

---

# A Computation and Communication Efficient Method for Distributed Nonconvex Problems in the Partial Participation Setting

---

Anonymous Author(s)

Affiliation

Address

email

## Abstract

1 We present a new method that includes three key components of distributed opti-  
2 mization and federated learning: variance reduction of stochastic gradients, partial  
3 participation, and compressed communication. We prove that the new method has  
4 optimal oracle complexity and state-of-the-art communication complexity in the  
5 partial participation setting. Regardless of the communication compression feature,  
6 our method successfully combines variance reduction and partial participation: we  
7 get the optimal oracle complexity, never need the participation of all nodes, and do  
8 not require the bounded gradients (dissimilarity) assumption.

## 9 1 Introduction

10 Federated and distributed learning have become very popular in recent years (Konečný et al., 2016;  
11 McMahan et al., 2017). The current optimization tasks require much computational resources and  
12 machines. Such requirements emerge in machine learning, where massive datasets and computations  
13 are distributed between cluster nodes (Lin et al., 2017; Ramesh et al., 2021). In federated learning,  
14 nodes, represented by mobile phones, laptops, and desktops, do not send their data to a server due to  
15 privacy and their huge number (Ramaswamy et al., 2019), and the server remotely orchestrates the  
16 nodes and communicates with them to solve an optimization problem.

17 As in classical optimization tasks, one of the main current challenges is to find **computationally**  
18 **efficient** optimization algorithms. However, the nature of distributed problems induces many other  
19 (Kairouz et al., 2021), including i) **partial participation** of nodes in algorithm steps: due to stragglers  
20 (Li et al., 2020) or communication delays (Vogels et al., 2021), ii) **communication bottleneck**: even  
21 if a node participates, it can be costly to transmit information to a server or other nodes (Alistarh  
22 et al., 2017; Ramesh et al., 2021; Kairouz et al., 2021; Sapio et al., 2019; Narayanan et al., 2019). It  
23 is necessary to develop a method that considers these problems.

## 24 2 Optimization Problem

25 Let us consider the nonconvex distributed optimization problem

$$\min_{x \in \mathbb{R}^d} \left\{ f(x) := \frac{1}{n} \sum_{i=1}^n f_i(x) \right\}, \quad (1)$$

26 where  $f_i : \mathbb{R}^d \rightarrow \mathbb{R}$  is a smooth nonconvex function for all  $i \in [n] := \{1, \dots, n\}$ . The full  
27 information about function  $f_i$  is stored on  $i^{\text{th}}$  node. The communication between nodes is maintained  
28 in the parameters server fashion (Kairouz et al., 2021): we have a server that receives compressed

information from nodes, updates a state, and broadcasts an updated model.<sup>1</sup> Since we work in the nonconvex world, our goal is to find an  $\varepsilon$ -solution ( $\varepsilon$ -stationary point) of (1): a (possibly random) point  $\hat{x} \in \mathbb{R}^d$ , such that  $\mathbb{E}[\|\nabla f(\hat{x})\|^2] \leq \varepsilon$ .

We consider three settings:

1. **Gradient Setting.** The  $i^{\text{th}}$  node has only access to the gradient  $\nabla f_i : \mathbb{R}^d \rightarrow \mathbb{R}^d$  of function  $f_i$ . Moreover, the following assumptions for the functions  $f_i$  hold.

**Assumption 1.** *There exists  $f^* \in \mathbb{R}$  such that  $f(x) \geq f^*$  for all  $x \in \mathbb{R}$ .*

**Assumption 2.** *The function  $f$  is  $L$ -smooth, i.e.,  $\|\nabla f(x) - \nabla f(y)\| \leq L\|x - y\|$  for all  $x, y \in \mathbb{R}^d$ .*

**Assumption 3.** *The functions  $f_i$  are  $L_i$ -smooth for all  $i \in [n]$ . Let us define  $\hat{L}^2 := \frac{1}{n} \sum_{i=1}^n L_i^2$ .<sup>2</sup>*

2. **Finite-Sum Setting.** The functions  $\{f_i\}_{i=1}^n$  have the finite-sum form

$$f_i(x) = \frac{1}{m} \sum_{j=1}^m f_{ij}(x), \quad \forall i \in [n], \quad (2)$$

where  $f_{ij} : \mathbb{R}^d \rightarrow \mathbb{R}$  is a smooth nonconvex function for all  $j \in [m]$ . We assume that Assumptions 1, 2 and 3 hold and the following assumption.

**Assumption 4.** *The function  $f_{ij}$  is  $L_{ij}$ -smooth for all  $i \in [n], j \in [m]$ . Let  $L_{\max} := \max_{i \in [n], j \in [m]} L_{ij}$ .*

3. **Stochastic Setting.** The function  $f_i$  is an expectation of a stochastic function,

$$f_i(x) = \mathbb{E}_{\xi} [f_i(x; \xi)], \quad \forall i \in [n], \quad (3)$$

where  $f_i : \mathbb{R}^d \times \Omega_{\xi} \rightarrow \mathbb{R}$ . For a fixed  $x \in \mathbb{R}$ ,  $f_i(x; \xi)$  is a random variable over some distribution  $\mathcal{D}_i$ , and, for a fixed  $\xi \in \Omega_{\xi}$ ,  $f_i(x; \xi)$  is a smooth nonconvex function. The  $i^{\text{th}}$  node has only access to a stochastic gradients  $\nabla f_i(\cdot; \xi_{ij})$  of the function  $f_i$  through the distribution  $\mathcal{D}_i$ , where  $\xi_{ij}$  is a sample from  $\mathcal{D}_i$ . We assume that Assumptions 1, 2 and 3 hold and the following assumptions.

**Assumption 5.** *For all  $i \in [n]$  and for all  $x \in \mathbb{R}^d$ , the stochastic gradient  $\nabla f_i(x; \xi)$  is unbiased and has bounded variance, i.e.,  $\mathbb{E}_{\xi} [\nabla f_i(x; \xi)] = \nabla f_i(x)$ , and  $\mathbb{E}_{\xi} [\|\nabla f_i(x; \xi) - \nabla f_i(x)\|^2] \leq \sigma^2$ , where  $\sigma^2 \geq 0$ .*

**Assumption 6.** *For all  $i \in [n]$  and for all  $x, y \in \mathbb{R}$ , the stochastic gradient  $\nabla f_i(x; \xi)$  satisfies the mean-squared smoothness property, i.e.,  $\mathbb{E}_{\xi} [\|\nabla f_i(x; \xi) - \nabla f_i(y; \xi)\|^2] \leq L_{\sigma}^2 \|x - y\|^2$ .*

We compare algorithms using the *oracle complexity*, i.e., the number of (stochastic) gradients that each node has to calculate to get  $\varepsilon$ -solution, and the *communication complexity*, i.e., the number of bits that each node has to send to the server to get  $\varepsilon$ -solution.

## 2.1 Unbiased Compressors

We use the concept of unbiased compressors to alleviate the communication bottleneck. The unbiased compressors quantize and/or sparsify vectors that the nodes send to the server.

**Definition 1.** A stochastic mapping  $\mathcal{C} : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is an *unbiased compressor* if there exists  $\omega \in \mathbb{R}$  such that

$$\mathbb{E}[\mathcal{C}(x)] = x, \quad \mathbb{E}[\|\mathcal{C}(x) - x\|^2] \leq \omega \|x\|^2, \quad (4)$$

for all  $x \in \mathbb{R}^d$ .

<sup>1</sup>Note that this strategy can be used in peer-to-peer communication, assuming that the server is an abstraction and all its algorithmic steps are performed on each node.

<sup>2</sup>Note that  $L \leq \hat{L}$ ,  $\hat{L} \leq L_{\max}$ , and  $\hat{L} \leq L_{\sigma}$ .

Table 1: Summary of methods that solve the problem (1) in the stochastic setting (3). Abbr.: *VR* (Variance Reduction) = Does a method have the optimal oracle complexity  $\mathcal{O}\left(\frac{\sigma^2}{\varepsilon} + \frac{\sigma}{\varepsilon^{3/2}}\right)$ ? *PP* (Partial Participation) = Does a method support partial participation from Section 2.2? *CC* = Does a method have the communication complexity equals to  $\mathcal{O}\left(\frac{d}{\sqrt{n}\varepsilon}\right)$ ?

Method	VR	PP	CC	Limitations
<b>SPIDER, SARAH, PAGE, STORM</b> (Fang et al., 2018; Nguyen et al., 2017) (Li et al., 2021a; Cutkosky and Orabona, 2019)	✓	✗	✗	—
<b>MARINA</b> (Gorbunov et al., 2021)	✓	✗ <sup>(a)</sup>	✓ <sup>(b)</sup>	Suboptimal convergence rate (see (Tyurin and Richtárik, 2023)).
<b>FedPAGE</b> (Zhao et al., 2021b)	✗	✗ <sup>(a)</sup>	✗	Suboptimal oracle complexity $\mathcal{O}\left(\frac{\sigma^2}{\varepsilon^2}\right)$ .
<b>FRECON</b> (Zhao et al., 2021a)	✗	✓	✓	—
<b>FedAvg</b> (McMahan et al., 2017; Karimireddy et al., 2020b)	✗	✓	✗	Bounded gradients (dissimilarity) assumption of $f_i$ .
<b>SCAFFOLD</b> (Karimireddy et al., 2020b)	✗	✓	✗	Suboptimal convergence rate <sup>(c)</sup> .
<b>MIME<sup>(c)</sup></b> (Karimireddy et al., 2020a)	✗ <sup>(d)</sup>	✓	✗	Calculates full gradient. Bounded gradients (dissimilarity) assumption of $f_i$ . Suboptimal oracle complexity $\mathcal{O}(1/\varepsilon^{3/2})$ in the setting (2).
<b>CE-LSGD (for Partial Participation)<sup>(c)</sup></b> (Patel et al., 2022) (concurrent work)	✓	✓	✗	Bounded gradients (dissimilarity) assumption of $f_i$ . Suboptimal oracle complexity $\mathcal{O}(1/\varepsilon^{3/2})$ in the setting (2).
<b>DASHA</b> (Tyurin and Richtárik, 2023)	✓ ✗	✗ or ✓	✓ ✓	—
<b>DASHA-PP</b> (new)	✓	✓	✓	—

<sup>(a)</sup> **MARINA** and **FedPAGE**, with a small probability, require the participation of all nodes so that they can not support partial participation from Section 2.2. Moreover, these methods provide suboptimal oracle complexities.

<sup>(b)</sup> On average, **MARINA** provides the compressed communication mechanism with complexity  $\mathcal{O}\left(\frac{d}{\sqrt{n}\varepsilon}\right)$ . However, with a small probability, this method sends non-compressed vectors.

<sup>(c)</sup> Note that **MIME** and **CE-LSGD** can not be directly compared with **DASHA-PP** because **MIME** and **CE-LSGD** consider the online version of the problem (1), and require more strict assumptions.

<sup>(d)</sup> Although **MIME** obtains the convergence rate  $\mathcal{O}\left(\frac{1}{\varepsilon^{3/2}}\right)$  of a variance reduced method, it requires the calculation of the full (exact) gradients.

<sup>(e)</sup> It can be seen when  $\sigma^2 = 0$ . Let us consider the  $s$ -nice sampling of the nodes, then **SCAFFOLD** requires  $\mathcal{O}\left(\frac{n^{3/2}}{\varepsilon s^{3/2}}\right)$  communication rounds to get  $\varepsilon$ -solution, while **DASHA-PP** requires  $\mathcal{O}\left(\frac{\sqrt{n}}{\varepsilon s}\right)$  communication rounds (see Theorem 4 with  $\omega = 0$ ,  $b = \frac{p_a}{2-p_a}$ , and  $p_a = \frac{s}{n}$ ).

We denote a set of stochastic mappings that satisfy Definition 1 as  $\mathbb{U}(\omega)$ . In our methods, the nodes make use of unbiased compressors  $\{\mathcal{C}_i\}_{i=1}^n$ . The community developed a large number of unbiased compressors, including **RandK** (see Definition 5) (Beznosikov et al., 2020; Stich et al., 2018), Adaptive sparsification (Wangni et al., 2018) and Natural compression and dithering (Horváth et al., 2019a). We are aware of correlated compressors by Szlendak et al. (2021) and quantizers by Suresh et al. (2022) that help in the homogeneous regimes, but in this work, we are mainly concentrated on generic heterogeneous regimes, though, for simplicity, assume the independence of the compressors.

**Assumption 7.**  $\mathcal{C}_i \in \mathbb{U}(\omega)$  for all  $i \in [n]$ , and the compressors are statistically independent.

## 2.2 Nodes Partial Participation Assumptions

We now try to formalize the notion of partial participation. Let us assume that we have  $n$  events  $\{i^{\text{th}} \text{ node is participating}\}$  with the following properties.

**Assumption 8.** The partial participation of nodes has the following distribution: exists constants  $p_a \in (0, 1]$  and  $p_{aa} \in [0, 1]$ , such that

Table 2: Summary of methods that solve the problem (1) in the finite-sum setting (2). Abbr.: VR (Variance Reduction) = Does a method have the optimal oracle complexity  $\mathcal{O}\left(m + \frac{\sqrt{m}}{\varepsilon}\right)$ ? PP and CC are defined in Table 1.

Method	VR	PP	CC	Limitations
SPIDER, PAGE (Fang et al., 2018; Li et al., 2021a)	✓	✗	✗	—
MARINA (Gorbunov et al., 2021)	✓	✗ <sup>(a)</sup>	✓ <sup>(b)</sup>	Suboptimal convergence rate (see (Tyurin and Richtárik, 2023)).
ZeroSARAH (Li et al., 2021b)	✓	✓	✗	Only homogeneous regime, i.e., the functions $f_i$ are equal.
FedPAGE (Zhao et al., 2021b)	✗	✗ <sup>(a)</sup>	✗	Suboptimal oracle complexity $\mathcal{O}\left(\frac{m}{\varepsilon}\right)$ .
DASHA (Tyurin and Richtárik, 2023)	✓	✗	✓	—
DASHA-PP (new)	✓	✓	✓	—

<sup>(a)</sup>, <sup>(b)</sup> : see Table 1.

- 78 1.  $\text{Prob}(i^{\text{th}} \text{ node is participating}) = p_a, \quad \forall i \in [n],$
- 79 2.  $\text{Prob}(i^{\text{th}} \text{ and } j^{\text{th}} \text{ nodes are participating}) = p_{aa},$
- 80  $\text{for all } i \neq j \in [n].$
- 81 3.  $p_{aa} \leq p_a^2, \tag{5}$

82 and these events from different communication rounds are independent.

83 We are not fighting for the full generality and believe that more complex sampling strategies can  
84 be considered in the analysis. For simplicity, we settle upon Assumption 8. Standard partial  
85 participation strategies, including  $s$ -nice sampling, where the server chooses uniformly  $s$  nodes  
86 without replacement ( $p_a = s/n$  and  $p_{aa} = s(s-1)/n(n-1)$ ), and independent participation, where each  
87 node independently participates with probability  $p_a$  (due to independence, we have  $p_{aa} = p_a^2$ ), satisfy  
88 Assumption 8. In the literature,  $s$ -nice sampling is one of the most popular strategies (Zhao et al.,  
89 2021a; Richtárik et al., 2021; Reddi et al., 2020; Konečný et al., 2016).

### 90 3 Motivation and Related Work

91 The main goal of our paper is to develop a method for the nonconvex distributed optimization that will  
92 include three key features: variance reduction of stochastic gradients, compressed communication,  
93 and partial participation. We now provide an overview of the literature (see also Table 1 and Table 2).

#### 94 1. Variance reduction of stochastic gradients

95 It is important to consider finite-sum (2) and stochastic (3) settings because, in machine learning  
96 tasks, either the number of local functions  $m$  is huge or the functions  $f_i$  is an expectation of a  
97 stochastic function due to the batch normalization (Ioffe and Szegedy, 2015) or random augmentation  
98 (Goodfellow et al., 2016), and it is infeasible to calculate the full gradients analytically. Let us recall  
99 the results from the nondistributed optimization. In the gradient setting, the optimal oracle complexity  
100 is  $\mathcal{O}(1/\varepsilon)$ , achieved by the vanilla gradient descent (GD) (Carmon et al., 2020; Nesterov, 2018). In  
101 the finite-sum setting and stochastic settings, the optimal oracle complexities are  $\mathcal{O}\left(m + \frac{\sqrt{m}}{\varepsilon}\right)$  and  
102  $\mathcal{O}\left(\frac{\sigma^2}{\varepsilon} + \frac{\sigma}{\varepsilon^{3/2}}\right)$  (Fang et al., 2018; Li et al., 2021a; Arjevani et al., 2019), accordingly, achieved by  
103 methods SPIDER, SARAH, PAGE, and STORM from (Fang et al., 2018; Nguyen et al., 2017; Li et al.,  
104 2021a; Cutkosky and Orabona, 2019).

#### 105 2. Compressed communication

106 In distributed optimization (Ramesh et al., 2021; Xu et al., 2021), lossy communication compression  
107 can be a powerful tool to increase the communication speed between the nodes and the server.

Different types of compressors are considered in the literature, including unbiased compressors (Alistarh et al., 2017; Beznosikov et al., 2020; Szlendak et al., 2021), contractive (biased) compressors (Richtárik et al., 2021), 3PC compressors (Richtárik et al., 2022). We will focus on unbiased compressors because methods DASHA and MARINA (Tyurin and Richtárik, 2023; Szlendak et al., 2021; Gorbunov et al., 2021) that employ unbiased compressors provide the current theoretical state-of-the-art (SOTA) communication complexities.

Many methods analyzed optimization methods with the unbiased compressors (Alistarh et al., 2017; Mishchenko et al., 2019; Horváth et al., 2019b; Gorbunov et al., 2021; Tyurin and Richtárik, 2023). In the gradient setting, the methods MARINA and DASHA by Gorbunov et al. (2021) and Tyurin and Richtárik (2023) establish the current SOTA communication complexity, each method needs  $\frac{1+\omega/\sqrt{n}}{\varepsilon}$  communication rounds to get an  $\varepsilon$ -solution. In the finite-sum and stochastic settings, the current SOTA communication complexity is attained by the DASHA method, while maintaining the optimal oracle complexities  $\mathcal{O}\left(m + \frac{\sqrt{m}}{\varepsilon\sqrt{n}}\right)$  and  $\mathcal{O}\left(\frac{\sigma^2}{\varepsilon n} + \frac{\sigma}{\varepsilon^{3/2}n}\right)$  per node.

### 3. Partial participation

From the beginning of federated learning era, the partial participation has been considered to be the essential feature of distributed optimization methods (McMahan et al., 2017; Konečný et al., 2016; Kairouz et al., 2021). However, previously proposed methods have limitations: i) methods MARINA and FedPAGE from (Gorbunov et al., 2021; Zhao et al., 2021b) still require synchronization of all nodes with a small probability. ii) in the stochastic settings, methods FedAvg, SCAFFOLD, and FRECON with the partial participation mechanism (McMahan et al., 2017; Karimireddy et al., 2020b; Zhao et al., 2021a) provide results without variance reduction techniques from (Fang et al., 2018; Li et al., 2021a; Cutkosky and Orabona, 2019) and, therefore, get suboptimal oracle complexities. Note that FRECON and DASHA reduce the variance *only from compressors* (in the partial participation and stochastic setting). iii) in the finite-sum setting, the ZeroSARAH method by Li et al. (2021b) focuses on the homogeneous regime only (the functions  $f_i$  are equal). iv) The MIME method by Karimireddy et al. (2020a) and the CE-LSGD method (for Partial Participation) by the concurrent paper (Patel et al., 2022) consider the online version of the problem (1). Therefore, MIME and CE-LSGD (for Partial Participation) require stricter assumptions, including the bounded inter-client gradient variance assumption. In the finite-sum setting (2), MIME and CE-LSGD obtain a suboptimal oracle complexity  $\mathcal{O}(1/\varepsilon^{3/2})$  while, in the full participation setting, it is possible to get the complexity  $\mathcal{O}(1/\varepsilon)$ .

## 4 Contributions

We propose a new method DASHA-PP for the nonconvex distributed optimization.

- As far as we know, this is the first method that includes three key ingredients of federated learning methods: *variance reduction of stochastic gradients*, *compressed communication*, and *partial participation*.
- Moreover, this is the first method that combines *variance reduction of stochastic gradients* and *partial participation* flawlessly: i) it gets the optimal oracle complexity ii) does not require the participation of all nodes iii) does not require the bounded gradients assumption of the functions  $f_i$ .
- We prove convergence rates and show that this method has *the optimal oracle complexity and the state-of-the-art communication complexity in the partial participation setting*. Moreover, in our work, we observe a nontrivial side-effect from mixing the variance reduction of stochastic gradients and partial participation. It is a general problem not related to our methods or analysis that we discuss in Section C.

## 5 Algorithm Description and Main Challenges Towards Partial Participation

We now present DASHA-PP (see Algorithm 1), a family of methods to solve the optimization problem (1). When we started investigating the problem, we took DASHA as a baseline method for two reasons: the family of algorithms DASHA provides the current state-of-the-art communication complexities in the *non-partial participation* setting, and, unlike MARINA, it does not send non-compressed gradients and does not synchronize all nodes. Let us briefly discuss the main idea of DASHA, its problem in the *partial participation* setting, and why the refinement of DASHA is not an exercise.

