i = \sum_{k=1}^{i} u_k, \ i \geq 1, \) and \( \bar{T}_0 = 0. \) Assumption 2.1 implies that \( \bar{T}_i < \bar{T}_{i+1} \) for all \( i \geq 0. \) Let \( N = \{ \hat{N}(t) : t \geq 0 \} \) be the associated renewal counting process in the limit, defined by \[
\hat{N}(t) = \max\{i \geq 0 : \bar{T}_i \leq t \}, \ t \geq 0.
\] (2.4)

Define the diffusion-scaled processes \( \hat{U}^n = n^{\gamma}(U^n - e) \) and \( \hat{D}^n = n^{\gamma}D^n = -\hat{U}^n, \) where \( e(t) \equiv t \) is the identity function, and \( U^n \) and \( D^n \) are the cumulative up and down times.

**Lemma 2.1.** Under Assumption 2.1,

\[
(U^n, D^n, N^n) \Rightarrow (e, 0, \hat{N}) \quad \text{in} \quad (\mathbb{D}_3, J_1) \quad \text{as} \quad n \to \infty,
\]

and

\[
(\hat{U}^n, \hat{D}^n) \Rightarrow (-\hat{J}, \hat{J}) \quad \text{in} \quad (\mathbb{D}_2, M_1) \quad \text{as} \quad n \to \infty,
\]

where the limit process \( \hat{J} = \{ \hat{J}(t) : t \geq 0 \} \) is defined by

\[
\hat{J}(t) = \sum_{i=1}^{\hat{N}(t)} d_i, \ t \geq 0,
\]

(2.7)

and \( \hat{N}(t) \) is defined in (2.4) and \( d_i \)’s are the limits in (2.3).

**Proof.** The proof of (2.5) follows directly from the Assumption 2.1 and the proof of (2.6) can be found in §5.4 of [23]. \( \blacksquare \)

### 3 Service times with finite support

In this section, we consider service time distributions that are i.i.d. and have a finite support. In particular, we assume that the service times \( \{ \eta_i : i \geq 1 \} \) are i.i.d. and have a distribution \( F \) with a finite positive support \( \{ x_1, ..., x_m \} \) with associated probabilities \( p_1, ..., p_m \) such that \( \sum_{i=1}^{m} p_i = 1. \) We say that a customer requiring service time \( x_i \) is a type \( i \) customer. Let \( A^n_i(t) \) be the cumulative number of arrivals of type \( i \) customers up to time \( t \) in the \( n^{th} \) system. Denote \( A^n = (A^n_1, ..., A^n_m). \) Let \( Disc(x) \) be the set of discontinuity points of \( x \) in \( [0, \infty) \) and \( \theta_s : \mathbb{D} \to \mathbb{D} \) be the shift operator defined by \( \theta_s(x)(t) = x(t + s) \) for \( t + s \geq 0 \) and \( \theta_s(x)(t) = 0 \) for \( t + s < 0 \) and \( t \geq 0. \) We make the following assumption on the diffusion-scaled arrival processes.

**Assumption 3.1.** There exist a deterministic non-decreasing function \( \Lambda(t) = (\Lambda_1(t), ..., \Lambda_m(t)) \) in \( \mathbb{D}^m \) and a stochastic process \( \bar{\Lambda} = (\bar{\Lambda}_1, ..., \bar{\Lambda}_m) \) in \( \mathbb{D}^m \) such that \( P(Disc(A_1) \cap Disc(\theta_s(A_j))) = 0 \) for all \( i, j = 1, ..., m \) and \( s \leq 0, \) and \( \bar{\Lambda}^n(t) \Rightarrow \bar{\Lambda}(t) \) in \( (\mathbb{D}_m, M_1) \) as \( n \to \infty, \) where \( \bar{\Lambda}^n = (\bar{\Lambda}_1^n, ..., \bar{\Lambda}_m^n) \) is defined by \( \bar{\Lambda}_i^n(t) := n^{\gamma}(n^{-1}A_i^n(t) - \Lambda_i(t)), \ t \geq 0, \ i = 1, ..., m. \)

The functions \( \Lambda_i(t), \ i = 1, ..., m, \) are centering terms in the FCLT. This assumption implies that the fluid-scaled processes satisfy an FLLN [36]: \( \bar{\Lambda}^n = n^{-1}A^n \Rightarrow \Lambda \) in \( (\mathbb{D}_m, M_1) \) as \( n \to \infty. \)

Let \( X_{n,i}(t,y) \) be the associated two-parameter processes for type \( i \) customers and \( X_{n,i}(t) \) be the associated total count process for type \( i \) customers. Then by (2.1) and (2.2), we can write

\[
X_{n,i}(t,y) = \begin{cases} 
A^n_i(t) - A^n_i(U^{n-1}(U^n(t) - (x_i - y))), & t \geq 0, \ 0 \leq y < x_i, \\
0, & t \geq 0, \ y \geq x_i,
\end{cases}
\]

(3.1)
and
\[ X_{n,i}(t) = A_i^n(t) - A_i^n(U^{n,-1}(U^n(t) - x_i)), \quad t \geq 0, \]  
(3.2)
where
\[ U^{n,-1}(t) = \sup\{s \geq 0 : U^n(s) \leq t\}, \quad t \geq 0. \]  
(3.3)
We also let \( U^{n,-1}(t) \equiv 0 \) for \( t < 0 \), so that the expressions (3.1) and (3.2) are well defined. We remark that the inverse process \( U^{n,-1} \) has sample paths in \( \mathbb{D} \) and can be represented as
\[ U^{n,-1}(t) = t + \sum_{k=1}^{N_n^u(t)} d_k^n, \quad t \geq 0, \]  
(3.4)
where
\[ \hat{N}_n^u(t) = \max\{i \geq 0 : \hat{T}_{n,i}^u \leq t\}, \quad t \geq 0, \]  
(3.5)
and \( \hat{T}_{n,i}^u = \sum_{k=1}^i u_k, \quad i \geq 1 \), and \( \hat{T}_{n,0} = 0 \). We remark that \( X^r_{n,i}(t,y) \) in (3.1) cannot be simply written as \( A^n_i(t) - A^n_i(U^{n,-1}(U^n(t) - (x_i - y)^+)) \) for all \( y \geq 0 \), as in the case without interruptions. By the definition of \( U^{n,-1}(t) \) in (3.3), the expression is only correct for \( 0 \leq y < x_i \). But in the FLLN and FCLT limits for \( X^r_{n,i}(t,y) \) in (3.6) and (3.8), we can use the expression \( t - (x_i - y)^+ \) to combine the two scenarios with \( 0 \leq y < x_i \) and \( y \geq x_i \).

We write \( X^r_n(t,y) = (X^r_{n,1}(t,y), ..., X^r_{n,m}(t,y)) \), \( X_n(t,y) = \sum_{i=1}^m X^r_{n,i}(t,y) \), \( X_n(t) = (X_{n,1}(t), ..., X_{n,m}(t)) \), and \( X_n(t) = \sum_{i=1}^m X_{n,i}(t) \). Now define the fluid-scaled processes \( \hat{X}^r_n(t,y) = n^{-1}X^r_n(t,y) \), \( \hat{X}^r_n(t,y) = n^{-1}X^r_n(t,y) \), \( \hat{X}^r_n(t,y) = n^{-1}X^r_n(t,y) \), and \( \hat{X}^r_n(t,y) = n^{-1}X^r_n(t,y) \). The fluid limits for these processes \( \hat{X}^r_n(t,y) \), \( \hat{X}^r_n(t,y) \), and \( \hat{X}^r_n(t,y) \) are stated in the following theorem. We remark that it is understood that for each \( i = 1, ..., m, \Lambda_i(t) \equiv 0 \) whenever \( t < 0 \), so that the limit processes in Theorem 3.1 are all well defined, and similarly for other relevant processes throughout the paper.