---

**Algorithm 1 DASHA-PP**

---

- 1: **Input:** starting point  $x^0 \in \mathbb{R}^d$ , stepsize  $\gamma > 0$ , momentum  $a \in (0, 1]$ , momentum  $b \in (0, 1]$ , probability  $p_{\text{page}} \in (0, 1]$  (only in **DASHA-PP-PAGE**), batch size  $B$  (only in **DASHA-PP-PAGE**, **DASHA-PP-FINITE-MVR** and **DASHA-PP-MVR**), probability  $p_a \in (0, 1]$  that a node is *participating*<sup>(a)</sup>, number of iterations  $T \geq 1$
  - 2: Initialize  $g_i^0 \in \mathbb{R}^d$ ,  $h_i^0 \in \mathbb{R}^d$  on the nodes and  $g^0 = \frac{1}{n} \sum_{i=1}^n g_i^0$  on the server
  - 3: Initialize  $h_{ij}^0 \in \mathbb{R}^d$  on the nodes and take  $h_i^0 = \frac{1}{m} \sum_{j=1}^m h_{ij}^0$  (only in **DASHA-PP-FINITE-MVR**)
  - 4: **for**  $t = 0, 1, \dots, T - 1$  **do**
  - 5:    $x^{t+1} = x^t - \gamma g^t$
  - 6:   Broadcast  $x^{t+1}, x^t$  to all *participating*<sup>(a)</sup> nodes
  - 7:   **for**  $i = 1, \dots, n$  in parallel **do**
  - 8:     **if**  $i^{\text{th}}$  node is *participating*<sup>(a)</sup> **then**
  - 9:       Calculate  $k_i^{t+1}$  using Algorithm 2, 3, 4 or 5
  - 10:        $h_i^{t+1} = h_i^t + \frac{1}{p_a} k_i^{t+1}$
  - 11:        $m_i^{t+1} = C_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right)$
  - 12:        $g_i^{t+1} = g_i^t + m_i^{t+1}$
  - 13:       Send  $m_i^{t+1}$  to the server
  - 14:     **else**
  - 15:        $h_{ij}^{t+1} = h_{ij}^t$  (only in **DASHA-PP-FINITE-MVR**)
  - 16:        $h_i^{t+1} = h_i^t, \quad g_i^{t+1} = g_i^t, \quad m_i^{t+1} = 0$
  - 17:     **end if**
  - 18:   **end for**
  - 19:    $g^{t+1} = g^t + \frac{1}{n} \sum_{i=1}^n m_i^{t+1}$
  - 20: **end for**
  - 21: **Output:**  $\hat{x}^T$  chosen uniformly at random from  $\{x^t\}_{k=0}^{T-1}$
- (a): For the formal description see Section 2.2.
- 

---

**Algorithm 2** Calculate  $k_i^{t+1}$  for **DASHA-PP** in the gradient setting. See line 9 in Alg. 1

---

- 1:  $k_i^{t+1} = \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))$
- 

---

**Algorithm 3** Calculate  $k_i^{t+1}$  for **DASHA-PP-PAGE** in the finite-sum setting. See line 9 in Alg. 1

---

- 1: Generate a random set  $I_i^t$  of size  $B$  from  $[m]$  *with replacement*
  - 2:  $k_i^{t+1} = \begin{cases} \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)), \\ \text{with probability } p_{\text{page}} \text{ on all } \textit{participating} \text{ nodes,} \\ \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)), \\ \text{with probability } 1 - p_{\text{page}} \text{ on all } \textit{participating} \text{ nodes} \end{cases}$
- 

---

**Algorithm 4** Calc.  $k_i^{t+1}$  for **DASHA-PP-FINITE-MVR** in the finite-sum setting. See line 9 in Alg. 1

---

- 1: Generate a random set  $I_i^t$  of size  $B$  from  $[m]$  *without replacement*
  - 2:  $k_{ij}^{t+1} = \begin{cases} \frac{m}{B} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))), & j \in I_i^t, \\ 0, & j \notin I_i^t \end{cases}$
  - 3:  $h_{ij}^{t+1} = h_{ij}^t + \frac{1}{p_a} k_{ij}^{t+1}$
  - 4:  $k_i^{t+1} = \frac{1}{m} \sum_{j=1}^m k_{ij}^{t+1}$
- 

---

**Algorithm 5** Calculate  $k_i^{t+1}$  for **DASHA-PP-MVR** in the stochastic setting. See line 9 in Alg. 1

---

- 1: Generate i.i.d. samples  $\{\xi_{ij}^{t+1}\}_{j=1}^B$  of size  $B$  from  $\mathcal{D}_i$ .
  - 2:  $k_i^{t+1} = \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1}) - b \left( h_i^t - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1}) \right)$
-



158 In fact, **DASHA** supports the partial participation of nodes *in the gradient setting*. Since the nodes  
159 only do the following steps (see full algorithm in Algorithm 6):

$$g_i^{t+1} = g_i^t + C_i (\nabla f_i(x^{t+1}) - (1-a)\nabla f_i(x^t) - ag_i^t).$$

160 The partial participation mechanism (independent participation from Section 2.2) can be easily  
161 implemented here if we redefine the compressor and use another one instead:

$$C_i^p := \begin{cases} \frac{1}{p}C_i, & \text{with pr. } p_a, \\ 0, & \text{with pr. } 1 - p_a. \end{cases} \Rightarrow g_i^{t+1} = \begin{cases} g_i^t + \frac{1}{p_a}C_i (\nabla f_i(x^{t+1}) - (1-a)\nabla f_i(x^t) - ag_i^t), & p_a \\ g_i^t, & 1 - p_a. \end{cases}$$

162 With probability  $1 - p$ , a node does not update  $g_i$  and does not send anything to the server. The main  
163 observation is that we can do this trick since  $g_i^{t+1}$  depends only on the vectors  $x^{t+1}$ ,  $x^t$ , and  $g_i^t$ .

164 However, we focus our attention on partial participation *in the finite-sum and stochastic settings*.  
165 Consider the nodes' steps in **DASHA-MVR** (see Algorithm 7) that is designed for the stochastic setting:

$$h_i^{t+1} = \nabla f_i(x^{t+1}; \xi_i^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})), \quad (6)$$

$$g_i^{t+1} = g_i^t + C_i (h_i^{t+1} - h_i^t - a(g_i^t - h_i^t)). \quad (7)$$

166 Even if we use the same trick for (7), we still have to update (6) in every iteration of the algorithm  
167 since  $g_i^{t+1}$  additionally depends on  $h_i^{t+1}$  and  $h_i^t$ . In other words, if a node does not update  $g_i$  and  
168 does not send anything to the server, it still has to update  $h_i$ , what is impossible without the points  
169  $x^{t+1}$  and  $x^t$ . One of the main challenges was to “guess” how to generalize (6) and (7) to the partial  
170 participation setting. We now provide a solution (**DASHA-PP-MVR** with the batch size  $B = 1$ ):

$$h_i^{t+1} = h_i^t + \frac{1}{p_a}k_i^{t+1}, \quad k_i^{t+1} = \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})), \quad (8)$$

$$g_i^{t+1} = g_i^t + C_i \left( \frac{1}{p_a}k_i^{t+1} - \frac{a}{p_a}(g_i^t - h_i^t) \right) \text{ with pr. } p_a, \text{ and } h_i^{t+1} = h_i^t, \quad g_i^{t+1} = g_i^t \text{ with pr. } 1 - p_a.$$

171 Now both control variables  $g_i^t$  and  $h_i^t$  do not change with the probability  $1 - p_a$ . However, when the  
172  $i^{\text{th}}$  node participates, this required changing the update rule of  $g_i^{t+1}$  and  $h_i^{t+1}$  to make the proof work.

173 The theoretical analysis of the new algorithm became more complicated: while in **DASHA**, the  
174 randomness from compressors is independent of the randomness from stochastic gradients (see (6)  
175 and (7)). In (8) (see also main Algorithm 1), these two randomnesses are coupled by the randomness  
176 from the partial participation. Going deeper into details, one can compare Lemma I.2 from (Tyurin  
177 and Richtárik, 2023) and Lemma 5. The former lemma does not use the knowledge about the update  
178 rule of  $h_i^{t+1}$ , works with one expectation  $\mathbb{E}_C[\cdot]$ , uses only (4), (16), and (17). The latter lemma  
179 additionally explicitly uses that  $h_i^{t+1} = h_i^t + \frac{1}{p_a}k_i^{t+1}$ , surgically copes with the expectations  $\mathbb{E}_C[\cdot]$   
180 and  $\mathbb{E}_{p_a}[\cdot]$  (for instance, it is not trivial in each order one should apply the expectations), and uses the  
181 sampling lemma (Lemma 1). The same reasoning applies to other part of the analysis.

182 At the first reading of the proofs, we suggest the reader follow the proof of Theorem 2 in the gradient  
183 setting, which takes a small part of the paper. Although the appendix seems to be dense and large, the  
184 size is justified by the fact that we consider four different settings and PL-condition ( $4 \times 2$  tracks of  
185 the proofs. The theory is designed so that the proofs do not repeat steps of each other and use one  
186 framework).

## 187 6 Theorems

188 We now present the convergence rates theorems of **DASHA-PP** in different settings. We will compare  
189 the theorems with the results of the current state-of-the-art methods, **MARINA** and **DASHA**, that work  
190 in the full participation setting. Suppose that **MARINA** or **DASHA** converges to  $\varepsilon$ -solution after  $T$   
191 communication rounds. Then, ideally, we would expect the convergence of the new algorithms to  
192  $\varepsilon$ -solution after up to  $T/p_a$  communication rounds due to the partial participation constraints<sup>3</sup>. The  
193 detailed analysis of the algorithms under Polyak-Łojasiewicz condition we provide in Section F. Let  
194 us define  $\Delta_0 := f(x^0) - f^*$ .

<sup>3</sup>We check this numerically in Section A.

## 195 6.1 Gradient Setting

196 **Theorem 2.** Suppose that Assumptions 1, 2, 3, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_a}{2-p_a}$ ,

$$\gamma \leq \left( L + \left[ \frac{48\omega(2\omega+1)}{np_a^2} + \frac{16}{np_a^2} \left( 1 - \frac{p_{aa}}{p_a} \right) \right]^{1/2} \hat{L} \right)^{-1},$$

197 and  $g_i^0 = h_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP), then  $\mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \leq \frac{2\Delta_0}{\gamma T}$ .

198 Let us recall the convergence rate of MARINA or DASHA, the number of communication rounds to get  
 199  $\varepsilon$ -solution equals  $\mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{\sqrt{n}} \hat{L} \right] \right)$ , while the rate of DASHA-PP equals  $\mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega+1}{p_a \sqrt{n}} \hat{L} \right] \right)$ .  
 200 Up to Lipschitz constants factors, we get the degeneration up to  $1/p_a$  factor due to the partial  
 201 participation. This is the perfect result since each worker sends useful information with the probability  
 202  $p_a$ .

## 203 6.2 Finite-Sum Setting

204 **Theorem 3.** Suppose that Assumptions 1, 2, 3, 4, 7, and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_{page} p_a}{2-p_a}$ ,  
 205 probability  $p_{page} \in (0, 1]$ ,

$$\gamma \leq \left( L + \left[ \frac{48\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{(1-p_{page})L_{\max}^2}{B} \right) + \frac{16}{np_a^2 p_{page}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{page})L_{\max}^2}{B} \right) \right]^{1/2} \right)^{-1}$$

206 and  $g_i^0 = h_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-PAGE) then  $\mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \leq$   
 207  $\frac{2\Delta_0}{\gamma T}$ .

208 We now choose  $p_{page}$  to balance heavy full gradient and light mini-batch calculations. Let us define  
 209  $\mathbb{1}_{p_a} := \sqrt{1 - \frac{p_{aa}}{p_a}} \in [0, 1]$ . Note that if  $p_a = 1$  then  $p_{aa} = 1$  and  $\mathbb{1}_{p_a} = 0$ .

210 **Corollary 1.** Let the assumptions from Theorem 3 hold and  $p_{page} = B/(m+B)$ . Then DASHA-PP-PAGE  
 211 needs

$$T := \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{1}{p_a} \sqrt{\frac{m}{n}} \left( \frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right) \quad (9)$$

212 communication rounds to get an  $\varepsilon$ -solution and the expected number of gradient calculations per  
 213 node equals  $\mathcal{O}(m + BT)$ .

214 The convergence rate the rate of the current state-of-the-art method DASHA-PAGE without partial  
 215 participation equals  $\mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{\sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \sqrt{\frac{m}{n}} \frac{L_{\max}}{B} \right] \right)$ . Let us closer compare it with (9).  
 216 As expected, we see that the second term w.r.t.  $\omega$  degenerates up to  $1/p_a$ . Surprisingly, the third term  
 217 w.r.t.  $\sqrt{m/n}$  can degenerate up to  $\sqrt{B}/p_a$  when  $\hat{L} \approx L_{\max}$ . Hence, in order to keep degeneration up to  
 218  $1/p_a$ , one should take the batch size  $B = \mathcal{O}(L_{\max}^2/\hat{L}^2)$ . This interesting effect we analyze separately  
 219 in Section C. The fact that the degeneration is up to  $1/p_a$  we check numerically in Section A.

220 In the following corollary, we consider RandK compressors (see Definition 5) and show that with  
 221 the particular choice of parameters, up to the Lipschitz constants factors, DASHA-PP-PAGE gets the  
 222 optimal oracle complexity and SOTA communication complexity. The choice of the compressor is  
 223 driven by simplicity, and the following analysis can be used for other unbiased compressors.

224 **Corollary 2.** Suppose that assumptions of Corollary 1 hold,  $B \leq \min \left\{ \frac{1}{p_a} \sqrt{\frac{m}{n}}, \frac{L_{\max}^2}{\frac{1}{p_a^2} \hat{L}^2} \right\}$ <sup>4</sup>, and we  
 225 use the unbiased compressor RandK with  $K = \Theta(Bd/\sqrt{m})$ . Then the communication complexity of  
 226 Algorithm 1 is

$$\mathcal{O} \left( d + \frac{L_{\max} \Delta_0 d}{p_a \varepsilon \sqrt{n}} \right), \quad (10)$$

---

<sup>4</sup>If  $\mathbb{1}_{p_a} = 0$ , then  $\frac{L_{\max}^2}{\frac{1}{p_a^2} \hat{L}^2} = +\infty$



227 and the expected number of gradient calculations per node equals

$$\mathcal{O}\left(m + \frac{L_{\max}\Delta_0\sqrt{m}}{p_a\varepsilon\sqrt{n}}\right). \quad (11)$$

228 The convergence rate of **DASHA-PP-FINITE-MVR** is provided in Section E.5. The conclusions are the  
229 same for the method.

### 230 6.3 Stochastic Setting

231 We define  $h^t := \frac{1}{n} \sum_{i=1}^n h_i^t$ .

232 **Theorem 4.** Suppose that Assumptions 1, 2, 3, 5, 6, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
233  $b \in \left(0, \frac{p_a}{2-p_a}\right]$ ,  $\gamma \leq \left(L + \left[\frac{48\omega(2\omega+1)}{np_a^2} \left(\hat{L}^2 + \frac{(1-b)^2 L_\sigma^2}{B}\right) + \frac{12}{np_a b} \left(\left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 + \frac{(1-b)^2 L_\sigma^2}{B}\right)\right]^{1/2}\right)^{-1}$ , and  
234  $g_i^0 = h_i^0$  for all  $i \in [n]$  in Algorithm 1 (**DASHA-PP-MVR**). Then

$$\begin{aligned} \mathbb{E}\left[\left\|\nabla f(\hat{x}^T)\right\|^2\right] &\leq \frac{1}{T} \left[ \frac{2\Delta_0}{\gamma} + \frac{2}{b} \left\|h^0 - \nabla f(x^0)\right\|^2 + \left(\frac{32b\omega(2\omega+1)}{np_a^2} + \frac{4\left(1 - \frac{p_{aa}}{p_a}\right)}{np_a}\right) \left(\frac{1}{n} \sum_{i=1}^n \left\|h_i^0 - \nabla f_i(x^0)\right\|^2\right) \right] \\ &+ \left(\frac{48b^2\omega(2\omega+1)}{p_a^2} + \frac{12b}{p_a}\right) \frac{\sigma^2}{nB}. \end{aligned}$$

235 In the next corollary, we choose momentum  $b$  and initialize vectors  $h_i^0$  to get  $\varepsilon$ -solution. Let us define

$$236 \mathbb{1}_{p_a} := \sqrt{1 - \frac{p_{aa}}{p_a}} \in [0, 1].$$

237 **Corollary 3.** Suppose that assumptions from Theorem 4 hold, momentum  $b =$   
238  $\Theta\left(\min\left\{\frac{p_a}{\omega} \sqrt{\frac{n\varepsilon B}{\sigma^2}}, \frac{p_a n \varepsilon B}{\sigma^2}\right\}\right)$ ,  $\frac{\sigma^2}{n\varepsilon B} \geq 1$ , and  $h_i^0 = \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0)$  for all  $i \in [n]$ ,  
239 and batch size  $B_{\text{init}} = \Theta\left(\frac{\sqrt{p_a B}}{b}\right)$ , then Algorithm 1 (**DASHA-PP-MVR**) needs

$$T := \mathcal{O}\left(\frac{\Delta_0}{\varepsilon} \left[L + \frac{\omega}{p_a \sqrt{n}} \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}}\right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left(\frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B}\right)\right] + \frac{\sigma^2}{\sqrt{p_a n \varepsilon B}}\right)$$

240 communication rounds to get an  $\varepsilon$ -solution and the number of stochastic gradient calculations per  
241 node equals  $\mathcal{O}(B_{\text{init}} + BT)$ .

242 The convergence rate of the **DASHA-SYNC-MVR**, the state-of-the-art method without partial participa-  
243 tion, equals  $\mathcal{O}\left(\frac{\Delta_0}{\varepsilon} \left[L + \frac{\omega}{\sqrt{n}} \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}}\right) + \frac{\sigma}{\sqrt{\varepsilon n}} \frac{L_\sigma}{B}\right] + \frac{\sigma^2}{n\varepsilon B}\right)$ . Similar to Section 6.2, we see that in  
244 the regimes when  $\hat{L} \approx L_\sigma$  the third term w.r.t.  $1/\varepsilon^{3/2}$  can degenerate up to  $\sqrt{B}/p_a$ . However, if we take  
245  $B = \mathcal{O}(L_\sigma^2/\hat{L}^2)$ , then the degeneration of the third term will be up to  $1/p_a$ . This effect we analyze in  
246 Section C. The fact that the degeneration is up to  $1/p_a$  we check numerically in Section A.

247 In the following corollary, we consider RandK compressors (see Definition 5) and show that with  
248 the particular choice of parameters, up to the Lipschitz constants factors, **DASHA-PP-MVR** gets the  
249 optimal oracle complexity and SOTA communication complexity of **DASHA-SYNC-MVR** method.

250 **Corollary 4.** Suppose that assumptions of Corollary 3 hold, batch size  $B \leq \min\left\{\frac{\sigma}{p_a \sqrt{\varepsilon n}}, \frac{L_\sigma^2}{\mathbb{1}_{p_a}^2 \hat{L}^2}\right\}$ ,  
251 we take RandK compressors with  $K = \Theta\left(\frac{Bd\sqrt{\varepsilon n}}{\sigma}\right)$ . Then the communication complexity equals

$$\mathcal{O}\left(\frac{d\sigma}{\sqrt{p_a} \sqrt{n\varepsilon}} + \frac{L_\sigma \Delta_0 d}{p_a \sqrt{n\varepsilon}}\right), \quad (12)$$

252 and the expected number of stochastic gradient calculations per node equals

$$\mathcal{O}\left(\frac{\sigma^2}{\sqrt{p_a} n\varepsilon} + \frac{L_\sigma \Delta_0 \sigma}{p_a \varepsilon^{3/2} n}\right). \quad (13)$$

We are aware that the initial batch size  $B_{\text{init}}$  can be suboptimal w.r.t.  $\omega$  in **DASHA-PP-MVR** in some regimes (see also (Tyurin and Richtárik, 2023)). This is a side effect of mixing the variance reduction of stochastic gradients and compression. However, Corollary 4 reveals that we can escape these regimes by choosing the parameter  $K$  of RandK compressors in a particular way. To get the complete picture, we analyze the same phenomenon under PL condition (see Section F) and provide a new method **DASHA-PP-SYNC-MVR** (see Section G).

## References

- Alistarh, D., Grubic, D., Li, J., Tomioka, R., and Vojnovic, M. (2017). QSGD: Communication-efficient SGD via gradient quantization and encoding. In *Advances in Neural Information Processing Systems (NIPS)*, pages 1709–1720.
- Arjevani, Y., Carmon, Y., Duchi, J. C., Foster, D. J., Srebro, N., and Woodworth, B. (2019). Lower bounds for non-convex stochastic optimization. *arXiv preprint arXiv:1912.02365*.
- Beznosikov, A., Horváth, S., Richtárik, P., and Safaryan, M. (2020). On biased compression for distributed learning. *arXiv preprint arXiv:2002.12410*.
- Carmon, Y., Duchi, J. C., Hinder, O., and Sidford, A. (2020). Lower bounds for finding stationary points i. *Mathematical Programming*, 184(1):71–120.
- Chang, C.-C. and Lin, C.-J. (2011). LIBSVM: a library for support vector machines. *ACM Transactions on Intelligent Systems and Technology (TIST)*, 2(3):1–27.
- Cutkosky, A. and Orabona, F. (2019). Momentum-based variance reduction in non-convex SGD. *arXiv preprint arXiv:1905.10018*.
- Fang, C., Li, C. J., Lin, Z., and Zhang, T. (2018). SPIDER: Near-optimal non-convex optimization via stochastic path integrated differential estimator. In *NeurIPS Information Processing Systems*.
- Goodfellow, I., Bengio, Y., Courville, A., and Bengio, Y. (2016). *Deep learning*, volume 1. MIT Press.
- Gorbunov, E., Burlachenko, K., Li, Z., and Richtárik, P. (2021). MARINA: Faster non-convex distributed learning with compression. In *38th International Conference on Machine Learning*.
- Horváth, S., Ho, C.-Y., Horvath, L., Sahu, A. N., Canini, M., and Richtárik, P. (2019a). Natural compression for distributed deep learning. *arXiv preprint arXiv:1905.10988*.
- Horváth, S., Kovalev, D., Mishchenko, K., Stich, S., and Richtárik, P. (2019b). Stochastic distributed learning with gradient quantization and variance reduction. *arXiv preprint arXiv:1904.05115*.
- Ioffe, S. and Szegedy, C. (2015). Batch normalization: Accelerating deep network training by reducing internal covariate shift. In *International Conference on Machine Learning*, pages 448–456. PMLR.
- Kairouz, P., McMahan, H. B., Avent, B., Bellet, A., Bennis, M., Bhagoji, A. N., Bonawitz, K., Charles, Z., Cormode, G., Cummings, R., et al. (2021). Advances and open problems in federated learning. *Foundations and Trends® in Machine Learning*, 14(1–2):1–210.
- Karimireddy, S. P., Jaggi, M., Kale, S., Mohri, M., Reddi, S. J., Stich, S. U., and Suresh, A. T. (2020a). Mime: Mimicking centralized stochastic algorithms in federated learning. *arXiv preprint arXiv:2008.03606*.
- Karimireddy, S. P., Kale, S., Mohri, M., Reddi, S., Stich, S., and Suresh, A. T. (2020b). Scaffold: Stochastic controlled averaging for federated learning. In *International Conference on Machine Learning*, pages 5132–5143. PMLR.
- Konečný, J., McMahan, H. B., Yu, F. X., Richtárik, P., Suresh, A. T., and Bacon, D. (2016). Federated learning: Strategies for improving communication efficiency. *arXiv preprint arXiv:1610.05492*.

297 Li, T., Sahu, A. K., Zaheer, M., Sanjabi, M., Talwalkar, A., and Smith, V. (2020). Federated  
298 optimization in heterogeneous networks. *Proceedings of Machine Learning and Systems*, 2:429–  
299 450.

300 Li, Z., Bao, H., Zhang, X., and Richtárik, P. (2021a). PAGE: A simple and optimal probabilistic  
301 gradient estimator for nonconvex optimization. In *International Conference on Machine Learning*,  
302 pages 6286–6295. PMLR.

303 Li, Z., Hanzely, S., and Richtárik, P. (2021b). ZeroSARAH: Efficient nonconvex finite-sum optimiza-  
304 tion with zero full gradient computation. *arXiv preprint arXiv:2103.01447*.

305 Lin, Y., Han, S., Mao, H., Wang, Y., and Dally, W. J. (2017). Deep gradient compression: Reducing  
306 the communication bandwidth for distributed training. *arXiv preprint arXiv:1712.01887*.

307 McMahan, B., Moore, E., Ramage, D., Hampson, S., and y Arcas, B. A. (2017). Communication-  
308 efficient learning of deep networks from decentralized data. In *Artificial intelligence and statistics*,  
309 pages 1273–1282. PMLR.

310 Mishchenko, K., Gorbunov, E., Takáč, M., and Richtárik, P. (2019). Distributed learning with  
311 compressed gradient differences. *arXiv preprint arXiv:1901.09269*.

312 Narayanan, D., Harlap, A., Phanishayee, A., Seshadri, V., Devanur, N. R., Ganger, G. R., Gibbons,  
313 P. B., and Zaharia, M. (2019). PipeDream: generalized pipeline parallelism for dnn training. In  
314 *Proceedings of the 27th ACM Symposium on Operating Systems Principles*, pages 1–15.

315 Nesterov, Y. (2018). *Lectures on convex optimization*, volume 137. Springer.

316 Nguyen, L., Liu, J., Scheinberg, K., and Takáč, M. (2017). SARAH: A novel method for machine  
317 learning problems using stochastic recursive gradient. In *The 34th International Conference on*  
318 *Machine Learning*.

319 Paszke, A., Gross, S., Massa, F., Lerer, A., Bradbury, J., Chanan, G., Killeen, T., Lin, Z., Gimelshein,  
320 N., Antiga, L., et al. (2019). Pytorch: An imperative style, high-performance deep learning library.  
321 In *Advances in Neural Information Processing Systems (NeurIPS)*.

322 Patel, K. K., Wang, L., Woodworth, B., Bullins, B., and Srebro, N. (2022). Towards optimal commu-  
323 nication complexity in distributed non-convex optimization. In *Advances in Neural Information*  
324 *Processing Systems*.

325 Ramaswamy, S., Mathews, R., Rao, K., and Beaufays, F. (2019). Federated learning for emoji  
326 prediction in a mobile keyboard. *arXiv preprint arXiv:1906.04329*.

327 Ramesh, A., Pavlov, M., Goh, G., Gray, S., Voss, C., Radford, A., Chen, M., and Sutskever, I. (2021).  
328 Zero-shot text-to-image generation. *arXiv preprint arXiv:2102.12092*.

329 Reddi, S., Charles, Z., Zaheer, M., Garrett, Z., Rush, K., Konečný, J., Kumar, S., and McMahan,  
330 H. B. (2020). Adaptive federated optimization. *arXiv preprint arXiv:2003.00295*.

331 Richtárik, P., Sokolov, I., and Fatkhullin, I. (2021). EF21: A new, simpler, theoretically better, and  
332 practically faster error feedback. In *Neural Information Processing Systems, 2021*.

333 Richtárik, P., Sokolov, I., Fatkhullin, I., Gasanov, E., Li, Z., and Gorbunov, E. (2022). 3PC: Three  
334 point compressors for communication-efficient distributed training and a better theory for lazy  
335 aggregation. *arXiv preprint arXiv:2202.00998*.

336 Sapio, A., Canini, M., Ho, C.-Y., Nelson, J., Kalnis, P., Kim, C., Krishnamurthy, A., Moshref, M.,  
337 Ports, D. R., and Richtárik, P. (2019). Scaling distributed machine learning with in-network  
338 aggregation. *arXiv preprint arXiv:1903.06701*.

339 Stich, S. U., Cordonnier, J.-B., and Jaggi, M. (2018). Sparsified SGD with memory. *Advances in*  
340 *Neural Information Processing Systems*, 31.

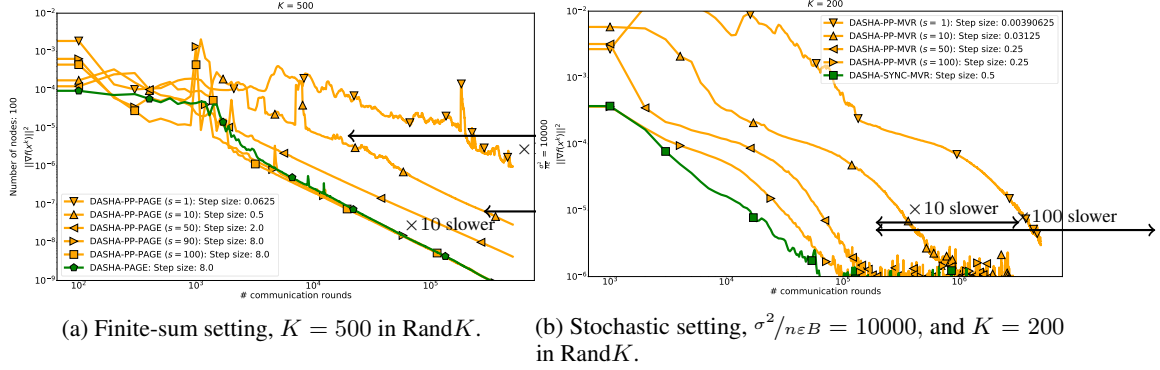
341 Suresh, A. T., Sun, Z., Ro, J. H., and Yu, F. (2022). Correlated quantization for distributed mean  
342 estimation and optimization. *arXiv preprint arXiv:2203.04925*.

- 343 Szlendak, R., Tyurin, A., and Richtárik, P. (2021). Permutation compressors for provably faster  
344 distributed nonconvex optimization. *arXiv preprint arXiv:2110.03300*.
- 345 Tyurin, A. and Richtárik, P. (2023). DASHA: Distributed nonconvex optimization with commu-  
346 nication compression and optimal oracle complexity. *International Conference on Learning*  
347 *Representations (ICLR)*.
- 348 Vogels, T., He, L., Koloskova, A., Karimireddy, S. P., Lin, T., Stich, S. U., and Jaggi, M. (2021).  
349 RelaySum for decentralized deep learning on heterogeneous data. *Advances in Neural Information*  
350 *Processing Systems*, 34.
- 351 Wangni, J., Wang, J., Liu, J., and Zhang, T. (2018). Gradient sparsification for communication-  
352 efficient distributed optimization. *Advances in Neural Information Processing Systems*, 31.
- 353 Xu, H., Ho, C.-Y., Abdelmoniem, A. M., Dutta, A., Bergou, E. H., Karatsenidis, K., Canini, M.,  
354 and Kalnis, P. (2021). Grace: A compressed communication framework for distributed machine  
355 learning. In *2021 IEEE 41st International Conference on Distributed Computing Systems (ICDCS)*,  
356 pages 561–572. IEEE.
- 357 Zhao, H., Burlachenko, K., Li, Z., and Richtárik, P. (2021a). Faster rates for compressed federated  
358 learning with client-variance reduction. *arXiv preprint arXiv:2112.13097*.
- 359 Zhao, H., Li, Z., and Richtárik, P. (2021b). FedPAGE: A fast local stochastic gradient method for  
360 communication-efficient federated learning. *arXiv preprint arXiv:2108.04755*.

361	<b>Contents</b>	
362	<b>1 Introduction</b>	<b>1</b>
363	<b>2 Optimization Problem</b>	<b>1</b>
364	2.1 Unbiased Compressors . . . . .	2
365	2.2 Nodes Partial Participation Assumptions . . . . .	3
366	<b>3 Motivation and Related Work</b>	<b>4</b>
367	<b>4 Contributions</b>	<b>5</b>
368	<b>5 Algorithm Description and Main Challenges Towards Partial Participation</b>	<b>5</b>
369	<b>6 Theorems</b>	<b>7</b>
370	6.1 Gradient Setting . . . . .	8
371	6.2 Finite-Sum Setting . . . . .	8
372	6.3 Stochastic Setting . . . . .	9
373	<b>A Numerical Verification of Theoretical Dependencies</b>	<b>15</b>
374	<b>B Original DASHA and DASHA-MVR Methods</b>	<b>16</b>
375	<b>C Problem of Estimating the Mean in the Partial Participation Setting</b>	<b>17</b>
376	<b>D Auxiliary facts</b>	<b>18</b>
377	D.1 Sampling Lemma . . . . .	18
378	D.2 Compressors Facts . . . . .	20
379	<b>E Proofs of Theorems</b>	<b>20</b>
380	E.1 Standard Lemmas in the Nonconvex Setting . . . . .	21
381	E.2 Generic Lemmas . . . . .	22
382	E.3 Proof for DASHA-PP . . . . .	25
383	E.4 Proof for DASHA-PP-PAGE . . . . .	29
384	E.5 Proof for DASHA-PP-FINITE-MVR . . . . .	40
385	E.6 Proof for DASHA-PP-MVR . . . . .	51
386	<b>F Analysis of DASHA-PP under Polyak-Łojasiewicz Condition</b>	<b>64</b>
387	F.1 Gradient Setting . . . . .	64
388	F.2 Finite-Sum Setting . . . . .	64
389	F.3 Stochastic Setting . . . . .	65
390	F.4 Proofs of Theorems . . . . .	66
391	F.4.1 Standard Lemma under Polyak-Łojasiewicz Condition . . . . .	66
392	F.4.2 Generic Lemma . . . . .	66

393	F.4.3 Proof for DASHA-PP under PL-condition . . . . .	68
394	F.4.4 Proof for DASHA-PP-PAGE under PL-condition . . . . .	70
395	F.4.5 Proof for DASHA-PP-MVR under PL-condition . . . . .	75
396	<b>G Description of DASHA-PP-SYNC-MVR</b>	<b>82</b>
397	G.1 Proof for DASHA-PP-SYNC-MVR . . . . .	84




 Figure 1: Classification task with the *real-sim* dataset.

Our main goal is to verify the dependeces from the theory. We compare **DASHA-PP** with **DASHA**. Clearly, **DASHA-PP** can not generally perform better than **DASHA**. In different settings, we verify that the bigger  $p_a$ , the closer **DASHA-PP** is to **DASHA**, i.e., **DASHA-PP** converges no slower than  $1/p_a$  times.

In all experiments, we take the *real-sim* dataset with dimension  $d = 20,958$  and the number of samples equals 72,309 from LIBSVM datasets (Chang and Lin, 2011) (under the 3-clause BSD license), and randomly split the dataset between  $n = 100$  nodes equally, ignoring residual samples. In the finite-sum setting, we solve a classification problem with functions

$$f_i(x) := \frac{1}{m} \sum_{j=1}^m \left( 1 - \frac{1}{1 + \exp(y_{ij} a_{ij}^\top x)} \right)^2,$$

where  $a_{ij} \in \mathbb{R}^d$  is the feature vector of a sample on the  $i^{\text{th}}$  node,  $y_{ij} \in \{-1, 1\}$  is the corresponding label, and  $m$  is the number of samples on the  $i^{\text{th}}$  node for all  $i \in [n]$ . In the stochastic setting, we consider functions

$$f_i(x_1, x_2) := \mathbb{E}_{j \sim [m]} \left[ -\log \left( \frac{\exp(a_{ij}^\top x_1 y_{ij})}{\sum_{y \in \{1, 2\}} \exp(a_{ij}^\top x_y y)} \right) + \lambda \sum_{y \in \{1, 2\}} \sum_{k=1}^d \frac{\{x_y\}_k^2}{1 + \{x_y\}_k^2} \right],$$

where  $x_1, x_2 \in \mathbb{R}^d$ ,  $\{\cdot\}_k$  is an indexing operation,  $a_{ij} \in \mathbb{R}^d$  is a feature of a sample on the  $i^{\text{th}}$  node,  $y_{ij} \in \{1, 2\}$  is a corresponding label,  $m$  is the number of samples located on the  $i^{\text{th}}$  node, constant  $\lambda = 0.001$  for all  $i \in [n]$ .

The code was written in Python 3.6.8 using PyTorch 1.9 (Paszke et al., 2019). A distributed environment was emulated on a machine with Intel(R) Xeon(R) Gold 6226R CPU @ 2.90GHz and 64 cores.

We use the standard setting in experiments<sup>5</sup> where all parameters except step sizes are taken as suggested in theory. Step sizes are finetuned from a set  $\{2^i \mid i \in [-10, 10]\}$ . We emulate the partial participation setting using  $s$ -nice sampling with the number of nodes  $n = 100$ . We consider the  $\text{Rand}K$  compressor and take the batch size  $B = 1$ . We plot the relation between communication rounds and values of the norm of gradients at each communication round.

In the finite-sum (Figure 1a) and in the stochastic setting (Figure 1b), we see that the bigger probability  $p_a = s/n$  to 1, the closer **DASHA-PP** to **DASHA**. Moreover, **DASHA-PP** with  $s = 10$  and  $s = 1$  converges approximately  $\times 10$  ( $= 1/p_a$ ) and  $\times 100$  ( $= 1/p_a$ ) times slower, accordingly. Our theory predicts such behavior.

<sup>5</sup>Code: <https://github.com/mysteryresearcher/dasha-partial-participation>

## 421 B Original DASHA and DASHA-MVR Methods

422 To simplify the discussion and explanation from the main part, we present the algorithms from (Tyurin  
423 and Richtárik, 2023)

---

### Algorithm 6 DASHA

---

```

1: Input: starting point  $x^0 \in \mathbb{R}^d$ , stepsize  $\gamma > 0$ , momentum  $a \in (0, 1]$ , number of iterations  $T \geq 1$ 
2: Initialize  $g_i^0 \in \mathbb{R}^d$  on the nodes and  $g^0 = \frac{1}{n} \sum_{i=1}^n g_i^0$  on the server
3: for  $t = 0, 1, \dots, T - 1$  do
4:    $x^{t+1} = x^t - \gamma g^t$ 
5:   Broadcast  $x^{t+1}, x^t$  to all participating(a) nodes
6:   for  $i = 1, \dots, n$  in parallel do
7:      $m_i^{t+1} = \mathcal{C}_i(\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - a(g_i^t - \nabla f_i(x^t)))$ 
8:      $g_i^{t+1} = g_i^t + m_i^{t+1}$ 
9:     Send  $m_i^{t+1}$  to the server
10:  end for
11:   $g^{t+1} = g^t + \frac{1}{n} \sum_{i=1}^n m_i^{t+1}$ 
12: end for
13: Output:  $\hat{x}^T$  chosen uniformly at random from  $\{x^t\}_{k=0}^{T-1}$ 

```

---



---

### Algorithm 7 DASHA-MVR (with batch size $B = 1$ )

---

```

1: Input: starting point  $x^0 \in \mathbb{R}^d$ , stepsize  $\gamma > 0$ , momentums  $a, b \in (0, 1]$ , number of iterations  $T \geq 1$ 
2: Initialize  $g_i^0 \in \mathbb{R}^d$  on the nodes and  $g^0 = \frac{1}{n} \sum_{i=1}^n g_i^0$  on the server
3: for  $t = 0, 1, \dots, T - 1$  do
4:    $x^{t+1} = x^t - \gamma g^t$ 
5:   Broadcast  $x^{t+1}, x^t$  to all participating(a) nodes
6:   for  $i = 1, \dots, n$  in parallel do
7:      $h_i^{t+1} = \nabla f_i(x^{t+1}; \xi_i^{t+1}) + (1 - b)(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})), \quad \xi_i^{t+1} \sim \mathcal{D}_i$ 
8:      $m_i^{t+1} = \mathcal{C}_i(h_i^{t+1} - h_i^t - a(g_i^t - h_i^t))$ 
9:      $g_i^{t+1} = g_i^t + m_i^{t+1}$ 
10:    Send  $m_i^{t+1}$  to the server
11:  end for
12:   $g^{t+1} = g^t + \frac{1}{n} \sum_{i=1}^n m_i^{t+1}$ 
13: end for
14: Output:  $\hat{x}^T$  chosen uniformly at random from  $\{x^t\}_{k=0}^{T-1}$ 

```

---

## 424 C Problem of Estimating the Mean in the Partial Participation Setting

425 We now provide the example to explain why the only choice of  $B = \mathcal{O}\left(\min\left\{\frac{1}{p_a}\sqrt{\frac{m}{n}}, \frac{L_{\max}^2}{p_a L^2}\right\}\right)$  and  
 426  $B = \mathcal{O}\left(\min\left\{\frac{\sigma}{p_a\sqrt{\varepsilon n}}, \frac{L_{\sigma}^2}{p_a^2 \hat{\mathcal{L}}^2}\right\}\right)$  in **DASHA-PP-PAGE** and **DASHA-PP-MVR**, accordingly, guarantees  
 427 the degeneration up to  $1/p_a$ . This is surprising, because in methods with the variance reduction of  
 428 stochastic gradients (Li et al., 2021a; Tyurin and Richtárik, 2023) we can take the size of batch size  
 429  $B = \mathcal{O}(\sqrt{\frac{m}{n}})$  and  $B = \mathcal{O}\left(\frac{\sigma}{\sqrt{\varepsilon n}}\right)$  and guarantee the optimality. Note that the smaller the batch size  
 430  $B$ , the more the server and the nodes have to communicate to get  $\varepsilon$ -solution.