**Theorem 3.1.** (FLLN) Under Assumptions 2.1 and 3.1, \( (\hat{X}^r_n(t,y), \hat{X}^r_n(t,y), \hat{X}_n(t), \hat{X}_n(t)) \Rightarrow (x^r(t,y), x^r(t,y), x(t)) \) in \((\mathbb{D}_{m+1}, M_1) \times (\mathbb{D}_{m+1}, M_1)\) as \( n \to \infty \), where \( x^r(t,y) = (x^r_1(t,y), ..., x^r_m(t,y)) \), \( x(t) = (x_1(t), ..., x_m(t)) \), and for each \( i = 1, ..., m \),
\[ x^r_i(t,y) = \Lambda_i(t) - \Lambda_i(t - (x_i - y)^+), \quad x_i(t) = x^r_i(t,0) = \Lambda_i(t) - \Lambda_i(t - x_i), \]  
(3.6)
\[ x^r(t,y) = \sum_{i=1}^m x^r_i(t,y) \text{ and } x(t) = \sum_{i=1}^m x_i(t). \]
Define the diffusion-scaled processes \( \hat{X}^r_{n,i}(t,y) = n^{-1}(\hat{X}^r_{n,i}(t,y) - x^r_i(t,y)) \), \( \hat{X}^r_n(t,y) = \sum_{i=1}^m \hat{X}^r_{n,i}(t,y) \), \( \hat{X}^r_{n,i}(t,0) = n^{-1}(\hat{X}^r_{n,i}(t) - x_i(t)) \), \( \hat{X}_n(t) = \hat{X}^r_n(t,0) = \sum_{i=1}^m \hat{X}^r_{n,i}(t) \), and write \( \hat{X}^r_n(t,y) = (\hat{X}^r_{n,1}(t,y), ..., \hat{X}^r_{n,m}(t,y)) \), \( \hat{X}_n(t) = (\hat{X}^r_{n,1}(t), ..., \hat{X}^r_{n,m}(t)) \). We show an FCLT for these diffusion-scaled processes.

**Theorem 3.2.** (FCLT) Under Assumptions 2.1 and 3.1, assuming that \( \Lambda \) is absolutely continuous with density \( \lambda := (\lambda_1, ..., \lambda_m) \in \mathbb{C}^m \) a.e.,
\[ (\hat{X}^r_n(t,y), \hat{X}^r_n(t,y), \hat{X}_n(t)) \Rightarrow (\hat{X}^r(t,y), \hat{X}^r(t,y), \hat{X}(t)) \]  
(3.7)
in \( (\mathbb{D}_{m+1}, M_1) \times (\mathbb{D}_{m+1}, M_1) \) as \( n \to \infty \), where \( \hat{X}^r(t,y) = (\hat{X}^r_1(t,y), ..., \hat{X}^r_m(t,y)) \), \( \hat{X}(t) = (\hat{X}_1(t), ..., \hat{X}_m(t)) \),
\[ \hat{X}^r_i(t,y) = \hat{A}_i(t) - \hat{A}_i(t - (x_i - y)^+) + \hat{J}_i^r(t,y), \quad \hat{X}_i(t) = \hat{X}^r_i(t,0), \]  
(3.8)
\[
\hat{J}_i^r(t, y) = \lambda_i (t - (x_i - y)^+) \sum_{k=\hat{N}(t-(x_i-y)^+)}^{\hat{N}(t)} d_k, \quad \hat{J}_i(t) = \hat{J}_i^r(t, 0),
\]
(3.9)

\[
\hat{X}_i^r(t, y) = \sum_{i=1}^m \hat{X}^r_i(t, y), \quad \hat{X}(t) = \sum_{i=1}^m \hat{X}_i(t), \quad \text{and } \hat{N}(t) \text{ is defined in (2.4). The processes } \hat{J}_i^r(t, y)
\]
and \( \hat{J}_i(t) \) are independent of the arrival limit processes \( \hat{A}_i(t) \) for each \( i = 1, ..., m \).

We remark that when the arrival rates are constant, that is, \( \Lambda_i(t) = \lambda_i t \), the fluid limits \( x_i^r(t, y) \)
and \( x_i^r(t) \) in (3.6) become
\[
x_i^r(t, y) = \lambda_i (t - (x_i - y)^+), \quad x_i(t) = \lambda_i (t - (x_i)^+).
\]
(3.10)

and the processes \( \hat{J}_i^r(t, y) \) and \( \hat{J}_i(t) \) in (3.9) become
\[
\hat{J}_i^r(t, y) = \lambda_i \sum_{k=\hat{N}(t-(x_i-y)^+)}^{\hat{N}(t)} d_k, \quad \hat{J}_i(t) = \lambda_i \sum_{k=\hat{N}(t-x_i)}^{\hat{N}(t)} d_k.
\]
(3.11)

As a consequence of Theorem 3.2, there exists a stochastic decomposition property for the limit process \( \hat{X}_i^r(t, y) \), that is, each of the processes \( \hat{X}_i^r(t, y) \) is equal to a sum of two independent processes, the arrival limit processes \( \hat{A}_i(t) - \hat{A}_i(t - (x_i - y)^+) \), and the jump processes due to service interruptions \( \hat{J}_i^r(t, y) \). This stochastic decomposition property also holds for \( \hat{X}_r(t, y), \hat{X}(t) \)
and \( \hat{X}(t) \). This can be intuitively explained as follows. Since the service times are deterministic, the realized service times with the interruptions are in fact the deterministic duration “inserted” with the random down times from the arrival time to service completion. The randomness from arrivals is captured in \( \hat{A}_i(t) - \hat{A}_i(t - (x_i - y)^+) \) as in [11], while the randomness from interruptions is captured in \( \hat{J}_i^r(t, y) \), which is exactly the number of down times (in the diffusion scale) during the time period to be counted in \( \hat{X}_i^r(t, y) \). This is also demonstrated in the covariance formulas below.

**Corollary 3.1.** In addition to the assumptions of Theorem 3.2, suppose that (i) the arrival rates are constant and the arrival limit process \( \hat{A} \) is a Brownian motion with mean 0 and covariance coefficient matrix \( \Sigma = (\sigma_{ij})_{i,j=1,...,m} \), where \( \sigma_{ii} = \lambda_i c_{ii}^2 \) for some constant \( c_{ii}^2 > 0 \), and \( \sigma_{ij} \geq 0 \) are some positive constants for all \( i \neq j \), and (ii) the limit counting process \( \hat{N} \) is a Poisson process with rate \( \lambda^u = 1/E[u_i] \in (0, \infty) \) and the limit up times \( u_i \) have a finite second moment. Then for each \( t > 0 \) and \( y \geq 0 \), the distribution of \( \hat{X}_i^r(t, y) \) is a sum of two independent random vectors, a multivariate Gaussian random vector \( \hat{N}(0, \Sigma^r(t, y)) \) where \( \Sigma^r_{ii}(t, y) = c_{ii}^2 x_i^r(t,y) \) and \( \Sigma^r_{ij}(t,y) = \sigma_{ij}(x_i^r(t,y) \wedge x_j^r(t,y)) \) with \( x_i^r(t,y) \) being defined in (3.10), and a random vector of compound Poisson variables, \( \hat{J}^r = (\hat{J}_1^r, ..., \hat{J}_m^r) \), which has mean \( E[\hat{J}_i^r(t,y)] = \lambda^u E[d_i] x_i^r(t,y) \), \( i = 1, ..., m \), and covariances: for \( i, j = 1, ..., m \), \( \text{Cov}(\hat{J}_i^r(t,y), \hat{J}_j^r(t,y)) = \lambda_i \lambda_j (\lambda^u((x_i - y)^+ \wedge (x_j - y)^+) E[d_i^2] - (\lambda^u)^2((x_i - y)^+ \wedge (x_j - y)^+)^2(E[d_i])^2). \)

### 4 Dependent service times with continuous distributions

In this section, we consider \( G/G/\infty \) systems with stationary and weakly dependent service times and a general continuous distribution function. Throughout this section, we assume that \( \gamma = 1/2 \) (see also Remark 4.2). We assume that the arrival processes \( A^n \) satisfy an FCLT.