431 Let us consider the task of estimating the mean of vectors in the distributed setting. Suppose that we  
 432 have  $n$  nodes, and each of them contains  $m$  vectors  $\{x_{ij}\}_{j=1}^m$ , where  $x_{ij} \in \mathbb{R}^d$  for all  $i \in [n], j \in [m]$ .  
 433 First, let us consider that each node samples a mini-batch  $I^i$  of size  $B$  with replacement and sends it  
 434 to the server. Then the server calculates the mean of the mini-batches from nodes. One can easily  
 435 show that the variance of the estimator is

$$\begin{aligned} & \mathbb{E} \left[ \left\| \frac{1}{nB} \sum_{i=1}^n \sum_{j \in I^i} x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2 \right] \\ &= \frac{1}{nB} \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \left\| x_{ij} - \frac{1}{m} \sum_{j=1}^m x_{ij} \right\|^2. \end{aligned} \quad (14)$$

436 Next, we consider the same task in the partial participation setting with  $s$ -nice sampling, i.e., we  
 437 sample a random set  $S \subset [n]$  of  $s \in [n]$  nodes without replacement and receive the mini-batches  
 438 only from the sampled nodes. Such sampling of nodes satisfy Assumption 8 with  $p_a = s/n$  and  
 439  $p_a = s(s-1)/n(n-1)$ . In this case, the variance of the estimator (See Lemma 1 with  $r_i = 0$  and  
 440  $s_i = \sum_{j \in I^i} x_{ij}$ ) is

$$\begin{aligned} & \mathbb{E} \left[ \left\| \frac{1}{sB} \sum_{i \in S} \sum_{j \in I^i} x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2 \right] \\ &= \frac{1}{sB} \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \underbrace{\left\| x_{ij} - \frac{1}{m} \sum_{j=1}^m x_{ij} \right\|^2}_{\mathcal{L}_{\max}^2} \\ &+ \frac{n-s}{s(n-1)} \frac{1}{n} \sum_{i=1}^n \underbrace{\left\| \frac{1}{m} \sum_{j=1}^m x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2}_{\hat{\mathcal{L}}^2}. \end{aligned} \quad (15)$$

441 Let us assume that  $s \leq n/2$ . Note that (14) scales with any  $B \geq 1$ , while (15) only scales when  
 442  $B = \mathcal{O}(\mathcal{L}_{\max}^2/\hat{\mathcal{L}}^2)$ . In other words, for large enough  $B$ , the variance in (15) does not significantly  
 443 improves with the growth of  $B$  due to the term  $\hat{\mathcal{L}}^2$ . In our proof, due to partial participation, the  
 444 variance from (15) naturally appears, and we get the same effect. As was mentioned in Sections 6.2  
 445 and 6.3, it can be seen in our convergence rate bounds.

## 446 D Auxiliary facts

447 We list auxiliary facts that we use in our proofs:

448 1. For all  $x, y \in \mathbb{R}^d$ , we have

$$\|x + y\|^2 \leq 2\|x\|^2 + 2\|y\|^2 \quad (16)$$

449 2. Let us take a *random vector*  $\xi \in \mathbb{R}^d$ , then

$$\mathbb{E} \left[ \|\xi\|^2 \right] = \mathbb{E} \left[ \|\xi - \mathbb{E}[\xi]\|^2 \right] + \|\mathbb{E}[\xi]\|^2. \quad (17)$$

### 450 D.1 Sampling Lemma

451 This section provides a lemma that we regularly use in our proofs, and it is useful for samplings that  
452 satisfy Assumption 8.

453 **Lemma 1.** *Suppose that a set  $S$  is a random subset of a set  $[n]$  such that*

454 1.  $\text{Prob}(i \in S) = p_a, \quad \forall i \in [n],$

455 2.  $\text{Prob}(i \in S, j \in S) = p_{aa}, \quad \forall i \neq j \in [n],$

456 3.  $p_{aa} \leq p_a^2,$

457 where  $p_a \in (0, 1]$  and  $p_{aa} \in [0, 1]$ . Let us take random independent vectors  $s_i \in \mathbb{R}^d$  for all  $i \in [n]$ ,  
458 nonrandom vector  $r_i \in \mathbb{R}^d$  for all  $i \in [n]$ , and random vectors

$$v_i = \begin{cases} r_i + \frac{1}{p_a} s_i, & i \in S, \\ r_i, & i \notin S, \end{cases}$$

459 then

$$\begin{aligned} & \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n v_i - \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n v_i \right] \right\|^2 \right] \\ &= \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E} \left[ \|s_i - \mathbb{E}[s_i]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\mathbb{E}[s_i]\|^2 + \frac{p_{aa} - p_a^2}{p_a^2} \left\| \frac{1}{n} \sum_{i=1}^n \mathbb{E}[s_i] \right\|^2 \\ &\leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E} \left[ \|s_i - \mathbb{E}[s_i]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\mathbb{E}[s_i]\|^2. \end{aligned}$$

460 *Proof.* Let us define additional constants  $p_{an}$  and  $p_{nn}$ , such that

461 1.  $\text{Prob}(i \in S, j \notin S) = p_{an}, \quad \forall i \neq j \in [n],$

462 2.  $\text{Prob}(i \notin S, j \notin S) = p_{nn}, \quad \forall i \neq j \in [n].$

463 Note, that

$$p_{an} = p_{aa} - p_a \quad (18)$$

464 and

$$p_{nn} = 1 - p_{aa} - 2p_{an}. \quad (19)$$

465 Using the law of total expectation and

$$\mathbb{E}[v_i] = p_a \left( r_i + \mathbb{E} \left[ \frac{1}{p_a} s_i \right] \right) + (1 - p_a) r_i = r_i + \mathbb{E}[s_i],$$

466 we have

$$\begin{aligned}
& \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n v_i - \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n v_i \right] \right\|^2 \right] \\
&= \frac{1}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \|v_i - (r_i + \mathbb{E}[s_i])\|^2 \right] \\
&\quad + \frac{1}{n^2} \sum_{i \neq j}^n \mathbb{E} [\langle v_i - (r_i + \mathbb{E}[s_i]), v_j - (r_j + \mathbb{E}[s_j]) \rangle] \\
&= \frac{p_a}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \left\| r_i + \frac{1}{p_a} s_i - (r_i + \mathbb{E}[s_i]) \right\|^2 \right] \\
&\quad + \frac{1-p_a}{n^2} \sum_{i=1}^n \|r_i - (r_i + \mathbb{E}[s_i])\|^2 \\
&\quad + \frac{p_{aa}}{n^2} \sum_{i \neq j}^n \mathbb{E} \left[ \left\langle r_i + \frac{1}{p_a} s_i - (r_i + \mathbb{E}[s_i]), r_j + \frac{1}{p_a} s_j - (r_j + \mathbb{E}[s_j]) \right\rangle \right] \\
&\quad + \frac{2p_{an}}{n^2} \sum_{i \neq j}^n \mathbb{E} \left[ \left\langle r_i + \frac{1}{p_a} s_i - (r_i + \mathbb{E}[s_i]), r_j - (r_j + \mathbb{E}[s_j]) \right\rangle \right] \\
&\quad + \frac{p_{nn}}{n^2} \sum_{i \neq j}^n \langle r_i - (r_i + \mathbb{E}[s_i]), r_j - (r_j + \mathbb{E}[s_j]) \rangle.
\end{aligned}$$

467 From the independence of random vectors  $s_i$ , we obtain

$$\begin{aligned}
& \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n v_i - \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n v_i \right] \right\|^2 \right] \\
&= \frac{p_a}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \left\| \frac{1}{p_a} s_i - \mathbb{E}[s_i] \right\|^2 \right] \\
&\quad + \frac{1-p_a}{n^2} \sum_{i=1}^n \|\mathbb{E}[s_i]\|^2 \\
&\quad + \frac{p_{aa}(1-p_a)^2}{n^2 p_a^2} \sum_{i \neq j}^n \langle \mathbb{E}[s_i], \mathbb{E}[s_j] \rangle \\
&\quad + \frac{2p_{an}(p_a-1)}{n^2 p_a} \sum_{i \neq j}^n \langle \mathbb{E}[s_i], \mathbb{E}[s_j] \rangle \\
&\quad + \frac{p_{nn}}{n^2} \sum_{i \neq j}^n \langle \mathbb{E}[s_i], \mathbb{E}[s_j] \rangle.
\end{aligned}$$

468 Using (18) and (19), we have

$$\begin{aligned}
& \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n v_i - \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n v_i \right] \right\|^2 \right] \\
&= \frac{p_a}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \left\| \frac{1}{p_a} s_i - \mathbb{E}[s_i] \right\|^2 \right] \\
&\quad + \frac{1-p_a}{n^2} \sum_{i=1}^n \|\mathbb{E}[s_i]\|^2
\end{aligned}$$

$$\begin{aligned}
& + \frac{p_{aa} - p_a^2}{n^2 p_a^2} \sum_{i \neq j}^n \langle E[s_i], E[s_j] \rangle \\
& \stackrel{(17)}{=} \frac{1}{n^2 p_a} \sum_{i=1}^n E \left[ \|s_i - E[s_i]\|^2 \right] \\
& + \frac{1 - p_a}{n^2 p_a} \sum_{i=1}^n \|E[s_i]\|^2 \\
& + \frac{p_{aa} - p_a^2}{n^2 p_a^2} \sum_{i \neq j}^n \langle E[s_i], E[s_j] \rangle \\
& = \frac{1}{n^2 p_a} \sum_{i=1}^n E \left[ \|s_i - E[s_i]\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|E[s_i]\|^2 \\
& + \frac{p_{aa} - p_a^2}{p_a^2} \left\| \frac{1}{n} \sum_{i=1}^n E[s_i] \right\|^2.
\end{aligned}$$

469 Finally, using that  $p_{aa} \leq p_a^2$ , we have

$$\begin{aligned}
& E \left[ \left\| \frac{1}{n} \sum_{i=1}^n v_i - E \left[ \frac{1}{n} \sum_{i=1}^n v_i \right] \right\|^2 \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n E \left[ \|s_i - E[s_i]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|E[s_i]\|^2.
\end{aligned}$$

470

□

## 471 D.2 Compressors Facts

472 We define the *RandK* compressor that chooses without replacement  $K$  coordinates, scales them by a  
473 constant factor to preserve unbiasedness and zero-out other coordinates.

**Definition 5.** Let us take a random subset  $S$  from  $[d]$ ,  $|S| = K$ ,  $K \in [d]$ . We say that a stochastic mapping  $\mathcal{C} : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is *RandK* if

$$\mathcal{C}(x) = \frac{d}{K} \sum_{j \in S} x_j e_j,$$

474 where  $\{e_i\}_{i=1}^d$  is the standard unit basis.

475 **Theorem 6.** If  $\mathcal{C}$  is *RandK*, then  $\mathcal{C} \in \mathbb{U} \left( \frac{d}{K} - 1 \right)$ .

476 See the proof in (Beznosikov et al., 2020).

## 477 E Proofs of Theorems

478 There are three different sources of randomness in Algorithm 1: the first one from vectors  $\{k_i^{t+1}\}_{i=1}^n$ ,  
479 the second one from compressors  $\{\mathcal{C}_i\}_{i=1}^n$ , and the third one from availability of nodes. We define  
480  $E_k[\cdot]$ ,  $E_C[\cdot]$  and  $E_{p_a}[\cdot]$  to be conditional expectations w.r.t.  $\{k_i^{t+1}\}_{i=1}^n$ ,  $\{\mathcal{C}_i\}_{i=1}^n$ , and availability,  
481 accordingly, conditioned on all previous randomness. Moreover, we define  $E_{t+1}[\cdot]$  to be a conditional  
482 expectation w.r.t. all randomness in iteration  $t+1$  conditioned on all previous randomness. Note,  
483 that  $E_{t+1}[\cdot] = E_k[E_C[E_{p_a}[\cdot]]]$ .

484 In the case of **DASHA-PP-PAGE**, there are two different sources of randomness from  $\{k_i^{t+1}\}_{i=1}^n$ .  
485 We define  $E_{p_{\text{page}}}[\cdot]$  and  $E_B[\cdot]$  to be conditional expectations w.r.t. the probabilistic switching and  
486 mini-batch indices  $I_i^t$ , accordingly, conditioned on all previous randomness. Note, that  $E_{t+1}[\cdot] =$   
487  $E_B[E_C[E_{p_a}[E_{p_{\text{page}}}[\cdot]]]]$  and  $E_{t+1}[\cdot] = E_B[E_{p_{\text{page}}}[E_C[E_{p_a}[\cdot]]]]$ .



## 488 E.1 Standard Lemmas in the Nonconvex Setting

489 We start the proof of theorems by providing standard lemmas from the nonconvex optimization.

490 **Lemma 2.** Suppose that Assumption 2 holds and let  $x^{t+1} = x^t - \gamma g^t$ . Then for any  $g^t \in \mathbb{R}^d$  and  
 491  $\gamma > 0$ , we have

$$f(x^{t+1}) \leq f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2. \quad (20)$$

492 *Proof.* Using  $L$ -smoothness, we have

$$\begin{aligned} f(x^{t+1}) &\leq f(x^t) + \langle \nabla f(x^t), x^{t+1} - x^t \rangle + \frac{L}{2} \|x^{t+1} - x^t\|^2 \\ &= f(x^t) - \gamma \langle \nabla f(x^t), g^t \rangle + \frac{L}{2} \|x^{t+1} - x^t\|^2. \end{aligned}$$

493 Next, due to  $-\langle x, y \rangle = \frac{1}{2} \|x - y\|^2 - \frac{1}{2} \|x\|^2 - \frac{1}{2} \|y\|^2$ , we obtain

$$f(x^{t+1}) \leq f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2.$$

494

□

495 **Lemma 3.** Suppose that Assumption 1 holds and

$$\mathbb{E} [f(x^{t+1})] + \gamma \Psi^{t+1} \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] + \gamma \Psi^t + \gamma C, \quad (21)$$

496 where  $\Psi^t$  is a sequence of numbers,  $\Psi^t \geq 0$  for all  $t \in [T]$ , constant  $C \geq 0$ , and constant  $\gamma > 0$ .  
 497 Then

$$\mathbb{E} [\|\nabla f(\hat{x}^T)\|^2] \leq \frac{2\Delta_0}{\gamma T} + \frac{2\Psi^0}{T} + 2C, \quad (22)$$

498 where a point  $\hat{x}^T$  is chosen uniformly from a set of points  $\{x^t\}_{t=0}^{T-1}$ .

499 *Proof.* By unrolling (21) for  $t$  from 0 to  $T - 1$ , we obtain

$$\frac{\gamma}{2} \sum_{t=0}^{T-1} \mathbb{E} [\|\nabla f(x^t)\|^2] + \mathbb{E} [f(x^T)] + \gamma \Psi^T \leq f(x^0) + \gamma \Psi^0 + \gamma TC.$$

500 We subtract  $f^*$ , divide inequality by  $\frac{\gamma T}{2}$ , and take into account that  $f(x) \geq f^*$  for all  $x \in \mathbb{R}$ , and  
 501  $\Psi^t \geq 0$  for all  $t \in [T]$ , to get the following inequality:

$$\frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E} [\|\nabla f(x^t)\|^2] \leq \frac{2\Delta_0}{\gamma T} + \frac{2\Psi^0}{T} + 2C.$$

502 It is left to consider the choice of a point  $\hat{x}^T$  to complete the proof of the lemma. □

**Lemma 4.** If  $0 < \gamma \leq (L + \sqrt{A})^{-1}$ ,  $L > 0$ , and  $A \geq 0$ , then

$$\frac{1}{2\gamma} - \frac{L}{2} - \frac{\gamma A}{2} \geq 0.$$

503 The lemma can be easily checked with the direct calculation.

## 504 E.2 Generic Lemmas

505 **Lemma 5.** Suppose that Assumptions 7 and 8 hold and let us consider sequences  $g_i^{t+1}$ ,  $h_i^{t+1}$ , and  
 506  $k_i^{t+1}$  from Algorithm 1, then

$$\begin{aligned} & \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \\ & \leq \frac{2\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \frac{a^2((2\omega + 1)p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2, \end{aligned} \quad (23)$$

507 and

$$\begin{aligned} & \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \\ & \leq \frac{2\omega}{p_a} \|k_i^{t+1}\|^2 + \left( \frac{a^2(2\omega + 1 - p_a)}{p_a} + (1-a)^2 \right) \|g_i^t - h_i^t\|^2 \quad \forall i \in [n]. \end{aligned} \quad (24)$$

508 *Proof.* First, we estimate  $\mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right]$ :

$$\begin{aligned} & \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \\ & = \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1} - \mathbb{E}_C [\mathbb{E}_{p_a} [g^{t+1} - h^{t+1}]]\|^2 \right] \right] + \|\mathbb{E}_C [\mathbb{E}_{p_a} [g^{t+1} - h^{t+1}]]\|^2, \end{aligned}$$

509 where we used (17). Due to Assumption 8, we have

$$\begin{aligned} & \mathbb{E}_C [\mathbb{E}_{p_a} [g_i^{t+1}]] \\ & = p_a \mathbb{E}_C \left[ g_i^t + \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right] + (1-p_a) g_i^t \\ & = g_i^t + p_a \mathbb{E}_C \left[ \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right] \\ & = g_i^t + k_i^{t+1} - a (g_i^t - h_i^t), \end{aligned}$$

510 and

$$\mathbb{E}_C [\mathbb{E}_{p_a} [h_i^{t+1}]] = p_a \mathbb{E}_C \left[ h_i^t + \frac{1}{p_a} k_i^{t+1} \right] + (1-p_a) h_i^t = h_i^t + k_i^{t+1}.$$

511 Thus, we can get

$$\begin{aligned} & \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \\ & = \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1} - \mathbb{E}_C [\mathbb{E}_{p_a} [g^{t+1} - h^{t+1}]]\|^2 \right] \right] + (1-a)^2 \|g^t - h^t\|^2. \end{aligned}$$

512 Due to the independence of compressors, we can use Lemma 1 with  $r_i = g_i^t - h_i^t$  and  $s_i =$

513  $p_a \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_i^{t+1}$ , and obtain

$$\begin{aligned} & \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \\ & \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_C \left[ \left\| p_a \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_i^{t+1} - \mathbb{E}_C \left[ p_a \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_i^{t+1} \right] \right\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \mathbb{E}_C \left[ p_a \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_i^{t+1} \right] \right\|^2 \end{aligned}$$

$$\begin{aligned}
& + (1-a)^2 \|g^t - h^t\|^2 \\
& = \frac{p_a}{n^2} \sum_{i=1}^n \mathbb{E}_{\mathcal{C}} \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& \quad + \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2.
\end{aligned}$$

514 From Assumption 7, we have

$$\begin{aligned}
& \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \\
& \leq \frac{\omega p_a}{n^2} \sum_{i=1}^n \left\| \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right\|^2 + \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \\
& = \frac{\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1} - a (g_i^t - h_i^t)\|^2 + \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \\
& \stackrel{(16)}{\leq} \frac{2\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \frac{a^2 ((2\omega + 1)p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2.
\end{aligned}$$

515 The second inequality can be proved almost in the same way:

$$\begin{aligned}
& \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \\
& = \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g_i^{t+1} - h_i^{t+1} - \mathbb{E}_{\mathcal{C}} [\mathbb{E}_{p_a} [g_i^{t+1} - h_i^{t+1}]]\|^2 \right] \right] + \|\mathbb{E}_{\mathcal{C}} [\mathbb{E}_{p_a} [g_i^{t+1} - h_i^{t+1}]]\|^2 \\
& = \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g_i^{t+1} - h_i^{t+1} - g_i^t + a (g_i^t - h_i^t) + h_i^t\|^2 \right] \right] + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& = p_a \mathbb{E}_{\mathcal{C}} \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \frac{1}{p_a} k_i^{t+1} + a (g_i^t - h_i^t) \right\|^2 \right] \\
& \quad + a^2 (1-p_a) \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(17)}{=} p_a \mathbb{E}_{\mathcal{C}} \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& \quad + a^2 \frac{(1-p_a)^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& \quad + a^2 (1-p_a) \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \leq \frac{\omega}{p_a} \|k_i^{t+1} - a (g_i^t - h_i^t)\|^2 \\
& \quad + \frac{a^2 (1-p_a)}{p_a} \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(16)}{\leq} \frac{2\omega}{p_a} \|k_i^{t+1}\|^2 + \frac{a^2 (2\omega + 1 - p_a)}{p_a} \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2.
\end{aligned}$$

516

□

517 **Lemma 6.** Suppose that Assumptions 2, 7, and 8 hold and let us take  $a = \frac{p_a}{2\omega+1}$ , then

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{n p_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& \quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{n p_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] + \frac{4\gamma\omega(2\omega+1)}{n p_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

518 *Proof.* Due to Lemma 2 and the update step from Line 4 in Algorithm 1, we have

$$\begin{aligned}
& \mathbb{E}_{t+1} [f(x^{t+1})] \\
& \leq \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2 \right] \\
& = \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - h^t + h^t - \nabla f(x^t)\|^2 \right] \\
& \stackrel{(17)}{\leq} \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \left( \|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2 \right) \right].
\end{aligned}$$