**Assumption 4.1.** There exist a deterministic function \( \Lambda \in \mathbb{C} \) with density \( \lambda \in \mathbb{D} \) a.e. and a stochastic process \( \Lambda \) such that \( \hat{A}^n(t) := \sqrt{n}(n^{-1}A^n(t) - \Lambda(t)) \Rightarrow A(t) \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \).
The service times are fixed with respect to \( n \) and so is their distribution function.

**Assumption 4.2.** The sequence of service times \( \{ \eta_i : i \geq 1 \} \) is weakly dependent and constitutes a one-sided stationary sequence. \( \eta_i \)'s have the same continuous c.d.f. \( F \) with \( F(0) = 0 \), and \( E[\eta_i^2] < \infty \), satisfying \( \sum_{i=1}^{\infty} (E[(E[\eta_i \mid \mathcal{F}_k^s])^2])^{1/2} < \infty \) for \( k = 1, 2, \ldots \), where \( \mathcal{F}_k^s := \sigma\{\eta_i : 1 \leq i \leq k\} \). Moreover, the sequence \( \{ \eta_i : i \geq 1 \} \) satisfies the \( \delta \)-mixing condition, that is, \( \sum_{k=1}^{\infty} \phi_k < \infty \), where \( \phi_k := \sup\{|P(B) - P(A)| : A \in \mathcal{F}_m^s, P(A) > 0, B \in \mathcal{G}_m^s, m \geq 1\} \), with \( \mathcal{G}_k^s := \sigma\{\eta_i : i \geq k\} \). Let \( F^c := 1 - F \).

Note that Assumption 4.2 includes many interesting examples of correlated service times as studied in Pang and Whitt [26, 27, 28]. For example, customers arrive in batches (deterministic or random batch sizes) and service times within each batch are (symmetrically) correlated while service times across different batches are independent. Other examples include EARMA sequences and first-order autoregressive sequences with exponential or general marginal distributions [13, 14, 15, 31], which have been used to study correlated service times in queueing; see Remark 4.3.

We first state the following FLLN for the fluid-scaled processes \( \hat{X}_n(t, y) = n^{-1}X_n(t, y) \) and \( \hat{X}_n(t) = n^{-1}X_n(t) \). The proof follows directly from Theorem 4.2.

**Theorem 4.1.** Under Assumptions 2.1, 4.1 and 4.2, \((\hat{X}_n^r(t, y), \hat{X}_n(t)) \Rightarrow (\hat{X}^r(t, y), \hat{X}(t)) \) in \( (\mathbb{D}, M_1) \times (\mathbb{D}, M_1) \) as \( n \to \infty \), where \( \hat{X}^r(t, y) = \int_0^t F^c(t - s + y) d\Lambda(s) \) and \( \hat{X}(t) = \hat{X}^r(t, 0) \) for \( t, y \geq 0 \).

Define the diffusion-scaled processes \( \hat{X}_n^r(t, y) = \sqrt{n}(\hat{X}_n^r(t, y) - \hat{X}^r(t, y)) \) and \( \hat{X}_n(t) = \sqrt{n}(\hat{X}_n(t) - \hat{X}(t)) \). We next state the FCLT for the diffusion-scaled processes \( \hat{X}_n^r(t, y) \) and \( \hat{X}_n(t) \).

**Theorem 4.2.** Under Assumptions 2.1, 4.1 and 4.2,

\[(\hat{X}_n^r(t, y), \hat{X}_n(t)) \Rightarrow (\hat{X}^r(t, y), \hat{X}(t)) \text{ in } (\mathbb{D}, M_1) \times (\mathbb{D}, M_1) \text{ as } n \to \infty, \tag{4.1}\]

where

\[
\hat{X}_1^r(t, y) = \hat{X}_1(t, y) + \hat{X}_2^r(t, y) + \hat{X}_3^r(t, y), \quad \hat{X}(t) = \hat{X}_1(t) + \hat{X}_2(t) + \hat{X}_3(t), \tag{4.2}
\]

\[
\hat{X}_1(t, y) = \int_0^t F^c(t - s + y) d\hat{\Lambda}(s), \tag{4.3}
\]

\[
\hat{X}_2(t, y) = \int_0^t \int_0^\infty \mathbf{1}(x > t - s + y) d\hat{K}(\Lambda(s), x), \tag{4.4}
\]

\[
\hat{X}_3(t, y) = \int_0^t (\hat{J}(t) - \hat{J}(s)) \lambda(s) dF^c(t - s + y), \tag{4.5}
\]

\[
\hat{X}_1(t) = \hat{X}_1(t, 0), \quad \hat{X}_2(t) = \hat{X}_2^r(t, 0), \quad \hat{X}_3(t) = \hat{X}_3^r(t, 0), \tag{4.6}
\]

the process \( \hat{J}(t) \) is defined in (2.7), the process \( \hat{K}(t, x) \) is a two-parameter continuous Gaussian process with mean zero and covariance function

\[
\text{Cov}(\hat{K}(t, x), \hat{K}(s, y)) = (t \wedge s) \Gamma_K(x, y), \quad t, s, x, y \geq 0, \tag{4.7}
\]

\[
\Gamma_K(x, y) = (F(x \wedge y) - F(x)F(y)) + \Gamma_k^c(x, y) < \infty, \quad x, y \geq 0, \tag{4.8}
\]

\[
\Gamma_k^c(x, y) = \sum_{k=2}^{\infty} (E[\gamma_1(x)\gamma_k(y)] + E[\gamma_1(y)\gamma_k(x)]), \quad x, y \geq 0, \tag{4.9}
\]

with \( \gamma_k(x) := \mathbf{1}(\eta_k \leq x) - F(x) \) for \( x \geq 0 \) and \( k \geq 1 \), and the integral in (4.4) with respect to the process \( \hat{K} \) is defined in the sense of mean-square integrals. The three processes \( \hat{X}_i^r(t, y), i = 1, 2, 3, \) are mutually independent, and so are the three processes \( \hat{X}_i(t), i = 1, 2, 3 \).
When the service times are i.i.d., the covariance function of \(\hat{K}(t,x)\) in (4.7)-(4.9) becomes 
\[
\text{Cov}(\hat{K}(t,x), \hat{K}(s,y))) = (t \land s)(F(x \land y) - F(x)F(y)),
\]
t, s, x, y \geq 0, and the process \(\hat{K}(t,x)\) is called a Kiefer process \([18, 25]\).

When \(\hat{N}(t)\) is Poisson, the jump process \(\hat{J}(t)\) in (2.7) is a compound Poisson process. If, in addition, the arrival rate is constant, by the stationarity of \(\hat{J}(t)\) and integration by parts, we can represent the process \(\hat{X}_2(t,y)\) in (4.5) as
\[
\hat{X}_2^n(t,y) = -\lambda \int_0^t \hat{J}(s)dF^c(s+y) = -\lambda \hat{J}(t)F^c(t+y) + \lambda \int_0^t F^c(s+y)d\hat{J}(s)
\]
\[
= -\lambda F^c(t+y)J([0,t] \times \mathbb{R}_+) + \lambda \int_0^t \int_0^\infty xF^c(s+y)d\hat{J}(ds,dx),
\]
where \(\hat{J}(s,x)\) is a Poisson random measure defined on \([0, \infty) \times \mathbb{R}_+\), with intensity \(\lambda^ud(s-x)\), where \(\lambda^u = 1/E[u_i] \in (0, \infty)\) and \(G(\cdot)\) is the distribution function of the limiting down times \(\{d_k : k \geq 1\}\).