519 Let us fix some constants  $\kappa, \eta \in [0, \infty)$  that we will define later. Combining the last inequality,  
520 bounds (23), (24) and using the law of total expectation, we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] \\
& + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& = \mathbb{E} [\mathbb{E}_{t+1} [f(x^{t+1})]] \\
& + \kappa \mathbb{E} [\mathbb{E}_C [\mathbb{E}_{p_a} [\|g^{t+1} - h^{t+1}\|^2]]] + \eta \mathbb{E} \left[ \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \left( \|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2 \right) \right] \\
& + \kappa \mathbb{E} \left[ \frac{2\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \frac{a^2((2\omega+1)p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \right] \\
& + \eta \mathbb{E} \left[ \frac{2\omega}{n p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \left( \frac{a^2(2\omega+1-p_a)}{p_a} + (1-a)^2 \right) \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& = \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( \gamma + \kappa(1-a)^2 \right) \mathbb{E} [\|g^t - h^t\|^2] \\
& + \left( \frac{\kappa a^2((2\omega+1)p_a - p_{aa})}{n p_a^2} + \eta \left( \frac{a^2(2\omega+1-p_a)}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( \frac{2\kappa\omega}{n p_a} + \frac{2\eta\omega}{p_a} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

521 Now, by taking  $\kappa = \frac{\gamma}{a}$ , we can see that  $\gamma + \kappa(1-a)^2 \leq \kappa$ , and thus

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] \\
& + \frac{\gamma}{a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma}{a} \mathbb{E} [\|g^t - h^t\|^2] \\
& + \left( \frac{\gamma a((2\omega+1)p_a - p_{aa})}{n p_a^2} + \eta \left( \frac{a^2(2\omega+1-p_a)}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( \frac{2\gamma\omega}{a n p_a} + \frac{2\eta\omega}{p_a} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

522 Next, by taking  $\eta = \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2}$  and considering the choice of  $a$ , one can show that  
 523  $\left( \frac{\gamma a((2\omega+1)p_a - p_{aa})}{np_a^2} + \eta \left( \frac{a^2(2\omega+1-p_a)}{p_a} + (1-a)^2 \right) \right) \leq \eta$ . Thus

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + \left( \frac{2\gamma(2\omega+1)\omega}{np_a^2} + \frac{2\gamma((2\omega+1)p_a - p_{aa})\omega}{np_a^3} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right]. \end{aligned}$$

524 Considering that  $p_{aa} \geq 0$ , we can simplify the last term and get

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + \frac{4\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right]. \end{aligned}$$

525 □

### 526 E.3 Proof for DASHA-PP

527 **Lemma 7.** Suppose that Assumptions 3 and 8 hold. For  $h_i^{t+1}$  and  $k_i^{t+1}$  from Algorithm 1 (DASHA-PP)  
 528 we have

1.

$$\begin{aligned} & \mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \\ & \leq \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} \|x^{t+1} - x^t\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

2.

$$\begin{aligned} & \mathbb{E}_{p_a} [\|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2] \\ & \leq \frac{2(1-p_a)}{p_a} L_i^2 \|x^{t+1} - x^t\|^2 + \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n]. \end{aligned}$$

3.

$$\|k_i^{t+1}\|^2 \leq 2L_i^2 \|x^{t+1} - x^t\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n].$$

529 *Proof.* First, let us proof the bound for  $\mathbb{E}_k [\mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]]$ :

$$\mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]$$

$$= \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \mathbb{E}_{p_a} [h^{t+1}] \right\|^2 \right] + \left\| \mathbb{E}_{p_a} [h^{t+1}] - \nabla f(x^{t+1}) \right\|^2.$$

530 Using

$$\mathbb{E}_{p_a} [h_i^{t+1}] = h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))$$

531 and (17), we have

$$\begin{aligned} & \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \\ &= \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \mathbb{E}_{p_a} [h^{t+1}] \right\|^2 \right] + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \end{aligned}$$

532 We can use Lemma 1 with  $r_i = h_i^t$  and  $s_i = k_i^{t+1}$  to obtain

$$\begin{aligned} & \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \\ & \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left\| k_i^{t+1} - k_i^{t+1} \right\|^2 + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| k_i^{t+1} \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ &= \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ & \stackrel{(16)}{\leq} \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ & \leq \frac{2(p_a - p_{aa}) \hat{L}^2}{n p_a^2} \left\| x^{t+1} - x^t \right\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \end{aligned}$$

533 In the last in inequality, we used Assumption 3. Now, we prove the second inequality:

$$\begin{aligned} & \mathbb{E}_{p_a} \left[ \left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \\ &= \mathbb{E}_{p_a} \left[ \left\| h_i^{t+1} - \mathbb{E}_{p_a} [h_i^{t+1}] \right\|^2 \right] + \left\| \mathbb{E}_{p_a} [h_i^{t+1}] - \nabla f_i(x^{t+1}) \right\|^2 \\ &= \mathbb{E}_{p_a} \left[ \left\| h_i^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\ &= \frac{(1-p_a)^2}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \quad + (1-p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\ &= \frac{(1-p_a)}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\ & \leq \frac{2(1-p_a)}{p_a} L_i^2 \left\| x^{t+1} - x^t \right\|^2 + \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \left\| h_i^t - \nabla f_i(x^t) \right\|^2. \end{aligned}$$

534 Finally, the third inequality of the theorem follows from (16) and Assumption 3.  $\square$

535 **Theorem 2.** Suppose that Assumptions 1, 2, 3, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_a}{2-p_a}$ ,

$$\gamma \leq \left( L + \left[ \frac{48\omega(2\omega+1)}{n p_a^2} + \frac{16}{n p_a^2} \left( 1 - \frac{p_{aa}}{p_a} \right) \right]^{1/2} \hat{L} \right)^{-1},$$

536 and  $g_i^0 = h_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP), then  $\mathbb{E} \left[ \left\| \nabla f(\hat{x}^T) \right\|^2 \right] \leq \frac{2\Delta_0}{\gamma T}$ .

537 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 6, Lemma 7,  
538 and the law of total expectation, we obtain

$$\mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \left\| g^{t+1} - h^{t+1} \right\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{n p_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \left\| g_i^{t+1} - h_i^{t+1} \right\|^2 \right]$$



$$\begin{aligned}
& + \nu \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& = \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \quad + \nu \mathbb{E} \left[ \mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \right] + \rho \mathbb{E} \left[ \mathbb{E}_{p_a} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& \quad + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& \quad + \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \mathbb{E} \left[ 2\hat{L}^2 \|x^{t+1} - x^t\|^2 + 2b^2 \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& \quad + \nu \mathbb{E} \left[ \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} \|x^{t+1} - x^t\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \right] \\
& \quad + \rho \mathbb{E} \left[ \frac{2(1-p_a)\hat{L}^2}{p_a} \|x^{t+1} - x^t\|^2 + \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

539 After rearranging the terms, we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& \quad + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& \quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma\omega(2\omega + 1)\hat{L}^2}{np_a^2} - \nu \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} - \rho \frac{2(1-p_a)\hat{L}^2}{p_a} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& \quad + (\gamma + \nu(1-b)^2) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& \quad + \left( \frac{8b^2\gamma\omega(2\omega + 1)}{np_a^2} + \nu \frac{2b^2(p_a - p_{aa})}{np_a^2} + \rho \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

540 By taking  $\nu = \frac{\gamma}{b}$ , one can show that  $(\gamma + \nu(1-b)^2) \leq \nu$ , and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \quad + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& \quad + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma\omega(2\omega+1)\widehat{L}^2}{np_a^2} - \frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} - \rho \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b(p_a - p_{aa})}{np_a^2} + \rho \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

541 Note that  $b = \frac{p_a}{2-p_a}$ , thus

$$\begin{aligned}
& \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b(p_a - p_{aa})}{np_a^2} + \rho \left( \frac{2b^2(1-p_a)}{p_a} + (1-b)^2 \right) \right) \\
& \leq \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b(p_a - p_{aa})}{np_a^2} + \rho(1-b) \right).
\end{aligned}$$

542 And if we take  $\rho = \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}$ , then

$$\left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b(p_a - p_{aa})}{np_a^2} + \rho(1-b) \right) \leq \rho,$$

543 and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma\omega(2\omega+1)\widehat{L}^2}{np_a^2} - \frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right. \\
& \quad \left. - \frac{16b\gamma\omega(2\omega+1)(1-p_a)\widehat{L}^2}{np_a^3} - \frac{4\gamma(p_a - p_{aa})(1-p_a)\widehat{L}^2}{np_a^3} \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

544 Let us simplify the last inequality. First, note that

$$\frac{16b\gamma\omega(2\omega+1)(1-p_a)\widehat{L}^2}{np_a^3} \leq \frac{16\gamma\omega(2\omega+1)\widehat{L}^2}{np_a^2},$$

545 due to  $b \leq p_a$ . Second,

$$\frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \leq \frac{4\gamma(p_a - p_{aa})\widehat{L}^2}{np_a^3},$$

546 due to  $b \geq \frac{p_a}{2}$ . All in all, we have

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{24\gamma\omega(2\omega+1)\hat{L}^2}{np_a^2} - \frac{8\gamma(p_a - p_{aa})\hat{L}^2}{np_a^3}\right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
\end{aligned}$$

547 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
&\mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
\end{aligned}$$

548 It is left to apply Lemma 3 with

$$\begin{aligned}
\Psi^t &= \frac{(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad + \frac{1}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\omega(2\omega+1)}{np_a^2} + \frac{2(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right]
\end{aligned}$$

549 to conclude the proof.  $\square$

#### 550 E.4 Proof for DASHA-PP-PAGE

551 Let us denote

$$\begin{aligned}
k_{i,1}^{t+1} &:= \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)), \\
k_{i,2}^{t+1} &:= \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)), \\
h_{i,1}^{t+1} &:= \begin{cases} h_i^t + \frac{1}{p_a} k_{i,1}^{t+1}, & i^{\text{th}} \text{ node is participating,} \\ h_i^t, & \text{otherwise,} \end{cases} \\
h_{i,2}^{t+1} &:= \begin{cases} h_i^t + \frac{1}{p_a} k_{i,2}^{t+1}, & i^{\text{th}} \text{ node is participating,} \\ h_i^t, & \text{otherwise,} \end{cases}
\end{aligned}$$

552  $h_1^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,1}^{t+1}$ , and  $h_2^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,2}^{t+1}$ . Note, that

$$h^{t+1} = \begin{cases} h_1^{t+1}, & \text{with probability } p_{\text{page}}, \\ h_2^{t+1}, & \text{with probability } 1 - p_{\text{page}}. \end{cases}$$

553 **Lemma 8.** Suppose that Assumptions 3, 4, and 8 hold. For  $h_i^{t+1}$  and  $k_i^{t+1}$  from Algorithm 1  
 554 (DASHA-PP-PAGE) we have

1.

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\ & \leq \left( \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} + \frac{(1 - p_{\text{page}})L_{\max}^2}{np_a B} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

2.

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\ & \leq \left( \frac{2(1 - p_a)L_i^2}{p_a} + \frac{(1 - p_{\text{page}})L_{\max}^2}{p_a B} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \left( \frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n]. \end{aligned}$$

3.

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|k_i^{t+1}\|^2 \right] \right] \\ & \leq \left( 2L_i^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n]. \end{aligned}$$

555 *Proof.* First, we prove the first inequality of the theorem:

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\ & = p_{\text{page}} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_1^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_2^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right]. \end{aligned}$$

556 Using

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] = \\ & = p_a h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) + (1 - p_a) h_i^t \\ & = h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)). \end{aligned}$$

557 and

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] = \\ & = p_a h_i^t + \mathbb{E}_B \left[ \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) \right] + (1 - p_a) h_i^t \\ & = h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t), \end{aligned}$$

558 we obtain

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\ & \stackrel{(17)}{=} p_{\text{page}} \mathbb{E}_{p_a} \left[ \|h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}]\|^2 \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}]\|^2 \right] \right] \end{aligned}$$

$$\begin{aligned}
& + p_{\text{page}} \left\| \mathbb{E}_{p_a} [h_1^{t+1}] - \nabla f(x^{t+1}) \right\|^2 + (1 - p_{\text{page}}) \left\| \mathbb{E}_B [\mathbb{E}_{p_a} [h_2^{t+1}]] - \nabla f(x^{t+1}) \right\|^2 \\
& = p_{\text{page}} \mathbb{E}_{p_a} \left[ \left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \left\| h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}] \right\|^2 \right] \\
& + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \left\| h^t - \nabla f(x^t) \right\|^2. \tag{25}
\end{aligned}$$

559 Next, we consider  $\mathbb{E}_{p_a} \left[ \left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right]$ . We can use Lemma 1 with  $r_i = h_i^t$  and  $s_i = k_{i,1}^{t+1}$   
560 to obtain

$$\begin{aligned}
& \mathbb{E}_{p_a} \left[ \left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left\| k_{i,1}^{t+1} - k_{i,1}^{t+1} \right\|^2 + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| k_{i,1}^{t+1} \right\|^2 \\
& = \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& \stackrel{(16)}{\leq} \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2.
\end{aligned}$$

561 From Assumption 3, we have

$$\begin{aligned}
& \mathbb{E}_{p_a} \left[ \left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] \\
& \leq \frac{2(p_a - p_{aa})\hat{L}^2}{n p_a^2} \left\| x^{t+1} - x^t \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2. \tag{26}
\end{aligned}$$

562 Now, we prove the bound for  $\mathbb{E}_B \left[ \left\| h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}] \right\|^2 \right]$ . Considering that mini-  
563 batches in the algorithm are independent, we can use Lemma 1 with  $r_i = h_i^t$  and  $s_i = k_{i,2}^{t+1}$   
564 to obtain

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}] \right\|^2 \right] \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - \mathbb{E}_{p_a} [k_{i,2}^{t+1}] \right\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \mathbb{E}_B [k_{i,2}^{t+1}] \right\|^2 \\
& = \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\
& = \frac{1}{n^2 p_a B^2} \sum_{i=1}^n \mathbb{E}_B \left[ \sum_{j \in I_i^t} \left\| (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\
& = \frac{1}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\
& \leq \frac{1}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) \right\|^2 + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2.
\end{aligned}$$

565 Next, we use Assumptions 3 and 4 to get

$$\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_2^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_2^{t+1} \right] \right] \right\|^2 \right] \right] \leq \left( \frac{L_{\max}^2}{np_a B} + \frac{(p_a - p_{aa}) \hat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2. \quad (27)$$

566 Applying (26) and (27) into (25), we get

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right] \right] \\ & \leq p_{\text{page}} \left( \frac{2(p_a - p_{aa}) \hat{L}^2}{np_a^2} \|x^{t+1} - x^t\|^2 + \frac{2(p_a - p_{aa}) b^2}{n^2 p_a^2 p_{\text{page}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right) + \\ & \quad + (1 - p_{\text{page}}) \left( \frac{L_{\max}^2}{np_a B} + \frac{(p_a - p_{aa}) \hat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h^t - \nabla f(x^t)\|^2 \\ & \leq \left( \frac{2(p_a - p_{aa}) \hat{L}^2}{np_a^2} + \frac{(1 - p_{\text{page}}) L_{\max}^2}{np_a B} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \frac{2(p_a - p_{aa}) b^2}{n^2 p_a^2 p_{\text{page}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

567 The proof of the second inequality almost repeats the previous one:

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \right] \\ & = p_{\text{page}} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,2}^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \\ & \stackrel{(17)}{=} p_{\text{page}} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & \quad + p_{\text{page}} \left\| \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] - \nabla f_i(x^{t+1}) \right\|^2 + (1 - p_{\text{page}}) \left\| \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] - \nabla f_i(x^{t+1}) \right\|^2 \\ & = p_{\text{page}} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & \quad + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2. \quad (28) \end{aligned}$$

568 Let us consider  $\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right]$ :

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & = \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \\ & = p_a \left\| h_i^t + \frac{1}{p_a} k_{i,1}^{t+1} - \left( h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right) \right\|^2 \\ & \quad + (1 - p_a) \left\| h_i^t - \left( h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right) \right\|^2 \\ & = \frac{(1 - p_a)^2}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \quad + (1 - p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & = \frac{1 - p_a}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2. \end{aligned}$$



569 Considering (16) and Assumption 3, we obtain

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & \leq \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{2(1-p_a)b^2}{p_a p_{\text{page}}^2} \|h_i^t - \nabla f_i(x^t)\|^2. \end{aligned} \quad (29)$$

570 Next, we obtain the bound for  $\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right]$ :

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & = p_a \mathbb{E}_B \left[ \left\| h_i^t + \frac{1}{p_a} k_{i,2}^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + (1-p_a) \mathbb{E}_B \left[ \left\| h_i^t - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & = p_a \mathbb{E}_B \left[ \left\| \frac{1}{p_a} k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + (1-p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \stackrel{(17)}{=} \frac{1}{p_a} \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{(1-p_a)^2}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \quad + (1-p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & = \frac{1}{p_a} \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{1-p_a}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \leq \frac{1}{p_a} \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2, \end{aligned} \quad (30)$$

571 where we used Assumption 3. By plugging (29) and (30) into (28), we get

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \right] \\ & \leq p_{\text{page}} \left( \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{2(1-p_a)b^2}{p_a p_{\text{page}}^2} \|h_i^t - \nabla f_i(x^t)\|^2 \right) \\ & \quad + (1-p_{\text{page}}) \left( \frac{1}{p_a} \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 \right) \\ & \quad + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\ & \leq \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{1-p_{\text{page}}}{p_a} \mathbb{E}_B \left[ \left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2. \end{aligned}$$

572 From the independence of elements in the mini-batch, we obtain

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \right] \\ & \leq \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{1-p_{\text{page}}}{p_a} \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \end{aligned}$$

$$\begin{aligned}
&= \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{1-p_{\text{page}}}{p_a B^2} \mathbb{E}_B \left[ \sum_{j \in I_i^t} \|(\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
&\quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
&= \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{1-p_{\text{page}}}{m p_a B} \sum_{j=1}^m \|(\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \\
&\quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{1-p_{\text{page}}}{m p_a B} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 \\
&\quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \left( \frac{2(1-p_a)L_i^2}{p_a} + \frac{(1-p_{\text{page}})L_{\max}^2}{p_a B} \right) \|x^{t+1} - x^t\|^2 \\
&\quad + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2,
\end{aligned}$$

573 where we used Assumption 4. Finally, we prove the last inequality:

$$\begin{aligned}
&\mathbb{E}_B \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|k_i^{t+1}\|^2 \right] \right] \\
&= p_{\text{page}} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) \right\|^2 \right] \\
&\stackrel{(17)}{=} p_{\text{page}} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + (1-p_{\text{page}}) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\stackrel{(16)}{\leq} 2p_{\text{page}} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + (1-p_{\text{page}}) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\leq 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2
\end{aligned}$$

$$+ (1 - p_{\text{page}}) \mathbb{E}_B \left[ \left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right].$$

574 Using the independence of elements in the mini-batch, we have

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|k_i^{t+1}\|^2 \right] \right] \\ & \leq 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2 \\ & \quad + \frac{1 - p_{\text{page}}}{B^2} \mathbb{E}_B \left[ \sum_{j \in I_i^t} \|(\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\ & = 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2 \\ & \quad + \frac{1 - p_{\text{page}}}{Bm} \sum_{j=1}^m \|(\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \\ & \leq 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2 \\ & \quad + \frac{1 - p_{\text{page}}}{Bm} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 \end{aligned}$$

575 It is left to consider Assumptions 3 and 4 to get

$$\begin{aligned} & \mathbb{E}_B \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|k_i^{t+1}\|^2 \right] \right] \\ & \leq \left( 2L_i^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2. \end{aligned}$$

576 □

577 **Theorem 3.** Suppose that Assumptions 1, 2, 3, 4, 7, and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_{\text{page}}p_a}{2-p_a}$ ,  
578 probability  $p_{\text{page}} \in (0, 1]$ ,

$$\gamma \leq \left( L + \left[ \frac{48\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) + \frac{16}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \right]^{1/2} \right)^{-1}$$

579 and  $g_i^0 = h_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-PAGE) then  $\mathbb{E} [\|\nabla f(\hat{x}^T)\|^2] \leq$   
580  $\frac{2\Delta_0}{\gamma T}$ .