**Remark 4.1.** There is a stochastic decomposition property for limit process \(\hat{X}_1(t,y)\): the variability from the arrival process is captured in \(\hat{X}_1(t,y)\), the variability and correlation of service times are captured in \(\hat{X}_2(t,y)\), and the impact of interruptions is captured in \(\hat{X}_3(t,y)\).

**Remark 4.2.** In Theorems 4.1–4.2, we have assumed that \(\gamma = 1/2\). This is the only scaling when a proper limit can be established to capture the variabilities from service times, the processes \(\hat{X}_1(t,y)\) and \(\hat{X}_2(t,y)\) in (4.4) and (4.6), respectively. This is because the FCLT for the sequential empirical processes driven by the service times requires \(\sqrt{n}\) scaling, see Lemma 6.1. When \(\gamma \in (0, 1/2)\), no limit exists for \((\hat{X}_1(t,y), \hat{X}_n(t))\), while when \(\gamma > 1/2\), an FCLT can be established for \((\hat{X}_1(t,y), \hat{X}_n(t))\), which will have the same limits in (4.2) with \(\hat{X}_2^n(t,y) = \hat{X}_2(t) \equiv 0\).

When the arrival limit process is a Brownian motion and the limiting counting process \(\hat{N}(t)\) is Poisson, we can characterize the transient and stationary distribution of the limit processes as in the following corollary. Its proof follows from direct calculations and is thus omitted.

**Corollary 4.1.** Under the Assumptions of Theorem 4.2, if, in addition, \(\hat{A}(t) = c_a B(\Lambda(t))\) for some \(c_a > 0\), \(\Lambda(t) = \int_0^t \lambda(s)ds\) and a standard Brownian motion \(B(t)\), and the limit counting process \(\hat{N}(t)\) is a Poisson process with rate \(\lambda^u = 1/E[u_i] \in (0, \infty)\) and the limit up times \(u_i\) have a finite second moment, the processes \(\hat{X}_1(t,y)\) and \(\hat{X}_2(t,y)\) have mean \(E[\hat{X}_1(t,y)] = \lambda^u E[d_1] \int_0^t (t-s)\lambda(s)dF^c(t-s+y)\), \(E[\hat{X}_2(t,y)] = E[\hat{X}_2(t,0)]\), and variance functions
\[
\text{Var}(\hat{X}_1(t,y)) = \int_0^t \lambda(s)\left( F^c(t+y-s) + (c_a^2 - 1)(F^c(t+y-s))^2 + \Gamma^c_K(t+y-s) \right)ds
\]
\[
+ \lambda^u E[d_1^2] \int_0^t \int_0^t \int_0^t (t-s \lor u)\lambda(s)\lambda(u)dF^c(t-s+y)dF^c(t-u+y),
\]
and \(\text{Var}(\hat{X}(t)) = \text{Var}(\hat{X}_1(t,0))\). When the arrival rate is constant, we obtain
\[
E[\hat{X}_1(\infty,y)] = \lambda \lambda^u E[d_1] \int_y^\infty (s-y)dF^c(s), \ E[\hat{X}_2(\infty)] = E[\hat{X}_1(\infty)] = \lambda E[\eta_1] \lambda^u E[d_1], \text{ and}
\]
\[
\text{Var}(\hat{X}_1(\infty,y)) = \lambda \int_y^\infty (F^c(s) + (c_a^2 - 1)((F^c(s))^2 + \Gamma^c_K(s,s)))ds
\]
Proof of Theorem 3.1. First, by (3.1) and (3.4), for \( n \), \( P \) and \( D \), Lemma 5.1.\]

§ support. We will need the following lemmas 5.2–5.5, whose proofs are in §7. In this section, we prove Theorems 3.1 and 3.2, when the service times are i.i.d. and have a finite autoregressive process, DAR(1), that is, \( \eta = \zeta_{i-1}\eta_{i-1} + (1 - \zeta_{i-1})\bar{\eta}_i, \ i \geq 2 \), where \( \{\zeta_i : i \geq 1\} \) is a sequence of i.i.d. Bernoulli random variables with \( P(\zeta_i = 1) = p \in (0,1) \) and \( \{\bar{\eta}_i : i \geq 2\} \) is a sequence of i.i.d. random variables with distribution \( F \). Then as shown in [26], the term \( \Gamma^r_K(s, s) \) in (4.11) is equal to \( \frac{2p}{1-p}F(s)F^r(s) \). Our contribution lies in the additional variabilities caused by service interruptions, characterized by the last terms in (4.10) and (4.11), which depend on the second moment of the limiting down times.

5 Proofs for service times with finite support

In this section, we prove Theorems 3.1 and 3.2, when the service times are i.i.d. and have a finite support. We will need the following lemmas 5.2–5.5, whose proofs are in §7.

Lemma 5.1. Let \((S, r)\) be a Banach space, and \( x_n, x \in \mathbb{D}([0, T], S) \) and \( x \in \mathbb{C}([0, T], S) \) for \( n \geq 1 \) and \( T > 0 \). The following is a necessary and sufficient condition for \( x_n \to x \) in the space \( \mathbb{D}([0, T], S) \) endowed with the Skorohod \( M_1 \) topology: whenever \( t_n \to t \) as \( n \to \infty \) for \( t_n, t \in [0, T] \), \( r(x_n(t_n), x(t)) \to 0 \) as \( n \to \infty \).

Proof. See, e.g., Section 3.6 in [7] and Lemma 2.1 in [33].

Lemma 5.2. Suppose that \( (x_n(t), y_n(t, s)) \to (x(t), y(t, s)) \) in \( (\mathbb{D}, M_1) \times (\mathbb{D}, J_1) \) as \( n \to \infty \) and \( y(t, s) \in \mathbb{C}_c \). Define \( z_n(t, s) := x_n(y_n(t, s)) \) and \( z(t, s) := x(y(t, s)) \) and denote \( z_n(t) := z_n(t, \cdot) \) and \( z(t) := z(t, \cdot) \). Suppose that \( y_n(t, s) \) and \( y(t, s) \) are nondecreasing in \( t \) and strictly increasing in \( s \). Then \( z_n(t, s) \) and \( z(t, s) \) are both in \( \mathbb{D}_D \), \( z_n(t) \) and \( z(t) \) are both continuous in \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \), and \( z_n \to z \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \).

Lemma 5.3. If \( x(t) \to x(t) \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \), then \( x(t) \to x(t) \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \).

Lemma 5.4. Suppose that \( (x^n(t), y^n(t, s)) \to (x(t), y(t, s)) \) in \( (\mathbb{D}, M_1) \times (\mathbb{D}, M_1) \) as \( n \to \infty \), and \( x(t) \) and \( y(t) \) do not have any common discontinuity points. Then \( x^n(t) + y^n(t, s) \to x(t) + y(t, s) \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \).