581 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 6, Lemma 8,  
582 and the law of total expectation, we obtain

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & \quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \end{aligned}$$

$$\begin{aligned}
& + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& = \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_{\text{page}}} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \right] \right] \\
& + \nu \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \right] \\
& + \rho \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{page}}} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \left( 2\hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{p_{\text{page}}} \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \nu \mathbb{E} \left( \left( \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} + \frac{(1-p_{\text{page}})L_{\max}^2}{np_a B} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \|h^t - \nabla f(x^t)\|^2 \right) \\
& + \rho \mathbb{E} \left( \left( \frac{2(1-p_a)\hat{L}^2}{p_a} + \frac{(1-p_{\text{page}})L_{\max}^2}{p_a B} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right)
\end{aligned}$$

583 After rearranging the terms, we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right)
\end{aligned}$$

$$\begin{aligned}
& -\nu \left( \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} + \frac{(1 - p_{\text{page}})L_{\max}^2}{np_a B} \right) - \rho \left( \frac{2(1 - p_a)\hat{L}^2}{p_a} + \frac{(1 - p_{\text{page}})L_{\max}^2}{p_a B} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( \gamma + \nu \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \right) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b^2\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{2\nu(p_a - p_{aa})b^2}{np_a^2 p_{\text{page}}} \right. \\
& \quad \left. + \rho \left( \frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Due to  $b = \frac{p_{\text{page}} p_a}{2 - p_a} \leq p_{\text{page}}$ , one can show that  $\left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \leq 1 - b$ . Thus, if we take  $\nu = \frac{\gamma}{b}$ , then

$$\left( \gamma + \nu \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \right) \leq \gamma + \nu(1 - b) = \nu,$$

584 therefore

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \right. \\
& \quad \left. - \frac{\gamma}{b} \left( \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} + \frac{(1 - p_{\text{page}})L_{\max}^2}{np_a B} \right) - \rho \left( \frac{2(1 - p_a)\hat{L}^2}{p_a} + \frac{(1 - p_{\text{page}})L_{\max}^2}{p_a B} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b^2\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{page}}} \right. \\
& \quad \left. + \rho \left( \frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Next, with the choice of  $b = \frac{p_{\text{page}} p_a}{2 - p_a}$ , we ensure that

$$\left( \frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \leq 1 - b.$$

If we take  $\rho = \frac{8b^2\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}}$ , then

$$\left( \frac{8b^2\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{page}}} + \rho \left( \frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \right) \leq \rho,$$

585 therefore

$$\mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right]$$

$$\begin{aligned}
& + \frac{\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \right. \\
& \quad \left. - \frac{\gamma}{bnp_a} \left( 2 \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \right. \\
& \quad \left. - \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^3 p_{\text{page}}} + \frac{2\gamma \left( 1 - \frac{p_{\text{aa}}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2(1-p_a) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \right] \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Let us simplify the inequality. First, due to  $b \geq \frac{p_{\text{page}} p_a}{2}$ , we have

$$\frac{\gamma}{bnp_a} \left( 2 \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \leq \frac{4\gamma}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right).$$

586 Second, due to  $b \leq p_a p_{\text{page}}$  and  $p_{\text{aa}} \leq p_a^2$ , we get

$$\begin{aligned}
& \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^3 p_{\text{page}}} + \frac{2\gamma \left( 1 - \frac{p_{\text{aa}}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2(1-p_a) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \left( \frac{8\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma \left( 1 - \frac{p_{\text{aa}}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2 \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \frac{16\gamma\omega(2\omega+1)}{np_a^2} \left( \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \quad + \frac{4\gamma \left( 1 - \frac{p_{\text{aa}}}{p_a} \right)}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \frac{16\gamma\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \quad + \frac{4\gamma}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{\text{aa}}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\text{max}}^2}{B} \right).
\end{aligned}$$

587 Combining all bounds together, we obtain the following simplified inequality:

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{24\gamma\omega(2\omega+1)}{np_a^2} \left( \widehat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right. \\
& \quad \left. - \frac{8\gamma}{np_a^2 p_{\text{page}}} \left( \left(1 - \frac{p_{\text{aa}}}{p_a}\right) \widehat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

588 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

589 It is left to apply Lemma 3 with

$$\begin{aligned}
\Psi^t &= \frac{(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{1}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2(p_a - p_{\text{aa}})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]
\end{aligned}$$

590 to conclude the proof.  $\square$

591 **Corollary 1.** Let the assumptions from Theorem 3 hold and  $p_{\text{page}} = B/(m+B)$ . Then DASHA-PP-PAGE  
592 needs

$$T := \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{p_a \sqrt{n}} \left( \widehat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{1}{p_a} \sqrt{\frac{m}{n}} \left( \frac{1_{p_a} \widehat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right) \quad (9)$$

593 communication rounds to get an  $\varepsilon$ -solution and the expected number of gradient calculations per  
594 node equals  $\mathcal{O}(m + BT)$ .

595 *Proof.* In the view of Theorem 3, it is enough to do

$$T := \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \sqrt{\frac{\omega^2}{np_a^2} \left( \widehat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right)} + \frac{1}{np_a^2 p_{\text{page}}} \left( \left(1 - \frac{p_{\text{aa}}}{p_a}\right) \widehat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right] \right)$$

596 steps to get  $\varepsilon$ -solution. Using the choice of  $p_{\text{mega}}$  and the definition of  $\mathbb{1}_{p_a}$ , we can get (9).

597 Note that the expected number of gradients calculations at each communication round equals  $p_{\text{mega}}m +$   
598  $(1 - p_{\text{mega}})B = \frac{2mB}{m+B} \leq 2B$ .  $\square$

599 **Corollary 2.** Suppose that assumptions of Corollary 1 hold,  $B \leq \min \left\{ \frac{1}{p_a} \sqrt{\frac{m}{n}}, \frac{L_{\max}^2}{1_{p_a}^2 L^2} \right\}^6$ , and we  
600 use the unbiased compressor RandK with  $K = \Theta(B^d/\sqrt{m})$ . Then the communication complexity of

---

<sup>6</sup>If  $\mathbb{1}_{p_a} = 0$ , then  $\frac{L_{\max}^2}{1_{p_a}^2 L^2} = +\infty$

601 Algorithm 1 is

$$\mathcal{O}\left(d + \frac{L_{\max}\Delta_0 d}{p_a \varepsilon \sqrt{n}}\right), \quad (10)$$

602 and the expected number of gradient calculations per node equals

$$\mathcal{O}\left(m + \frac{L_{\max}\Delta_0 \sqrt{m}}{p_a \varepsilon \sqrt{n}}\right). \quad (11)$$

603 *Proof.* The communication complexity equals

$$\mathcal{O}(d + KT) = \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ KL + K \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + K \frac{1}{p_a} \sqrt{\frac{m}{n}} \left( \frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right).$$

604 Since  $B \leq \frac{L_{\max}^2}{\mathbb{1}_{p_a}^2 \hat{L}^2}$ , we have  $\frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \leq \frac{2L_{\max}}{B}$  and

$$\mathcal{O}(d + KT) = \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ KL + K \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + K \frac{1}{p_a} \sqrt{\frac{m}{n}} \frac{L_{\max}}{B} \right] \right).$$

605 Note that  $K = \Theta\left(\frac{Bd}{\sqrt{m}}\right) = \mathcal{O}\left(\frac{d}{p_a \sqrt{n}}\right)$  and  $\omega + 1 = \frac{d}{K}$  due to Theorem 6, thus

$$\begin{aligned} \mathcal{O}(d + KT) &= \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ \frac{d}{p_a \sqrt{n}} L + \frac{d}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{d}{p_a \sqrt{n}} L_{\max} \right] \right) \\ &= \mathcal{O}\left(d + \frac{L_{\max}\Delta_0 d}{p_a \varepsilon \sqrt{n}}\right). \end{aligned}$$

606 Using the same reasoning, the expected number of gradient calculations per node equals

$$\begin{aligned} \mathcal{O}(m + BT) &= \mathcal{O}\left(m + \frac{\Delta_0}{\varepsilon} \left[ BL + B \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + B \frac{1}{p_a} \sqrt{\frac{m}{n}} \left( \frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right) \\ &= \mathcal{O}\left(m + \frac{\Delta_0}{\varepsilon} \left[ BL + B \frac{d}{K p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + B \frac{1}{p_a} \sqrt{\frac{m}{n}} \frac{L_{\max}}{B} \right] \right) \\ &= \mathcal{O}\left(m + \frac{\Delta_0}{\varepsilon} \left[ \frac{1}{p_a} \sqrt{\frac{m}{n}} L + \frac{\sqrt{m}}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{1}{p_a} \sqrt{\frac{m}{n}} L_{\max} \right] \right) \\ &= \mathcal{O}\left(m + \frac{L_{\max}\Delta_0 \sqrt{m}}{p_a \varepsilon \sqrt{n}}\right). \end{aligned}$$

607 □

## 608 E.5 Proof for DASHA-PP-FINITE-MVR

609 **Lemma 9.** Suppose that Assumptions 3, 4, and 8 hold. For  $h_i^{t+1}$ ,  $h_{ij}^{t+1}$  and  $k_i^{t+1}$  from Algorithm 1  
610 (DASHA-PP-FINITE-MVR) we have

1.

$$\begin{aligned} &\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\ &\leq \left( \frac{2L_{\max}^2}{np_a B} + \frac{2(p_a - p_{aa}) \hat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \\ &\quad + \frac{2(p_a - p_{aa}) b^2}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \frac{2b^2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\ &\quad + (1 - b)^2 \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$



2.

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
& \leq \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
& \quad + \frac{2b^2}{p_a B m} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n].
\end{aligned}$$

3.

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] \\
& \leq \frac{2 \left( 1 - \frac{p_a B}{m} \right) L_{\max}^2}{\frac{p_a B}{m}} \|x^{t+1} - x^t\|^2 \\
& \quad + \left( \frac{2 \left( 1 - \frac{p_a B}{m} \right) b^2}{\frac{p_a B}{m}} + (1-b)^2 \right) \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2, \quad \forall i \in [n], \forall j \in [m].
\end{aligned}$$

4.

$$\begin{aligned}
& \mathbb{E}_B \left[ \|k_i^{t+1}\|^2 \right] \\
& \leq \left( \frac{2L_{\max}^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 \\
& \quad + \frac{2b^2}{Bm} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n].
\end{aligned}$$

611 *Proof.* We start by proving the first inequality. Note that

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} [h_i^{t+1}] \right] \\
& = p_a \left( h_i^t + \frac{1}{p_a} \mathbb{E}_B [k_i^{t+1}] \right) + (1-p_a)h_i^t \\
& = h_i^t + \frac{1}{m} \sum_{j=1}^m \frac{B}{m} \cdot \frac{m}{B} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))) + \left( 1 - \frac{B}{m} \right) \cdot 0 \\
& = \nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t)),
\end{aligned}$$

612 thus

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\
& \stackrel{(17)}{=} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \mathbb{E}_B [\mathbb{E}_{p_a} [h^{t+1}]]\|^2 \right] \right] + (1-b)^2 \|h^t - \nabla f(x^t)\|^2.
\end{aligned}$$

613 We can use Lemma 1 with  $r_i = h_i^t$  and  $s_i = k_i^{t+1}$  to obtain

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_B \left[ \|k_i^{t+1} - \mathbb{E}_B [k_i^{t+1}]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\mathbb{E}_B [k_i^{t+1}]\|^2 \\
& \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\
& = \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_B \left[ \left\| \frac{1}{m} \sum_{j=1}^m k_{ij}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))) \right\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2.
\end{aligned}$$

614 Next, we again use Lemma 1 with  $r_i = 0$ ,  $s_i = \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))$ ,

615  $p_a = \frac{B}{m}$ , and  $p_{aa} = \frac{B(B-1)}{m(m-1)}$ :

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left( \frac{m-B}{Bm(m-1)} \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \right) \\
& \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& \quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
& \leq \frac{1}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \\
& \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& \quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
& \stackrel{(16)}{\leq} \frac{2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) \right\|^2 + \frac{2b^2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| h_{ij}^t - \nabla f_{ij}(x^t) \right\|^2 \\
& \quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
& \quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2.
\end{aligned}$$

616 Due to Assumptions 3 and 4, we have

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right] \\
& \leq \left( \frac{2L_{\max}^2}{n p_a B} + \frac{2(p_a - p_{aa})\hat{L}^2}{n p_a^2} \right) \left\| x^{t+1} - x^t \right\|^2 \\
& \quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 + \frac{2b^2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \left\| h_{ij}^t - \nabla f_{ij}(x^t) \right\|^2 \\
& \quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2.
\end{aligned}$$

617 Let us get the bound for the second inequality:

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \\
& \stackrel{(17)}{=} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \left\| h_i^{t+1} - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \right] \\
& \quad + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
& = p_a \mathbb{E}_B \left[ \left\| h_i^t + \frac{1}{p_a} h_i^{t+1} - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \\
& \quad + (1-p_a) \left\| h_i^t - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \\
& \quad + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2
\end{aligned}$$

$$\begin{aligned}
&\stackrel{(17)}{=} \frac{1}{p_a} \mathbb{E}_B \left[ \|k_i^{t+1} - \mathbb{E}_B[k_i^{t+1}]\|^2 \right] \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

618 Let us use Lemma 1 with  $r_i = 0$ ,  $s_i = \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))$ ,  $p_a = \frac{B}{m}$ , and  
619  $p_{aa} = \frac{B(B-1)}{m(m-1)}$ :

$$\begin{aligned}
&\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
&\leq \frac{1}{p_a} \left( \frac{m-B}{Bm(m-1)} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \right) \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \frac{1}{p_a B m} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\stackrel{(16)}{\leq} \frac{2}{p_a B m} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 + \frac{2(1-p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\quad + \frac{2b^2}{p_a B m} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
&\quad + \frac{2b^2}{p_a B m} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2,
\end{aligned}$$

620 where we used Assumptions 3 and 4. We continue the proof by considering

$$\begin{aligned}
621 &\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right]: \\
&\mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] \\
&\stackrel{(17)}{=} \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_{ij}^{t+1} - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t)))\|^2 \right] \right] \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&= \frac{p_a B}{m} \mathbb{E}_B \left[ \left\| h_{ij}^t + \frac{m}{B p_a} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))) - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t))) \right\|^2 \right] \\
&\quad + \left( 1 - \frac{p_a B}{m} \right) \|h_{ij}^t - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t)))\|^2 \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&= \frac{\left( 1 - \frac{p_a B}{m} \right)^2}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + \left( 1 - \frac{p_a B}{m} \right) \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2
\end{aligned}$$

$$\begin{aligned}
& + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
& = \frac{\left(1 - \frac{p_a B}{m}\right)}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
& + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
& \stackrel{(16)}{\leq} \frac{2\left(1 - \frac{p_a B}{m}\right)}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 + \left(\frac{2\left(1 - \frac{p_a B}{m}\right)b^2}{\frac{p_a B}{m}} + (1-b)^2\right) \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2.
\end{aligned}$$

622 It is left to consider Assumption 4:

$$\begin{aligned}
& \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] \\
& \leq \frac{2\left(1 - \frac{p_a B}{m}\right) L_{\max}^2}{\frac{p_a B}{m}} \|x^{t+1} - x^t\|^2 + \left(\frac{2\left(1 - \frac{p_a B}{m}\right)b^2}{\frac{p_a B}{m}} + (1-b)^2\right) \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2.
\end{aligned}$$

623 Finally, we obtain the bound for the last inequality of the lemma:

$$\begin{aligned}
& \mathbb{E}_B \left[ \|k_i^{t+1}\|^2 \right] \\
& \stackrel{(17)}{=} \mathbb{E}_B \left[ \|k_i^{t+1} - \mathbb{E}_B[k_i^{t+1}]\|^2 \right] \\
& + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2.
\end{aligned}$$

624 Using Lemma 1, we get

$$\begin{aligned}
& \mathbb{E}_B \left[ \|k_i^{t+1}\|^2 \right] \\
& \leq \frac{m-B}{Bm(m-1)} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
& + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& \leq \frac{1}{Bm} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
& + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& \stackrel{(16)}{\leq} \frac{2}{Bm} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 + 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& + \frac{2b^2}{Bm} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
& \leq \left( \frac{2L_{\max}^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 \\
& + \frac{2b^2}{Bm} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2,
\end{aligned}$$

625 where we used Assumptions 3 and 4. □

**Theorem 7.** Suppose that Assumptions 1, 2, 3, 4, 7, and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{\frac{p_a B}{m}}{2 - \frac{p_a B}{m}}$ ,

$$\gamma \leq \left( L + \sqrt{\frac{148\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{L_{\max}^2}{B} \right) + \frac{72m}{np_a^2 B} \left( \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 + \frac{L_{\max}^2}{B} \right)} \right)^{-1},$$

626  $g_i^0 = h_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  and  $h_{ij}^0 = \nabla f_{ij}(x^0)$  for all  $i \in [n], j \in [m]$  in Algorithm 1  
 627 (DASHA-PP-FINITE-MVR) then  $\mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \leq \frac{2\Delta_0}{\gamma T}$ .

628 *Proof.* Let us fix constants  $\nu, \rho, \delta \in [0, \infty)$  that we will define later. Considering Lemma 6, Lemma 9,  
 629 and the law of total expectation, we obtain

$$\begin{aligned}
 & \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
 & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
 & + \delta \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \\
 & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
 & + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
 & + \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \\
 & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
 & + \delta \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \\
 & = \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
 & + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
 & + \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \mathbb{E} \left[ \mathbb{E}_B \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \right] \\
 & + \nu \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \right] \right] \\
 & + \rho \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\
 & + \delta \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] \right] \\
 & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
 & + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
 & + \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \mathbb{E} \left[ \left( \frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{Bmn} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \frac{2b^2}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]
 \end{aligned}$$

$$\begin{aligned}
& + \nu \mathbb{E} \left( \left( \frac{2L_{\max}^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \frac{2b^2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
& \quad \left. + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \right) \\
& + \rho \mathbb{E} \left( \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad + \frac{2b^2}{p_a B n m} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \Bigg) \\
& + \delta \mathbb{E} \left( \frac{2 \left(1 - \frac{p_a B}{m}\right) L_{\max}^2}{\frac{p_a B}{m}} \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \left( \frac{2 \left(1 - \frac{p_a B}{m}\right) b^2}{\frac{p_a B}{m}} + (1-b)^2 \right) \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right).
\end{aligned}$$

Due to  $b = \frac{\frac{p_a B}{m}}{2 - \frac{p_a B}{m}} \leq \frac{p_a}{2 - p_a}$ , we have

$$\left( \frac{2 \left(1 - \frac{p_a B}{m}\right) b^2}{\frac{p_a B}{m}} + (1-b)^2 \right) \leq 1 - b$$

and

$$\left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \leq 1 - b.$$

630 Moreover, we consider that  $1 - \frac{p_a B}{m} \leq 1$ , therefore

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& + \delta \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{4\gamma\omega(2\omega + 1)}{np_a^2} \mathbb{E} \left[ \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{Bmn} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \frac{2b^2}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \nu \mathbb{E} \left( \left( \frac{2L_{\max}^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right.
\end{aligned}$$

$$\begin{aligned}
& + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \frac{2b^2}{n^2 p_a B m} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
& + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \Big) \\
& + \rho \mathbb{E} \left( \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \left. + \frac{2b^2}{p_a B n m} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + (1-b) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right) \\
& + \delta \mathbb{E} \left( \frac{2mL_{\max}^2}{p_a B} \|x^{t+1} - x^t\|^2 + (1-b) \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right).
\end{aligned}$$

631 After rearranging the terms, we get

$$\begin{aligned}
& \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
& + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
& + \delta \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2\right] \\
& \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) \right. \\
& \quad \left. - \nu \left( \frac{2L_{\max}^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) - \delta \frac{2mL_{\max}^2}{p_a B} \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
& + (\gamma + \nu(1-b)^2) \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\nu(p_a - p_{aa})b^2}{np_a^2} + \rho(1-b) \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
& + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{2\nu b^2}{np_a B} + \frac{2\rho b^2}{p_a B} + \delta(1-b) \right) \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2\right].
\end{aligned}$$

632 Thus, if we take  $\nu = \frac{\gamma}{b}$ , then  $\gamma + \nu(1-b)^2 \leq \nu$  and

$$\begin{aligned}
& \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
& + \delta \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2\right]
\end{aligned}$$

$$\begin{aligned}
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \right. \\
&\quad \quad \left. - \left( \frac{2\gamma L_{\max}^2}{bn p_a B} + \frac{2\gamma(p_a - p_{aa})\hat{L}^2}{bn p_a^2} \right) - \rho \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\hat{L}^2}{p_a} \right) - \delta \frac{2mL_{\max}^2}{p_a B} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})b}{np_a^2} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{2\gamma b}{np_a B} + \frac{2\rho b^2}{p_a B} + \delta(1-b) \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right].
\end{aligned}$$

Next, if we take  $\rho = \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}$ , then

$$\left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})b}{np_a^2} + \rho(1-b) \right) = \rho,$$

633 therefore

$$\begin{aligned}
&\mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
&\quad + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
&\quad + \delta \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \\
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \right. \\
&\quad \quad - \left( \frac{2\gamma L_{\max}^2}{bn p_a B} + \frac{2\gamma(p_a - p_{aa})\hat{L}^2}{bn p_a^2} \right) - \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\hat{L}^2}{p_a} \right) \\
&\quad \quad \left. - \delta \frac{2mL_{\max}^2}{p_a B} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{2\gamma b}{np_a B} + \frac{16b^3\gamma\omega(2\omega+1)}{np_a^3 B} + \frac{4b^2\gamma(p_a - p_{aa})}{nBp_a^3} + \delta(1-b) \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right].
\end{aligned}$$



634 Due to  $b \leq p_a$  and  $\frac{p_a - p_{aa}}{p_a} \leq 1$ , we have

$$\begin{aligned} & \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{2\gamma b}{np_aB} + \frac{16b^3\gamma\omega(2\omega+1)}{np_a^3B} + \frac{4b^2\gamma(p_a - p_{aa})}{nBp_a^3} \\ & \leq \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{2\gamma b}{np_aB} + \frac{16b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{4\gamma b}{np_aB} \\ & = \frac{24b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma b}{np_aB}. \end{aligned}$$