Proof of Theorem 3.1. First, by (3.1) and (3.4), for \( i = 1, \ldots, m \), we can write \( \bar{X}_{n,i}^r(t, y) \) as the following: for \( t \geq 0 \) and \( 0 \leq y < x_i \),

\[
\bar{X}_{n,i}^r(t, y) = \bar{A}_i^r(t) - \bar{A}_i^r(U^n(t) - (x_i - y))
\]

\[
= \bar{A}_i^r(t) - \bar{A}_i^r \left( U^n(t) - (x_i - y) + \sum_{k=1}^{N^n(U^n(t) - (x_i - y))} d_k^n \right),
\] (5.1)

and for \( t \geq 0 \) and \( y \geq x_i, \bar{X}_{n,i}^r(t, y) = 0 \). Thus, the main focus is on proving the convergence of \( \bar{X}_{n,i}^r(t, y) \) for \( t \geq 0 \) and \( 0 \leq y < x_i, i = 1, \ldots, m \). In the following of this proof (convergences in
Suppose that $\hat{N}_n \Rightarrow \hat{N}$ in $(\mathbb{D}, J_1)$ as $n \to \infty$. This, together with (2.5) in Lemma 2.1, implies that

$$
\left( U^n(t) - (x_i - y), \hat{N}_n^n(U^n(t) - (x_i - y)), i = 1, \ldots, m \right) \Rightarrow \left( t - (x_i - y), \hat{N}(t - (x_i - y)), i = 1, \ldots, m \right)
$$

in the in $((\mathbb{D}_m)_2, J_1)$ as $n \to \infty$. Now by Assumption 2.1 and the convergence in (5.3),

$$
\left( \hat{N}_n^n(U^n(t) - (x_i - y)) \sum_{k=1}^{n} d_k^n, i = 1, \ldots, m \right) \Rightarrow (0, i = 1, \ldots, m),
$$

in $((\mathbb{D}), J_1)$ as $n \to \infty$.

By Assumption 3.1 on the arrival processes and Lemma 5.2, we obtain

$$
\left( \hat{A}_i^n \left( U^n(t) - (x_i - y) + \sum_{k=1}^{n} d_k^n, i = 1, \ldots, m \right) \right) \Rightarrow (A_i(t - (x_i - y)), i = 1, \ldots, m)
$$

in $((\mathbb{D}_m), M_1)$ as $n \to \infty$. To apply Lemma 5.2, it is easy to see that the sample paths of the processes $U^n(t) - (x_i - y) + \sum_{k=1}^{n} d_k^n$ and the functions $t - (x_i - y)^+$ are nondecreasing in $t$ and strictly increasing in $y$ for $y \geq 0$, for each $i = 1, \ldots, m$.

Now, by applying the continuous mapping theorem, Lemmas 5.3 and 5.4, and by combining the convergence of $\bar{X}_{n,i}(t, y)$ for $t \geq 0$ and $y \geq x_i$, we obtain the convergence of $\bar{X}_n(t, y)$. The convergence of $\bar{X}_n(t, y)$, and thus that of $\bar{X}_n(t)$, follow directly from the relationship between the total count process and the two-parameter process in (2.2). It can also be proved directly by applying the continuous mapping theorem to the composition and addition mappings in the space $(\mathbb{D}, M_1)$.

We then prove Theorem 3.2. For the proof, we need the following lemma.

**Lemma 5.5.** Suppose that $x(t, s) \in \mathbb{D}$ as a function of $t$ is continuous in $\mathbb{D}([0, \infty), (\mathbb{D}, L_1))$ and $y_n(t, s) \to y(t, s)$ in $(\mathbb{D}, M_1)$, and $y(t, s)$ as a function of $t$ is continuous in $\mathbb{D}([0, \infty), (\mathbb{D}, L_1))$. Then $x(t, s)y_n(t, s) \to x(t, s)y(t, s)$ in $(\mathbb{D}, M_1)$ as $n \to \infty$.

**Proof of Theorem 3.2.** First, by (5.1), for each $i = 1, \ldots, m$, we can write the processes $\bar{X}_{n,i}(t, y)$ as the following: for $t \geq 0$ and $0 \leq y < x_i$,

$$
\bar{X}_{n,i}(t, y) = \bar{A}_i^n(t) - \hat{A}_i^n \left( U^n(t) - (x_i - y) + \sum_{k=1}^{n} d_k^n \right) + \hat{J}_{n,i}(t, y),
$$

(5.7)
where
\[
\hat{J}_{n,i}^r(t, y) = n^\gamma \left( \Lambda_i(t - (x_i - y)) - \Lambda_i \left( U^n(t) - (x_i - y) + \sum_{k=1}^{\hat{N}_n^u(U^n(t) - (x_i - y))} d_k^n \right) \right),
\]
(5.8)
and for \( t \geq 0 \) and \( y \geq x_i \), \( \hat{X}_{n,i}^r(t, y) = 0 \). As in the proof of Theorem 3.1, we only need to focus on proving the convergence of \( \hat{X}_{n,i}^r(t, y) \) for \( t \geq 0 \) and \( 0 \leq y < x_i \), \( i = 1, ..., m \), and in the following of the proof (convergences in (5.9)-(5.13)), the space \( \mathbb{D}_\mathbb{D} \) is restricted to functions \( z(t, y) \) with \( t \in [0, \infty) \) and \( y \in [0, x_i) \) for each \( i = 1, ..., m \).

By Assumption 3.1, the convergence in (5.5) and Lemma 5.2, we obtain
\[
\left( \hat{A}_i(t - (x_i - y)), i = 1, ..., m \right) \Rightarrow \left( \hat{A}_i(t - (x_i - y)), i = 1, ..., m \right)
\]
(5.9)
in \( \left( (\mathbb{D}_\mathbb{D})_m, M_1 \right) \) as \( n \to \infty \). Again, note that the sample paths of the processes \( U^n(t) - (x_i - y) + \sum_{k=1}^{\hat{N}_n^u(U^n(t) - (x_i - y))} d_k^n \) and the functions \( t - (x_i - y) \) are continuous as functions of \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \), and are nondecreasing in \( t \) and strictly increasing in \( y \) for \( y \in [0, x_i) \), for \( i = 1, ..., m \).

Next, we focus on the processes \( \hat{J}_{n,i}^r(t, y) \). We note that the continuity assumption of \( \Lambda_i \) is necessary because if \( t - (x_i - y)^+ \) is a discontinuity point of \( \Lambda_i \), the processes \( \hat{J}_{n,i}^r(t, y) \) in (5.8) will converge to infinity at those discontinuity points. In (5.8), we apply Taylor expansion and obtain
\[
\hat{J}_{n,i}^r(t, y) = \lambda_i(t - (x_i - y)) \hat{Z}_{n,i}^r(t, y) + o(1/n^\gamma),
\]
(5.10)
where
\[
\hat{Z}_{n,i}^r(t, y) = \hat{D}^n(t) - \sum_{k=1}^{\hat{N}_n^u(U^n(t) - (x_i - y))} (n^\gamma d_k^n).
\]
(5.11)
By Assumption 2.1, the convergence of \( \hat{N}_n^u \) in (5.2), Lemma 2.1, we can easily show that
\[
\left( \sum_{k=1}^{\hat{N}_n^u(U^n(t) - (x_i - y))} (n^\gamma d_k^n), i = 1, ..., m \right) \Rightarrow \left( \sum_{k=1}^{\hat{N}(t - (x_i - y))} d_k, i = 1, ..., m \right)
\]
(5.12)
in \( \left( (\mathbb{D}_\mathbb{D})_m, J_1 \right) \) as \( n \to \infty \).

Thus, by Assumption 2.1, the convergence of \( \hat{D}^n(t) \) in (2.6) of Lemma 2.1, and by Lemma 5.4, we obtain that
\[
\left( \hat{Z}_{n,i}^r(t, y), i = 1, ..., m \right) \Rightarrow \left( \sum_{k=\hat{N}(t - (x_i - y))} d_k, i = 1, ..., m \right)
\]
(5.13)
in \( \left( (\mathbb{D}_\mathbb{D})_m, M_1 \right) \) as \( n \to \infty \). We denote \( \hat{J}_n^r = (\hat{J}_{n,1}^r, ..., \hat{J}_{n,m}^r) \). Recall the associated processes \( J_i^r = (J_{i,1}^r, ..., J_{i,m}^r) \), where \( J_i^r \) is defined in (3.9). By Lemma 5.5, we obtain that \( \hat{J}_n^r \Rightarrow J_i^r \) in \( \left( (\mathbb{D}_\mathbb{D})_m, M_1 \right) \) as \( n \to \infty \).

Finally, by the fact that the limiting processes \( \hat{A}_i \) and \( \hat{J}_i^r \) do not have simultaneous jumps, and by the continuity of summation in Lemma 5.4, and by combining the convergence of \( \hat{X}_{n,i}^r(t, y) \) for
$t \geq 0$ and $y \geq x_i$, $i = 1, ..., m$, we obtain the convergence in (3.7). The convergence of $\hat{X}_n(t)$, and thus that of $\hat{X}_n(t)$, follow directly from the relationship between the total count process and the two-parameter processes in (2.2). In fact, a direct proof for the convergence of $X_n(t)$ and $\hat{X}_n(t)$ can be done by simply applying the continuous mapping theorem of composition, multiplication and summation in the $(\mathbb{D}, M_1)$ topology. The details are omitted for brevity. This completes the proof this theorem.

6 Proofs for dependent service times with continuous distributions

In this section, we prove Theorem 4.2. As in [18, 25, 28], we can represent the processes $\hat{X}_n^r(t, y)$ in (4.1) as Lebesgue-Stieltjes integrals with respect to the sequential empirical process driven by the sequence of service times. Let $\hat{K}_n(t, x)$ be sequential empirical process driven by the service times $\{\eta_i : i \geq 1\}$, defined by $\hat{K}_n(t, x) = \frac{1}{\sqrt{n}} \sum_{i=1}^{[nt]} (1(\eta_i \leq x) - F(x))$, $t \geq 0, \ x \geq 0$. We now state an FCLT for the processes $\hat{K}_n(t, x)$, whose proof can be found in §4.1 for Theorem 2.1 of [28].

Lemma 6.1. Under Assumption 4.2, $\hat{K}_n(t, x) \Rightarrow \hat{K}(t, x)$ in $(\mathbb{D}_c, J_1)$ as $n \to \infty$, where $\hat{K}$ is the two-parameter continuous Gaussian process defined in Theorem 4.2.

Lemma 6.2. The processes $X_n^r(t, y)$ in (4.1) can be represented as $X_n^r(t, y) = \hat{X}_{n,1}^r(t, y) + \hat{X}_{n,2}^r(t, y) + \hat{X}_{n,3}^r(t, y)$, where

$$\hat{X}_{n,1}^r(t, y) = \int_0^t F^c(U^n(t) - U^n(s) + y) d\hat{A}^n(s),$$

$$\hat{X}_{n,2}^r(t, y) = \int_0^t \int_0^\infty 1(x > U^n(t) - U^n(s) + y) d\hat{K}_n(\hat{A}^n(s), x),$$

$$\hat{X}_{n,3}^r(t, y) = \int_0^t \sqrt{n}(F^c(U^n(t) - U^n(s) + y) - F^c(t - s + y)) d\Lambda(s),$$

where the integrals in (6.1) and (6.2) with respect to $\hat{A}^n$ and $\hat{K}_n$, respectively, are all defined as Stieltjes integrals for functions of bounded variation as integrators.

The proof of this lemma follows from a direct calculation and the observation that the processes $X_n^r(t, y)$ defined in (2.1), can be written as

$$X_n^r(t, y) = \int_0^t \int_0^\infty 1(x > U^n(t) - U^n(s) + y) d\left(\sum_{i=1}^{A^n(t)} 1(\eta_i \leq x)\right).$$

We omit the details of this proof for brevity.

We first show the convergence of $\hat{X}_{n,1}^r(t, y)$. We need the following lemma, whose proof is in §7.

Lemma 6.3. For $x \in \mathbb{C}$ and $z \in \mathbb{D}$, define a mapping $\phi : \mathbb{C} \times \mathbb{D} \to \mathbb{D}_\mathbb{C}$ by

$$\phi(x, z)(t, y) = z(t) F^c(y) - \int_0^t z(s) dF^c(x(t) - x(s) + y), \quad t, y \geq 0.$$ 

Suppose that $(x_n, z_n) \to (x, z)$ in $(\mathbb{C}, ||\cdot||) \times (\mathbb{D}, M_1)$ as $n \to \infty$ where $x_n, x \in \mathbb{C}$. Then $\phi(x_n, z_n) \to \phi(x, z)$ in $(\mathbb{D}_\mathbb{D}, M_1)$ as $n \to \infty$. 

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We are ready to prove the convergence of $\hat{X}^r_{n,1}(t,y)$ and $\hat{X}_{n,1}(t)$.

**Lemma 6.4.** Under Assumptions 2.1, 4.1 and 4.2, $(\hat{X}^r_{n,1}, \hat{X}_{n,1}) \Rightarrow (\hat{X}^r_1, \hat{X}_1)$ in $(\mathbb{D}, M_1) \times (\mathbb{D}, M_1)$ as $n \to \infty$.

**Proof.** First, by (6.1) and (4.3), and integration by parts, we have

$$
\hat{X}^r_{n,1}(t,y) = \hat{A}^n(t)F^c(y) - \int_0^t \hat{A}^n(s)dF^c(U^n(t) - U^n(s) + y),
$$

$$
\hat{X}^r_1(t,y) = \hat{A}(t)F^c(y) - \int_0^t \hat{A}(s)dF^c(t - s + y).
$$

Then it is clear that $\hat{X}^r_{n,1}(t,y) = \phi(U^n, \hat{A}^n)(t,y)$, and $\hat{X}^r_1(t,y) = \phi(e, \hat{A})(t,y)$, where $e(t) \equiv t$ is the identity mapping. Recall that $U^n$ has continuous non-decreasing sample paths almost surely. Now by the continuity of summation in $(\mathbb{D}, M_1)$, Lemmas 5.3 and 6.3, and Assumption 4.2, we obtain the convergence of $\hat{X}^r_{n,1}(t,y)$. The convergence of $\hat{X}^r_{n,1}(t,y)$ follows from that of $\hat{X}^r_{n,1}(t,y)$. It can also be proved directly by showing the continuity in the $M_1$ topology of the following mapping $\psi : \mathbb{C}_1 \times \mathbb{D} \to \mathbb{D}$, defined by

$$
\psi(x, z)(t) = z(t)F^c(0) - \int_0^t z(s)dF^c(x(t) - x(s)), \quad t \geq 0.
$$

It is easy to check that the claim is true. Thus, the proof is complete. ■

We next prove the convergence of $X^r_{n,3}(t,y)$ and $\hat{X}_{n,3}(t)$.

**Lemma 6.5.** Under Assumptions 2.1, 4.1 and 4.2, $(\hat{X}^r_{n,3}, \hat{X}_{n,3}) \Rightarrow (\hat{X}^r_3, \hat{X}_3)$ in $(\mathbb{D}, M_1) \times (\mathbb{D}, M_1)$ as $n \to \infty$.

**Proof.** First, by (6.3) and Taylor expansion, we have

$$
X^r_{n,3}(t,y) = -\int_0^t (\hat{D}^n(s) - \hat{D}^n(t))\lambda(s)dF^c(t - s + y) + o(1/\sqrt{n})
$$

$$
= \hat{D}^n(t)\int_0^t \lambda(s)dF^c(t - s + y) - \int_0^t \hat{D}^n(s)\lambda(s)dF^c(t - s + y) + o(1/\sqrt{n}). \quad (6.4)
$$

For the first term on the right hand side of (6.4), by Lemma 2.1, we have $\hat{D}^n \Rightarrow \hat{J}$ in $(\mathbb{D}, M_1)$ and by a similar argument in the proof of Lemma 6.3, the function $\int_0^t \lambda(s)dF^c(t - s + y)$ as a function of $t$ is continuous in $(\mathbb{D}, (0, \infty), (\mathbb{D}, L_1))$. Thus by Lemmas 5.3 and 5.5, it follows that $\hat{D}^n(t)\int_0^t \lambda(s)dF^c(t - s + y) \Rightarrow \hat{J}(t)\int_0^t \lambda(s)dF^c(t - s + y)$ in $(\mathbb{D}, M_1)$ as $n \to \infty$. For the second term on the right hand side of (6.4), we can apply Lemmas 6.3 and 2.1. To see this, we let $z_n(s)$ and $x_n(s)$ in Lemma 6.3 be the sample paths of $\hat{D}^n(s)\lambda(s)$ and $\hat{J}(s)\lambda(s)$, respectively, and $x_n(s)$ and $x(s)$ be $e(s) \equiv s$. Thus, by the continuity of summation in $(\mathbb{D}, M_1)$, we obtain $\hat{X}^r_{n,3} \Rightarrow \hat{X}^r_3$ in $(\mathbb{D}, M_1)$.