635 Let us take  $\delta = \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB}$ . Thus

$$\left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{2\gamma b}{np_aB} + \frac{16b^3\gamma\omega(2\omega+1)}{np_a^3B} + \frac{4b^2\gamma(p_a - p_{aa})}{nBp_a^3} + \delta(1-b) \right) \leq \delta$$

636 and

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \right. \\ & \quad \left. - \left( \frac{2\gamma L_{\max}^2}{bnp_aB} + \frac{2\gamma(p_a - p_{aa})\hat{L}^2}{bnp_a^2} \right) - \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \left( \frac{2L_{\max}^2}{p_aB} + \frac{2(1-p_a)\hat{L}^2}{p_a} \right) \right. \\ & \quad \left. - \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \frac{2mL_{\max}^2}{p_aB} \right] \mathbb{E}[\|x^{t+1} - x^t\|^2] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] \\ & + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2\right]. \end{aligned}$$

637 Let us simplify the term near  $\mathbb{E}[\|x^{t+1} - x^t\|^2]$ . Due to  $b \leq p_a$ ,  $\frac{p_a - p_{aa}}{p_a} \leq 1$ , and  $1 - p_a \leq 1$ , we

638 have

$$\begin{aligned} & \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \\ & + \left( \frac{2\gamma L_{\max}^2}{bnp_aB} + \frac{2\gamma(p_a - p_{aa})\hat{L}^2}{bnp_a^2} \right) \end{aligned}$$

$$\begin{aligned}
& + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \\
& + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{6\gamma}{np_a B} \right) \frac{2mL_{\max}^2}{p_a B} \\
& \leq \frac{12\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) \\
& + \left( \frac{6\gamma L_{\max}^2}{bnp_a B} + \frac{6\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right) \\
& + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{6\gamma}{np_a B} \right) \frac{2mL_{\max}^2}{p_a B}
\end{aligned}$$

639 Considering that  $b \leq \frac{p_a B}{m}$  and  $b \geq \frac{p_a B}{2m}$ , we obtain

$$\begin{aligned}
& \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) \\
& + \left( \frac{2\gamma L_{\max}^2}{bnp_a B} + \frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right) \\
& + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \left( \frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \\
& + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{6\gamma}{np_a B} \right) \frac{2mL_{\max}^2}{p_a B} \\
& \leq \frac{36\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) + \left( \frac{18\gamma L_{\max}^2}{bnp_a B} + \frac{6\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right) \\
& \leq \frac{36\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) + \left( \frac{36m\gamma L_{\max}^2}{np_a^2 B^2} + \frac{12m\gamma(p_a - p_{aa})\widehat{L}^2}{Bnp_a^3} \right).
\end{aligned}$$

640 All in all, we have

$$\begin{aligned}
& \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
& + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2 B} + \frac{6\gamma}{np_a B} \right) \mathbb{E}\left[\frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2\right] \\
& \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{36\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2L_{\max}^2}{B} + 2\widehat{L}^2 \right) - \left( \frac{36m\gamma L_{\max}^2}{np_a^2 B^2} + \frac{12m\gamma(p_a - p_{aa})\widehat{L}^2}{Bnp_a^3} \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right]
\end{aligned}$$

$$+ \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right].$$

641 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\ & + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\ & + \left( \frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right]. \end{aligned}$$

642 It is left to apply Lemma 3 with

$$\begin{aligned} \Psi^t &= \frac{(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ &+ \frac{1}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\ &+ \left( \frac{8b\omega(2\omega+1)}{np_a^2} + \frac{2(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\ &+ \left( \frac{24b\omega(2\omega+1)}{np_a^2B} + \frac{6}{np_aB} \right) \mathbb{E} \left[ \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \right] \end{aligned}$$

643 to conclude the proof.  $\square$

## 644 E.6 Proof for DASHA-PP-MVR

645 Let us denote  $\nabla f_i(x^{t+1}; \xi_i^{t+1}) := \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1})$ .

646 **Lemma 10.** Suppose that Assumptions 3, 5, 6 and 8 hold. For  $h_i^{t+1}$  and  $k_i^{t+1}$  from Algorithm 1  
647 (DASHA-PP-MVR) we have

1.

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \right] \\ & \leq \frac{2b^2\sigma^2}{np_aB} + \left( \frac{2(1-b)^2L_\sigma^2}{np_aB} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \frac{2(p_a - p_{aa})b^2}{n^2p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

2.

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\ & \leq \frac{2b^2\sigma^2}{p_a B} + \left( \frac{2(1-b)^2 L_\sigma^2}{p_a B} + \frac{2(1-p_a) L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n]. \end{aligned}$$

3.

$$\mathbb{E}_k \left[ \|k_i^{t+1}\|^2 \right] \leq \frac{2b^2\sigma^2}{B} + \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n].$$

648 *Proof.* First, let us proof the bound for  $\mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right]$ :

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\ & = \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \mathbb{E}_k [\mathbb{E}_{p_a} [h^{t+1}]]\|^2 \right] \right] + \|\mathbb{E}_k [\mathbb{E}_{p_a} [h^{t+1}]] - \nabla f(x^{t+1})\|^2. \end{aligned}$$

649 Using

$$\mathbb{E}_k [\mathbb{E}_{p_a} [h_i^{t+1}]] = h_i^t + \mathbb{E}_k [k_i^{t+1}] = h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))$$

650 and (17), we have

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\ & = \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \mathbb{E}_k [\mathbb{E}_{p_a} [h^{t+1}]]\|^2 \right] \right] + (1-b)^2 \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

651 We can use Lemma 1 with  $r_i = h_i^t$  and  $s_i = k_i^{t+1}$  to obtain

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\ & \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \|k_i^{t+1} - \mathbb{E}_k [k_i^{t+1}]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\mathbb{E}_k [k_i^{t+1}]\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ & = \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})) \right. \\ & \quad \left. - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\ & \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ & \stackrel{(16)}{\leq} \frac{2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \|b(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1}))\|^2 \right] \\ & \quad + \frac{2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \|(1-b)(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)))\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\ & \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ & = \frac{2b^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \end{aligned}$$

$$\begin{aligned}
& + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \\
& = \frac{2b^2}{n^2 p_a B^2} \sum_{i=1}^n \sum_{j=1}^B \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^{t+1}) \right\|^2 \right] \\
& + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2.
\end{aligned}$$

652 In the last equality, we use the independence of elements in the mini-batches. Due to Assumption 5,  
653 we get

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right] \\
& \leq \frac{2b^2 \sigma^2}{np_a B} \\
& + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
& \stackrel{(16)}{\leq} \frac{2b^2 \sigma^2}{np_a B} \\
& + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \\
& = \frac{2b^2 \sigma^2}{np_a B} \\
& + \frac{2(1-b)^2}{n^2 p_a B^2} \sum_{i=1}^n \sum_{j=1}^B \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
& + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2,
\end{aligned}$$

654 where we use the independence of elements in the mini-batches. Using Assumptions 3 and 6, we  
655 obtain

$$\mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right]$$

$$\begin{aligned}
&\leq \frac{2b^2\sigma^2}{np_aB} + \left( \frac{2(1-b)^2L_\sigma^2}{np_aB} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \\
&\quad + \frac{2(p_a - p_{aa})b^2}{n^2p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2.
\end{aligned}$$

656 Now, we prove the second inequality:

$$\begin{aligned}
&\mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
&= \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \mathbb{E}_k [\mathbb{E}_{p_a} [h_i^{t+1}]]\|^2 \right] \right] \\
&\quad + \|\mathbb{E}_k [\mathbb{E}_{p_a} [h_i^{t+1}]] - \nabla f_i(x^{t+1})\|^2 \\
&= \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \right] \\
&\quad + \|h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) - \nabla f_i(x^{t+1})\|^2 \\
&= \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \right] \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&= p_a \mathbb{E}_k \left[ \left\| h_i^t + \frac{1}{p_a} k_i^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \\
&\quad + (1-p_a) \|h_i^t - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&= p_a \mathbb{E}_k \left[ \left\| \frac{1}{p_a} k_i^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \\
&\quad + (1-p_a) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\stackrel{(17)}{=} \frac{1}{p_a} \mathbb{E}_k \left[ \|k_i^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\
&\quad + \frac{(1-p_a)^2}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-p_a) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&= \frac{1}{p_a} \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&= \frac{1}{p_a} \mathbb{E}_k \left[ \|b(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})) + (1-b)(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)))\|^2 \right] \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\stackrel{(16)}{\leq} \frac{2b^2}{p_a} \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
&\quad + \frac{2(1-b)^2}{p_a} \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

657 Considering the independence of elements in the mini-batch, we obtain

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
& = \frac{2b^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
& \quad + \frac{2(1-b)^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
& \quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& \quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2. \\
& \stackrel{(16)}{\leq} \frac{2b^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
& \quad + \frac{2(1-b)^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
& \quad + \frac{2(1-p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2
\end{aligned}$$

658 Next, we use Assumptions 3, 6, 5, to get

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
& \leq \frac{2b^2\sigma^2}{p_a B} + \left( \frac{2(1-b)^2 L_\sigma^2}{p_a B} + \frac{2(1-p_a)L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
& \quad + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

659 It is left to prove the bound for  $\mathbb{E}_k \left[ \|k_i^{t+1}\|^2 \right]$ :

$$\begin{aligned}
& \mathbb{E}_k \left[ \|k_i^{t+1}\|^2 \right] \\
& = \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1}))\|^2 \right] \\
& \stackrel{(17)}{=} \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\
& \quad + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& = \mathbb{E}_k \left[ \|b(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})) + (1-b)(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)))\|^2 \right] \\
& \quad + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
& \stackrel{(16)}{\leq} 2b^2 \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
& \quad + 2(1-b)^2 \mathbb{E}_k \left[ \|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
& \quad + 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

660 Using Assumptions 3, 6, 5 and the independence of elements in the mini-batch, we get

$$\begin{aligned} & \mathbb{E}_k \left[ \|k_i^{t+1}\|^2 \right] \\ & \leq \frac{2b^2\sigma^2}{B} + \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2. \end{aligned}$$

661

□

662 **Theorem 4.** Suppose that Assumptions 1, 2, 3, 5, 6, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
 663  $b \in \left(0, \frac{p_a}{2-p_a}\right]$ ,  $\gamma \leq \left(L + \left[\frac{48\omega(2\omega+1)}{np_a^2} \left(\widehat{L}^2 + \frac{(1-b)^2L_\sigma^2}{B}\right) + \frac{12}{np_ab} \left(\left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 + \frac{(1-b)^2L_\sigma^2}{B}\right)\right]^{1/2}\right)^{-1}$ , and  
 664  $g_i^0 = h_i^0$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-MVR). Then

$$\begin{aligned} \mathbb{E} \left[ \|\nabla f(\widehat{x}^T)\|^2 \right] & \leq \frac{1}{T} \left[ \frac{2\Delta_0}{\gamma} + \frac{2}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{32b\omega(2\omega+1)}{np_a^2} + \frac{4\left(1 - \frac{p_{aa}}{p_a}\right)}{np_a} \right) \left( \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \right] \\ & + \left( \frac{48b^2\omega(2\omega+1)}{p_a^2} + \frac{12b}{p_a} \right) \frac{\sigma^2}{nB}. \end{aligned}$$

665 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 6, Lemma 10,  
 666 and the law of total expectation, we obtain

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & = \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \mathbb{E}_k \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \right] \\ & + \nu \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \right] \right] \\ & + \rho \mathbb{E} \left[ \mathbb{E}_B \left[ \mathbb{E}_{p_a} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \end{aligned}$$



$$\begin{aligned}
& + \frac{4\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \frac{2b^2\sigma^2}{B} + \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 + 2b^2 \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \nu \mathbb{E} \left( \frac{2b^2\sigma^2}{np_a B} + \left( \frac{2(1-b)^2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \right) \\
& + \rho \mathbb{E} \left( \frac{2b^2\sigma^2}{p_a B} + \left( \frac{2(1-b)^2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right).
\end{aligned}$$

667 After rearranging the terms, we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right. \\
& \quad \left. - \nu \left( \frac{2(1-b)^2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2(1-b)^2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + (\gamma + \nu(1-b)^2) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\nu(p_a - p_{aa})b^2}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \nu \frac{2b^2}{np_a} + \rho \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

668 By taking  $\nu = \frac{\gamma}{b}$ , one can show that  $(\gamma + \nu(1-b)^2) \leq \nu$ , and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right.
\end{aligned}$$

$$\begin{aligned}
& -\frac{\gamma}{b} \left( \frac{2(1-b)^2 L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa}) \widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2(1-b)^2 L_\sigma^2}{p_a B} + \frac{2(1-p_a) \widehat{L}^2}{p_a} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma (p_a - p_{aa}) b}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma b}{np_a} + \rho \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

669 Note that  $b \leq \frac{p_a}{2-p_a}$ , thus

$$\begin{aligned}
& \left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma (p_a - p_{aa}) b}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \\
& \leq \left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma (p_a - p_{aa}) b}{np_a^2} + \rho (1-b) \right).
\end{aligned}$$

670 And if we take  $\rho = \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a-p_{aa})}{np_a^2}$ , then

$$\left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma (p_a - p_{aa}) b}{np_a^2} + \rho (1-b) \right) \leq \rho,$$

671 and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a-p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right. \\
& \quad \left. - \frac{\gamma}{np_a b} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2 \left( 1 - \frac{p_{aa}}{p_a} \right) \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^3} + \frac{2\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2} \right) \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2(1-p_a) \widehat{L}^2 \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a-p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{8b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{2\gamma b}{np_a} + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a-p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

672 Let us simplify the inequality. First, due to  $b \leq p_a$  and  $(1-p_a) \leq \left( 1 - \frac{p_{aa}}{p_a} \right)$ , we have

$$\left( \frac{8b\gamma\omega(2\omega+1)}{np_a^3} + \frac{2\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2} \right) \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2(1-p_a) \widehat{L}^2 \right)$$

$$\begin{aligned}
&= \frac{8b\gamma\omega(2\omega+1)}{np_a^3} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2(1-p_a)\widehat{L}^2 \right) \\
&\quad + \frac{2\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2(1-p_a)\widehat{L}^2 \right) \\
&\leq \frac{8\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2\widehat{L}^2 \right) \\
&\quad + \frac{2\gamma}{np_a b} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2 \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right),
\end{aligned}$$

673 therefore

$$\begin{aligned}
&\mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{12\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right. \\
&\quad \quad \left. - \frac{3\gamma}{np_a b} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2 \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
&\quad + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b}{np_a} + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B} \\
&= \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{24\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \quad \left. - \frac{6\gamma}{np_a b} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
&\quad + \left( \frac{8b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma b}{np_a} + \left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

674 Also, we can simplify the last term:

$$\begin{aligned}
&\left( \frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \\
&= \frac{16b^3\gamma\omega(2\omega+1)}{np_a^3} + \frac{4b^2\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2}
\end{aligned}$$

$$\leq \frac{16b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{4b\gamma}{np_a},$$

675 thus

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{24\gamma\omega(2\omega+1)}{np_a^2} \left(\frac{(1-b)^2 L_\sigma^2}{B} + \hat{L}^2\right)\right. \\ & \quad \left. - \frac{6\gamma}{np_a b} \left(\frac{(1-b)^2 L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2\right)\right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \left(\frac{24b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{6\gamma b}{np_a}\right) \frac{\sigma^2}{B}. \end{aligned}$$

676 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \left(\frac{24b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{6\gamma b}{np_a}\right) \frac{\sigma^2}{B}. \end{aligned}$$

677 It is left to apply Lemma 3 with

$$\begin{aligned} \Psi^t &= \frac{(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ &+ \frac{1}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{8b\omega(2\omega+1)}{np_a^2} + \frac{2(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \end{aligned}$$

678 and  $C = \left(\frac{24b^2\omega(2\omega+1)}{p_a^2} + \frac{6b}{p_a}\right) \frac{\sigma^2}{nB}$  to conclude the proof.  $\square$

679 **Corollary 3.** Suppose that assumptions from Theorem 4 hold, momentum  $b =$   
 680  $\Theta\left(\min\left\{\frac{p_a}{\omega} \sqrt{\frac{n\varepsilon B}{\sigma^2}}, \frac{p_a n \varepsilon B}{\sigma^2}\right\}\right)$ ,  $\frac{\sigma^2}{n\varepsilon B} \geq 1$ , and  $h_i^0 = \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0)$  for all  $i \in [n]$ ,

681 and batch size  $B_{\text{init}} = \Theta\left(\frac{\sqrt{p_a B}}{b}\right)$ , then Algorithm 1 (DASHA-PP-MVR) needs

$$T := \mathcal{O}\left(\frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left( \frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right] + \frac{\sigma^2}{\sqrt{p_a n \varepsilon B}} \right)$$

682 communication rounds to get an  $\varepsilon$ -solution and the number of stochastic gradient calculations per  
683 node equals  $\mathcal{O}(B_{\text{init}} + BT)$ .

684 *Proof.* Using the result from Theorem 4, we have

$$\begin{aligned} & \mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \\ & \leq \frac{1}{T} \left[ 2\Delta_0 \left( L + \sqrt{\frac{48\omega(2\omega+1)}{np_a^2}} \left( \hat{L}^2 + \frac{(1-b)^2 L_\sigma^2}{B} \right) + \frac{12}{np_a b} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1-b)^2 L_\sigma^2}{B} \right) \right) \right. \\ & \quad \left. + \frac{2}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{32b\omega(2\omega+1)}{np_a^2} + \frac{4\left(1 - \frac{p_{aa}}{p_a}\right)}{np_a} \right) \left( \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \right] \\ & \quad + \left( \frac{48b^2\omega(2\omega+1)}{p_a^2} + \frac{12b}{p_a} \right) \frac{\sigma^2}{nB} \end{aligned}$$

685 We choose  $b$  to ensure  $\left( \frac{48b^2\omega(2\omega+1)}{p_a^2} + \frac{12b}{p_a} \right) \frac{\sigma^2}{nB} = \Theta(\varepsilon)$ . Note that  $\frac{1}{b} =$

686  $\Theta\left(\max\left\{\frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{n\varepsilon B}}, \frac{\sigma^2}{p_a n \varepsilon B}\right\}\right) \leq \Theta\left(\max\left\{\frac{\omega^2}{p_a}, \frac{\sigma^2}{p_a n \varepsilon B}\right\}\right)$ , thus

$$\begin{aligned} & \mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \\ & = \mathcal{O} \left( \frac{1}{T} \left[ \Delta_0 \left( L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \right. \right. \\ & \quad \left. \left. + \frac{1}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{b\omega^2}{np_a^2} + \frac{1}{np_a} \right) \left( \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \right] + \varepsilon \right), \end{aligned}$$

687 where  $\mathbb{1}_{p_a} = \sqrt{1 - \frac{p_{aa}}{p_a}}$ . It enough to take the following  $T$  to get  $\varepsilon$ -solution.

$$\begin{aligned} T = \mathcal{O} & \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \right. \right. \\ & \quad \left. \left. + \frac{1}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{b\omega^2}{np_a^2} + \frac{1}{np_a} \right) \left( \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \right] \right). \end{aligned}$$

688 Let us bound the norms:

$$\begin{aligned} \mathbb{E} \left[ \|h^0 - \nabla f(x^0)\|^2 \right] & = \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0) - \nabla f(x^0) \right\|^2 \right] \\ & = \frac{1}{n^2 B_{\text{init}}^2} \sum_{i=1}^n \sum_{k=1}^{B_{\text{init}}} \mathbb{E} \left[ \|\nabla f_i(x^0; \xi_{ik}^0) - \nabla f_i(x^0)\|^2 \right] \end{aligned}$$

$$\leq \frac{\sigma^2}{nB_{\text{init}}}.$$

689 Using the same reasoning, one can get  $\frac{1}{n} \sum_{i=1}^n \mathbb{E} \left[ \|h_i^0 - \nabla f_i(x^0)\|^2 \right] \leq \frac{\sigma^2}{B_{\text{init}}}$ . Combining all inequalities, we have

$$T = \mathcal{O} \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \right. \right. \\ \left. \left. + \frac{\sigma^2}{bnB_{\text{init}}} + \frac{b\omega^2\sigma^2}{np_a^2 B_{\text{init}}} + \frac{\sigma^2}{np_a B_{\text{init}}} \right] \right).$$

691 Using the choice of  $B_{\text{init}}$  and  $b$ , we obtain

$$T = \mathcal{O} \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \right. \right. \\ \left. \left. + \frac{\sigma^2}{\sqrt{p_a} n B} + \frac{b^2 \omega^2 \sigma^2}{np_a^{5/2} B} + \frac{b\sigma^2}{p_a^{3/2} n B} \right] \right) \\ = \mathcal{O} \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \right. \right. \\ \left. \left. + \frac{\sigma^2}{\sqrt{p_a} n B} + \frac{\varepsilon}{\sqrt{p_a}} \right] \right) \\ = \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} + \frac{1}{\sqrt{p_a}} \right).$$