The convergence of $\hat{X}^r_{n,3}(t,y)$ follows directly from that of $\hat{X}^r_{n,3}(t,y)$. It can also be shown directly by proving the continuity in the $M_1$ topology of the following mapping $\hat{\psi} : \mathbb{D} \to \mathbb{D}$, where $\hat{\psi}(x)(t) = x(t)\int_0^t \lambda(s)dF^c(t - s) - \int_0^t x(s)\lambda(s)dF^c(t - s), \quad t \geq 0$. It is easy to check that the claim is true. Thus, the proof is complete. ■

We now focus on the proof of the convergence of $\hat{X}^r_{n,2}(t,y)$ and $\hat{X}_{n,2}(t)$. It turns out that under the assumptions of Theorem 4.2, we can prove their convergence in the stronger Skorohod $J_1$ topology, which implies the convergence in the $M_1$ topology.
Lemma 6.6. Under Assumptions 2.1, 4.1 and 4.2,

\[(\hat{X}_{n,2}^r, \hat{X}_{n,2}) \Rightarrow (\hat{X}_2^r, \hat{X}_2) \quad \text{in} \quad (\mathbb{D}_B, J_1) \times (\mathbb{D}, J_1) \quad \text{as} \quad n \to \infty.\]  

(6.5)

Proof. Define the process \(\hat{Y}_{n,2}^r(t, y):\)

\[
\hat{Y}_{n,2}^r(t, y) = \int_0^t \int_0^\infty 1(x > t - s + y) d\hat{K}_n(\hat{A}^n(s), x), \quad t, y \geq 0.
\]

By Theorem 3.2 in [28] (note that the convergence therein is directly proved for the associated two-parameter process \(X'(t, y)\) tracking the number of customers in the system at time \(t\) that have received an amount of service less than or equal to \(t\), but the convergence of the two-parameter processes \(Y(t, y)\) is implied and can be easily obtained as in [25] by exploiting the relationship between the two-parameter processes tracking elapsed and residual times), we have \(\hat{Y}_{n,2}^r(t, y) \Rightarrow \hat{X}_2^r(t, y)\) in \((\mathbb{D}_B, J_1)\) as \(n \to \infty\), where \(\hat{X}_2^r(t, y)\) is defined in (4.4). Thus, by Theorem 3.1 in [3], it suffices to show that for each \(\epsilon > 0\) and for each \(T > 0\) and \(T' > 0,\)

\[
\lim_{n \to \infty} P\left( \sup_{t \in [0, T], y \in [0, T']} |\hat{Y}_{n,2}^r(t, y) - \hat{X}_{n,2}^r(t, y)| > \epsilon \right) = 0. \tag{6.6}
\]

We provide a sketch proof of (6.6) here. The tightness of \(\hat{X}_{n,2}^r\) follows first a straightforward extension of that for \(\hat{Y}_{n,2}^r\). We next discretize the domain into rectangles as in Definition 3.1 in [28], and denote the corresponding processes \(\hat{Y}_{n,2,k}^r(t, y)\) and \(\hat{X}_{n,2,k}^r(t, y)\) for discretization size \(k\). It is then easy to show that the finite dimensional distributions of \(\hat{Y}_{n,2,k}^r - \hat{X}_{n,2,k}^r\) converge to 0 for each \(k\) as \(n \to \infty\). Finally, by Theorem 3.2 in [3], it suffices to show that

\[
\lim_{k \to \infty} \limsup_{n \to \infty} P \left( \left| (\hat{Y}_{n,2,k}^r(t, y) - \hat{X}_{n,2,k}^r(t, y)) - (\hat{Y}_{n,2}^r(t, y) - \hat{X}_{n,2}^r(t, y)) \right| > \epsilon \right) = 0,
\]

for each \(t \geq 0\) and \(0 \leq y \leq t\). This follows from a similar argument as in the proof of Lemma 4.2 in [28], by noting the convergence \(U^n \Rightarrow e\). This completes the proof. ■

Proof of Theorem 4.2. The convergence in (4.1) follows from Lemmas 6.2, 6.4, 6.5 and 6.6 and the continuous mapping theorem. ■

7 Appendix

In this appendix, we collect the proofs for the lemmas used in §5–6.

Proof of Lemma 5.2. We first show \(z_n(t)\) is continuous in \(t\) in \(\mathbb{D}((0, \infty), (\mathbb{D}, L_1))\) for each \(n\). To see this, let \(t_k, t \in [0, T]\) for \(T > 0\), and \(t_k \to t\) as \(k \to \infty\), we need to show \(|z_n(t_k) - z_n(t)|_{L_1} \to 0\) as \(k \to \infty\), for each \(n \geq 1\), that is, for each \(Y > 0,\)

\[
\int_0^Y \left| x_n(y_n(t_k, s)) - x_n(y_n(t, s)) \right| ds \to 0 \quad \text{as} \quad k \to \infty. \tag{7.1}
\]

Let \(Disc_s(x_n, y_n)\) be the set of \(s\) values such that \(s\) is a discontinuity point of \(y_n(t, s)\) or \(y_n(t, s)\) is a discontinuity point of \(x_n\) for the above \(t \geq 0\). Since \(y_n \in \mathbb{D}_B\) is strictly increasing in \(s\), we have that there can only be countably many points in \(Disc_s(x_n, y_n)\) for the above \(t \geq 0\), which has Lebesgue measure zero.
Now for each \( s \notin Disc^c(x_n, y_n) \), we obtain that for each \( n \geq 1 \), \( |x_n(y_n(t_k, s)) - x_n(y_n(t, s))| \to 0 \) as \( k \to \infty \). Since \( x_n(y_n(t, s)) \in \mathbb{D}_D \), \( x_n(y_n(t, s)) \) is bounded in finite time intervals, and thus by the bounded convergence theorem together with the fact that \( Disc^c(x, y_n) \) has Lebesgue measure zero, we obtain the convergence in (7.1). Similarly, \( z(t) \) is also continuous in \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \).

Finally, we prove \( z_n \to z \) in \( (\mathbb{D}_D, M_1) \) as \( n \to \infty \). Since \( z_n(t) \) and \( z(t) \) are continuous in \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \), by Lemma 5.1, it suffices to show that for any \( T > 0 \) and any sequence \( t_n, t \in [0, T] \) satisfying \( t_n \to t \) as \( n \to \infty \), \( \|z_n(t_n) - z(t)\|_{L_1} \to 0 \) as \( n \to \infty \), that is, for each \( Y > 0 \),

\[
\int_0^Y |x_n(y_n(t_n, s)) - x(y(t, s))| \, ds \to 0 \quad \text{as} \quad n \to \infty. 
\]  

(7.2)

Let \( Disc^c(x) \) be the set of \( s \) values such that \( y(t, s) \) is a discontinuity point of \( x \) for the above \( t \geq 0 \). Since \( y \) is nondecreasing in \( t \) and strictly increasing in \( s \), there can only be countably many points in \( Disc^c(x) \), which has Lebesgue measure zero.

Now for each \( s \notin Disc^c(x) \), since \( y_n \to y \) in the space \( \mathbb{D}_D \) endowed with the Skorohod \( J_1 \) topology (and equivalently with the uniform topology due to the fact that \( y \in C_C \)), by Proposition 3.6.5 in [7], we have \( |y_n(t_n, s) - y(t, s)| \to 0 \) as \( n \to \infty \). Thus, by the convergence \( x_n \to x \) in \( (\mathbb{D}, M_1) \), we have that \( x_n(t) \to x(t) \) uniformly at each continuity point of \( x \), and since \( s \notin Disc^c(x) \), we obtain that \( |x_n(y_n(t_n, s)) - y(x, s)| \to 0 \) as \( n \to \infty \). This, together with the fact that \( Disc^c(x) \) has Lebesgue measure zero, implies that (7.2) holds and thus, the lemma is proved. ■

Proof of Lemma 5.3. Since \( x_n(t) \) and \( x(t) \) are constant functions when they are regarded functions in \( \mathbb{D}_D \), the same parametric representations for \( x_n \) and \( x \) used for the convergence of \( x_n(t) \to x(t) \) in \( (\mathbb{D}, M_1) \) can be also used for the convergence of \( x_n(t) \to x(t) \) in \( (\mathbb{D}_D, M_1) \). We omit the details for brevity. ■

Proof of Lemma 5.4. The continuity of summation in the \( M_1 \) topology follows from Lemma 5.3 and Theorem III.3.1 in [29]. ■

Proof of Lemma 5.5. Due to the continuity of \( x \) and \( y \) as functions of \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \), by Lemma 5.1, it suffices to prove that for any \( T > 0 \) and any sequence \( t_n, t \in [0, T] \) satisfying \( t_n \to t \) as \( n \to \infty \), we have \( \|x(t_n)y_n(t_n) - x(t)y(t)\|_{L_1} \to 0 \) as \( n \to \infty \). Recall that \( x(t_n) := x(t_n, \cdot) \) and similarly for \( y_n(t_n), x(t) \) and \( y(t) \). Now,

\[
\|x(t_n)y_n(t_n) - x(t)y(t)\|_{L_1} \\
\leq \|x(t_n)y_n(t_n) - x(t_n)y(t)\|_{L_1} + \|x(t_n)y(t) - x(t)y(t)\|_{L_1} \\
\leq \|x(t_n)\|_{L_1} \|y_n(t_n) - y(t)\|_{L_1} + \|y(t)\|_{L_1} \|x(t_n) - x(t)\|_{L_1} \to 0.
\]

The convergence follows from the facts that \( \text{sup}_n \|x(t_n)\|_{L_1} < \infty, \|y(t)\|_{L_1} < \infty, y_n(t, s) \to y(t, s) \) in \( (\mathbb{D}_D, M_1) \), and the continuity property of \( y(t, s) \) and \( x(t, s) \) as functions of \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \).

■

Proof of Lemma 6.3. The proof proceeds in four steps.

Step 1: Since \( z_n \to z \) in \( (\mathbb{D}, M_1) \) as \( n \to \infty \), by Lemma 5.3 we have \( z_n \to z \) in \( (\mathbb{D}_D, M_1) \) as \( n \to \infty \). Since \( F \) is continuous, by Lemma 5.5, we have that \( z_n(t)F^c(y) \to z(t)F^c(y) \) in \( (\mathbb{D}_D, M_1) \) as \( n \to \infty \).

Step 2: We claim that \( \int_0^t z(s)dF^c(x(t) - x(s) + y) \) is continuous in \( t \) in \( \mathbb{D}([0, \infty), (\mathbb{D}, L_1)) \). Suppose \( t_n, t \in [0, T] \) for \( T > 0 \) and \( t_n \to t \) as \( n \to \infty \), for each \( y \geq 0 \), we have

\[
\left| \int_0^{t_n} dF^c(x(t_n) - x(s) + y) - \int_0^t dF^c(x(t) - x(s) + y) \right| \\
\leq \left| \int_0^t dF^c(x(t_n) - x(s) + y) - \int_0^t dF^c(x(t_n) - x(s) + y) \right|
\]

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Since $x$ and $F$ are both continuous functions, $F^c(x(t_n) - x(s) + y) - F^c(x(t) - x(s) + y) \to 0$ as $n \to \infty$, and thus the same holds for the first integral on the right hand side of (7.3). The second integral goes to zero because $t_n \to t$ and the continuity of $F$. (So far, we have proved the pointwise convergence). Given $Y > 0$, and $T > 0$ such that $t_n, t \in [0, T]$, $z(t) \in \mathbb{D}$ implies $z(t)$ is bounded on $[0, T]$. So we have $\int_0^Y \int_0^t z(s)dF^c(x(t) - x(s) + y)dy$ is bounded for each $Y > 0$. By (7.3) and the bounded convergence theorem, we obtain
\[
\int_0^Y \int_0^t z(s)dF^c(x(t_n) - x(s) + y) - \int_0^t z(s)dF^c(x(t) - x(s) + y) \to 0,
\]
as $n \to \infty$. Thus, we have proved the claim. A similar argument also shows that $\int_0^t z_n(s)dF^c(x_n(t) - x(s) + y)$ is continuous in $t$ in $\mathbb{D}([0, \infty), (\mathbb{D}, L_1))$ since we assume $x_n \in \mathbb{C}$.

Step 3: We prove that
\[
\int_0^t z_n(s)dF^c(x_n(t) - x_n(s) + y) \to \int_0^t z(s)dF^c(x(t) - x(s) + y)
\]
as $n \to \infty$. By the claim proved in Step 2, by Lemma 5.1, it suffices to prove that for each $T > 0$ and $Y > 0$, for $t_n, t \in [0, T]$ and any sequence $t_n \to t$ as $n \to \infty$, we have
\[
\int_0^Y \int_0^{t_n} z_n(s)dF^c(x_n(t_n) - x_n(s) + y) - \int_0^t z(s)dF^c(x(t) - x(s) + y) \to 0
\]
as $n \to \infty$. We first prove pointwise convergence and then $L_1$ convergence. For each $y \geq 0$,
\[
\int_0^{t_n} z_n(s)dF^c(x_n(t_n) - x_n(s) + y) - \int_0^t z(s)dF^c(x(t) - x(s) + y)
\]
\[
\leq \left| \int_0^{t_n} z_n(s)d(F^c(x_n(t_n) - x_n(s) + y) - F^c(x(t) - x(s) + y)) \right|
\]
\[
+ \left| \int_0^{t_n} (z_n(s) - z(s))dF^c(x(t) - x(s) + y) \right| + \left| \int_t^{t_n} z(s)dF(x(t) - x(s) + y) \right|.
\]
The first integral on the right hand side of (7.5) converges to zero as $n \to \infty$ because of the continuity of $F$ and $x_n(t_n) \to x(t)$ uniformly as $n \to \infty$ (Recall that $x_n$ and $x$ are continuous). Since $z_n(t) \to z(t)$ in $(\mathbb{D}, M_1)$, the set $\{s \in [0, \infty) : |z_n(s) - z(s)| \to 0\}$ has Lebesgue measure 0, and thus the second integral on the right hand side of (7.5) goes to zero as $n \to \infty$. It is evident that the third integral on the right hand side of (7.5) also goes to zero as $n \to \infty$. Thus, we have proved that for each $y > 0$,
\[
\int_0^{t_n} z_n(s)dF^c(x_n(t) - x_n(s) + y) - \int_0^t z(s)dF^c(x(t) - x(s) + y) \to 0
\]
as $n \to \infty$. Given any $Y > 0$, we obtain the convergence in (7.4) by applying the bounded convergence theorem.

Step 4: By the continuity of summation in $(\mathbb{D}, M_1)$, the conclusion of the lemma holds.

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