692 Using  $\frac{\sigma^2}{n\varepsilon B} \geq 1$ , we can conclude the proof of the inequality. The number of stochastic gradients that  
693 each node calculates equals  $B_{\text{init}} + 2BT = \mathcal{O}(B_{\text{init}} + BT)$ .  $\square$

694 **Corollary 4.** Suppose that assumptions of Corollary 3 hold, batch size  $B \leq \min \left\{ \frac{\sigma}{p_a \sqrt{\varepsilon n}}, \frac{L_\sigma^2}{\mathbb{1}_{p_a}^2 \hat{L}^2} \right\}$ ,  
695 we take RandK compressors with  $K = \Theta \left( \frac{Bd\sqrt{\varepsilon n}}{\sigma} \right)$ . Then the communication complexity equals

$$\mathcal{O} \left( \frac{d\sigma}{\sqrt{p_a} \sqrt{n\varepsilon}} + \frac{L_\sigma \Delta_0 d}{p_a \sqrt{n\varepsilon}} \right), \quad (12)$$

696 and the expected number of stochastic gradient calculations per node equals

$$\mathcal{O} \left( \frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{L_\sigma \Delta_0 \sigma}{p_a \varepsilon^{3/2} n} \right). \quad (13)$$

697 *Proof.* The communication complexity equals

$$\mathcal{O}(d + KT) = \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ KL + K \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + K \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right] + K \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right).$$

698 Due to  $B \leq \frac{L_\sigma^2}{\mathbb{1}_{p_a}^2 \hat{L}^2}$ , we have  $\mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \leq \frac{2L_\sigma}{\sqrt{B}}$  and

$$\mathcal{O}(d + KT) = \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ KL + K \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + K \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \frac{L_\sigma}{\sqrt{B}} \right] + K \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right).$$

699 From Theorem 6, we have  $\omega + 1 = \frac{d}{K}$ . Since  $K = \Theta\left(\frac{Bd\sqrt{\varepsilon n}}{\sigma}\right) = \mathcal{O}\left(\frac{d}{p_a \sqrt{n}}\right)$ , the communication  
700 complexity equals

$$\begin{aligned} \mathcal{O}(d + KT) &= \mathcal{O}\left(d + \frac{\Delta_0}{\varepsilon} \left[ \frac{d}{p_a \sqrt{n}} L + \frac{d}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{d}{p_a \sqrt{n}} L_\sigma \right] + \frac{d\sigma}{\sqrt{p_a} \sqrt{n} \varepsilon} \right) \\ &= \mathcal{O}\left(\frac{d\sigma}{\sqrt{p_a} \sqrt{n} \varepsilon} + \frac{L_\sigma \Delta_0 d}{p_a \sqrt{n} \varepsilon}\right) \end{aligned}$$

701 And the expected number of stochastic gradient calculations per node equals

$$\begin{aligned} &\mathcal{O}(B_{\text{init}} + BT) \\ &= \mathcal{O}\left(\frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{B\omega}{\sqrt{p_a}} \sqrt{\frac{\sigma^2}{n \varepsilon B}} + \frac{\Delta_0}{\varepsilon} \left[ BL + B \frac{\omega}{p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + B \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \left( \mathbb{1}_{p_a} \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right] + B \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right) \\ &= \mathcal{O}\left(\frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{Bd}{K \sqrt{p_a}} \sqrt{\frac{\sigma^2}{n \varepsilon B}} + \frac{\Delta_0}{\varepsilon} \left[ BL + B \frac{d}{K p_a \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + B \sqrt{\frac{\sigma^2}{p_a^2 \varepsilon n^2 B}} \frac{L_\sigma}{\sqrt{B}} \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon} \right) \\ &= \mathcal{O}\left(\frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon \sqrt{B}} + \frac{\Delta_0}{\varepsilon} \left[ \frac{\sigma}{p_a \sqrt{\varepsilon n}} L + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} L_\sigma \right] \right) \\ &= \mathcal{O}\left(\frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{L_\sigma \Delta_0 \sigma}{p_a \varepsilon^{3/2} n}\right). \end{aligned}$$

702

□

## F Analysis of DASHA-PP under Polyak-Łojasiewicz Condition

In this section, we provide the theoretical convergence rates of DASHA-PP under Polyak-Łojasiewicz Condition.

**Assumption 9.** The function  $f$  satisfy (Polyak-Łojasiewicz) PL-condition:

$$\|\nabla f(x)\|^2 \geq 2\mu(f(x) - f^*), \quad \forall x \in \mathbb{R}, \quad (31)$$

where  $f^* = \inf_{x \in \mathbb{R}^d} f(x) > -\infty$ .

Under Polyak-Łojasiewicz condition, a (random) point  $\hat{x}$  is  $\varepsilon$ -solution, if  $\mathbb{E}[f(\hat{x})] - f^* \leq \varepsilon$ .

We now provide the convergence rates of DASHA-PP under PL-condition.

### F.1 Gradient Setting

**Theorem 8.** Suppose that Assumption 1, 2, 3, 7, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_a}{2-p_a}$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega+1)}{np_a^2} + \frac{48}{np_a^2} \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}} \right)^{-1}, \frac{a}{4\mu} \right\},$$

and  $h_i^0 = g_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP), then  $\mathbb{E}[f(x^T)] - f^* \leq (1 - \gamma\mu)^T \Delta_0$ .

Let us provide bounds up to logarithmic factors and use  $\tilde{\mathcal{O}}(\cdot)$  notation. The provided theorem states that to get  $\varepsilon$ -solution DASHA-PP have to run

$$\tilde{\mathcal{O}} \left( \frac{\omega+1}{p_a} + \frac{L}{\mu} + \frac{\omega \hat{L}}{p_a \mu \sqrt{n}} + \frac{\hat{L}}{p_a \mu \sqrt{n}} \right),$$

communication rounds. The method DASHA from (Tyurin and Richtárik, 2023), have to run

$$\tilde{\mathcal{O}} \left( \omega + \frac{L}{\mu} + \frac{\omega \hat{L}}{\mu \sqrt{n}} \right),$$

communication rounds to get  $\varepsilon$ -solution. The difference is the same as in the general nonconvex case (see Section 6.1). Up to Lipschitz constants factors, we get the degeneration up to  $1/p_a$  factor due to the partial participation.

### F.2 Finite-Sum Setting

**Theorem 9.** Suppose that Assumption 1, 2, 3, 7, 4, 8, and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ , probability  $p_{page} = \frac{B}{m+B}$ ,  $b = \frac{p_{page} p_a}{2-p_a}$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{(1-p_{page})L_{\max}^2}{B} \right) + \frac{48}{np_a^2 p_{page}} \left( \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 + \frac{(1-p_{page})L_{\max}^2}{B} \right)} \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

and  $h_i^0 = g_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-PAGE), then  $\mathbb{E}[f(x^T)] - f^* \leq (1 - \gamma\mu)^T \Delta_0$ .

The provided theorem states that to get  $\varepsilon$ -solution DASHA-PP have to run

$$\tilde{\mathcal{O}} \left( \frac{\omega+1}{p_a} + \frac{m}{p_a B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{\sqrt{m}}{p_a \mu \sqrt{n} B} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) \right),$$

communication rounds. The method DASHA-PAGE from (Tyurin and Richtárik, 2023), have to run

$$\tilde{\mathcal{O}} \left( \omega + \frac{m}{B} + \frac{L}{\mu} + \frac{\omega}{\mu \sqrt{n}} \left( \hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{\sqrt{m}}{\mu \sqrt{n} B} \left( \frac{L_{\max}}{\sqrt{B}} \right) \right),$$

communication rounds to get  $\varepsilon$ -solution. We can guarantee the degeneration up to  $1/p_a$  factor due to the partial participation only if  $B = \mathcal{O}\left(\frac{L_{\max}^2}{L^2}\right)$ . The same conclusion we have in Section 6.2.



### 726 F.3 Stochastic Setting

**Theorem 10.** Suppose that Assumption 1, 2, 3, 7, 5, 6, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
 $b \in \left(0, \frac{p_a}{2-p_a}\right]$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega+1)}{np_a^2} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \hat{L}^2 \right)} + \frac{40}{np_a b} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 \right) \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

727 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-MVR), then

$$\begin{aligned} & \mathbb{E} [f(x^T) - f^*] \\ & \leq (1 - \gamma\mu)^T \left( \Delta_0 + \frac{2\gamma}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{40\gamma b\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \\ & \quad + \frac{1}{\mu} \left( \frac{100b^2\omega(2\omega+1)}{p_a^2} + \frac{20b}{p_a} \right) \frac{\sigma^2}{nB}. \end{aligned}$$

728 The provided theorems states that to get  $\varepsilon$ -solution DASHA-PP have to run

$$\tilde{\mathcal{O}} \left( \underbrace{\frac{\omega+1}{p_a} + \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}}}_{\mathcal{P}_2} + \frac{\sigma^2}{p_a \mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \underbrace{\frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right)}_{\mathcal{P}_1} \right) \quad (32)$$

729 communication rounds. We take  $b = \Theta \left( \min \left\{ \frac{p_a}{\omega} \sqrt{\frac{\mu n \varepsilon B}{\sigma^2}}, \frac{p_a \mu n \varepsilon B}{\sigma^2} \right\} \right) \geq$

730  $\Theta \left( \min \left\{ \frac{p_a}{\omega^2}, \frac{p_a \mu n \varepsilon B}{\sigma^2} \right\} \right).$

731 The method DASHA-SYNC-MVR from (Tyurin and Richtárik, 2023), have to run

$$\tilde{\mathcal{O}} \left( \omega + \frac{\sigma^2}{\mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{\mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{n \mu^{3/2} \sqrt{\varepsilon B}} \left( \frac{L_\sigma}{\sqrt{B}} \right) \right) \quad (33)$$

732 communication rounds to get  $\varepsilon$ -solution<sup>7</sup>.

733 In the stochastic setting, the comparison is a little bit more complicated. As in the finite-sum setting,  
 734 we have to take  $B = \mathcal{O} \left( \frac{L_\sigma^2}{\hat{L}^2} \right)$  to guarantee the degeneration up to  $1/p_a$  of the term  $\mathcal{P}_1$  from (32).

735 However, DASHA-PP-MVR has also suboptimal term  $\mathcal{P}_2$ . This suboptimality is tightly connected with  
 736 the suboptimality of  $B_{\text{init}}$  in the general nonconvex case, which we discuss in Section 6.3, and it also  
 737 appears in the analysis of DASHA-MVR (Tyurin and Richtárik, 2023). Let us provide the counterpart  
 738 of Corollary 4. The corollary reveals that we can escape regimes when  $\mathcal{P}_2$  is the bottleneck by  
 739 choosing the parameters of the compressors.

740 **Corollary 5.** Suppose that assumptions of Theorem 10 hold, batch size  $B \leq \min \left\{ \frac{\sigma}{p_a \sqrt{\mu \varepsilon n}}, \frac{L_\sigma^2}{\hat{L}^2} \right\}$ ,

741 we take RandK compressors with  $K = \Theta \left( \frac{B d \sqrt{\mu \varepsilon n}}{\sigma} \right)$ . Then the communication complexity equals

$$\tilde{\mathcal{O}} \left( \frac{d\sigma}{p_a \sqrt{\mu \varepsilon n}} + \frac{dL_\sigma}{p_a \mu \sqrt{n}} \right),$$

742 and the expected number of stochastic gradient calculations per node equals

$$\tilde{\mathcal{O}} \left( \frac{\sigma^2}{p_a \mu n \varepsilon} + \frac{\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon}} \right).$$

743 Up to Lipschitz constants, DASHA-PP-MVR has the state-of-the-art oracle complexity under PL-  
 744 condition (see (Li et al., 2021a)). Moreover, DASHA-PP-MVR has the state-of-the-art communication  
 745 complexity of DASHA for a small enough  $\mu$ .

<sup>7</sup>For simplicity, we omitted  $\frac{d}{\zeta_C}$  term from the complexity in the stochastic setting, where  $\zeta_C$  is defined in Definition 12. For instance, for the RandK compressor (see Definition 5 and Theorem 6),  $\zeta_C = K$  and  $\frac{d}{\zeta_C} = \Theta(\omega)$ .

#### 746 **F.4 Proofs of Theorems**

747 The following proofs almost repeat the proofs from Section E. And one of the main changes is that  
748 instead of Lemma 3, we use the following lemma.

##### 749 **F.4.1 Standard Lemma under Polyak-Łojasiewicz Condition**

750 **Lemma 11.** *Suppose that Assumptions 1 and 9 hold and*

$$\mathbb{E} [f(x^{t+1})] + \gamma \Psi^{t+1} \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] + (1 - \gamma\mu)\gamma \Psi^t + \gamma C,$$

751 *where  $\Psi^t$  is a sequence of numbers,  $\Psi^t \geq 0$  for all  $t \in [T]$ , constant  $C \geq 0$ , constant  $\mu > 0$ , and*  
752 *constant  $\gamma \in (0, 1/\mu)$ . Then*

$$\mathbb{E} [f(x^T) - f^*] \leq (1 - \gamma\mu)^T ((f(x^0) - f^*) + \gamma \Psi^0) + \frac{C}{\mu}. \quad (34)$$

753 *Proof.* We subtract  $f^*$  and use PŁ-condition (31) to get

$$\begin{aligned} \mathbb{E} [f(x^{t+1}) - f^*] + \gamma \Psi^{t+1} &\leq \mathbb{E} [f(x^t) - f^*] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] + \gamma \Psi^t + \gamma C \\ &\leq (1 - \gamma\mu) \mathbb{E} [f(x^t) - f^*] + (1 - \gamma\mu)\gamma \Psi^t + \gamma C \\ &= (1 - \gamma\mu) (\mathbb{E} [f(x^t) - f^*] + \gamma \Psi^t) + \gamma C. \end{aligned}$$

754 Unrolling the inequality, we have

$$\begin{aligned} \mathbb{E} [f(x^{t+1}) - f^*] + \gamma \Psi^{t+1} &\leq (1 - \gamma\mu)^{t+1} ((f(x^0) - f^*) + \gamma \Psi^0) + \gamma C \sum_{i=0}^t (1 - \gamma\mu)^i \\ &\leq (1 - \gamma\mu)^{t+1} ((f(x^0) - f^*) + \gamma \Psi^0) + \frac{C}{\mu}. \end{aligned}$$

755 It is left to note that  $\Psi^t \geq 0$  for all  $t \in [T]$ . □

##### 756 **F.4.2 Generic Lemma**

757 We now provide the counterpart of Lemma 6.

758 **Lemma 12.** *Suppose that Assumptions 2, 7, 8 and 9 hold and let us take  $a = \frac{p_a}{2\omega+1}$ , then*

$$\begin{aligned} &\mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ &\leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ &\quad + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ &\quad + \frac{10\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right]. \end{aligned}$$

759 *Proof.* Let us fix some constants  $\kappa, \eta \in [0, \infty)$  that we will define later. Using the same reasoning as  
760 in Lemma 6, we can get

$$\begin{aligned} &\mathbb{E} [f(x^{t+1})] \\ &\quad + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ &\leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \end{aligned}$$

$$\begin{aligned}
& + \left( \gamma + \kappa (1 - a)^2 \right) \mathbb{E} \left[ \|g^t - h^t\|^2 \right] \\
& + \left( \frac{\kappa a^2 ((2\omega + 1) p_a - p_{aa})}{np_a^2} + \eta \left( \frac{a^2 (2\omega + 1 - p_a)}{p_a} + (1 - a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( \frac{2\kappa\omega}{np_a} + \frac{2\eta\omega}{p_a} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

761 Let us take  $\kappa = \frac{2\gamma}{a}$ . One can show that  $\gamma + \kappa (1 - a)^2 \leq (1 - \frac{a}{2}) \kappa$ , and thus

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] \\
& + \frac{2\gamma}{a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma}{a} \mathbb{E} [\|g^t - h^t\|^2] \\
& + \left( \frac{2\gamma a ((2\omega + 1) p_a - p_{aa})}{np_a^2} + \eta \left( \frac{a^2 (2\omega + 1 - p_a)}{p_a} + (1 - a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( \frac{4\gamma\omega}{anp_a} + \frac{2\eta\omega}{p_a} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

762 Considering the choice of  $a$ , one can show that  $\left( \frac{a^2 (2\omega + 1 - p_a)}{p_a} + (1 - a)^2 \right) \leq 1 - a$ . If we take  
763  $\eta = \frac{4\gamma((2\omega+1)p_a-p_{aa})}{np_a^2}$ , then  $\left( \frac{2\gamma a ((2\omega + 1) p_a - p_{aa})}{np_a^2} + \eta \left( \frac{a^2 (2\omega + 1 - p_a)}{p_a} + (1 - a)^2 \right) \right) \leq (1 - \frac{a}{2}) \eta$  and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] \\
& + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1) p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left( 1 - \frac{a}{2} \right) \frac{4\gamma((2\omega + 1) p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( \frac{2\gamma(2\omega + 1)\omega}{np_a^2} + \frac{8\gamma((2\omega + 1) p_a - p_{aa})\omega}{np_a^3} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \\
& \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left( 1 - \frac{a}{2} \right) \frac{4\gamma((2\omega + 1) p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{10\gamma(2\omega + 1)\omega}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right].
\end{aligned}$$

764 It is left to consider that  $\gamma \leq \frac{a}{2\mu}$ , and therefore  $1 - \frac{a}{2} \leq 1 - \gamma\mu$ . □

765 **E.4.3 Proof for DASHA-PP under PL-condition**

**Theorem 8.** Suppose that Assumption 1, 2, 3, 7, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_a}{2-p_a}$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega+1)}{np_a^2} + \frac{48}{np_a^2} \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}} \right)^{-1}, \frac{a}{4\mu} \right\},$$

766 and  $h_i^0 = g_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP), then  $\mathbb{E}[f(x^T)] - f^* \leq$   
 767  $(1 - \gamma\mu)^T \Delta_0$ .

768 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 12, Lemma 7,  
 769 and the law of total expectation, we obtain

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}\left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2}\right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2\right] \\ & + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & + \frac{10\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E}\left[2\widehat{L}^2 \|x^{t+1} - x^t\|^2 + 2b^2 \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \nu \mathbb{E}\left[\frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \|x^{t+1} - x^t\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2\right] \\ & + \rho \mathbb{E}\left[\frac{2(1-p_a)\widehat{L}^2}{p_a} \|x^{t+1} - x^t\|^2 + \left(\frac{2b^2(1-p_a)}{p_a} + (1-b)^2\right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right]. \end{aligned}$$

770 After rearranging the terms, we get

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{20\gamma\omega(2\omega+1)\widehat{L}^2}{np_a^2} - \nu \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} - \rho \frac{2(1-p_a)\widehat{L}^2}{p_a}\right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\ & + (\gamma + \nu(1-b)^2) \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] \\ & + \left(\frac{20b^2\gamma\omega(2\omega+1)}{np_a^2} + \nu \frac{2b^2(p_a - p_{aa})}{np_a^2} + \rho \left(\frac{2b^2(1-p_a)}{p_a} + (1-b)^2\right)\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right]. \end{aligned}$$

771 By taking  $\nu = \frac{2\gamma}{b}$ , one can show that  $(\gamma + \nu(1-b)^2) \leq (1 - \frac{b}{2})\nu$ , and

$$\mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right]$$

$$\begin{aligned}
& + \frac{2\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
\leq & \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{20\gamma\omega(2\omega + 1)\hat{L}^2}{np_a^2} - \frac{4\gamma(p_a - p_{aa})\hat{L}^2}{bnp_a^2} - \rho \frac{2(1 - p_a)\hat{L}^2}{p_a} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{20b^2\gamma\omega(2\omega + 1)}{np_a^2} + \frac{4\gamma b(p_a - p_{aa})}{np_a^2} + \rho \left( \frac{2b^2(1 - p_a)}{p_a} + (1 - b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

772 Note that  $b = \frac{p_a}{2 - p_a}$ , thus

$$\begin{aligned}
& \left( \frac{20b^2\gamma\omega(2\omega + 1)}{np_a^2} + \frac{4\gamma b(p_a - p_{aa})}{np_a^2} + \rho \left( \frac{2b^2(1 - p_a)}{p_a} + (1 - b)^2 \right) \right) \\
& \leq \left( \frac{20b^2\gamma\omega(2\omega + 1)}{np_a^2} + \frac{4\gamma b(p_a - p_{aa})}{np_a^2} + \rho(1 - b) \right).
\end{aligned}$$

773 And if we take  $\rho = \frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}$ , then

$$\left( \frac{20b^2\gamma\omega(2\omega + 1)}{np_a^2} + \frac{4\gamma b(p_a - p_{aa})}{np_a^2} + \rho(1 - b) \right) \leq \left( 1 - \frac{b}{2} \right) \rho,$$

774 and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
\leq & \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{20\gamma\omega(2\omega + 1)\hat{L}^2}{np_a^2} - \frac{4\gamma(p_a - p_{aa})\hat{L}^2}{bnp_a^2} \right. \\
& \quad \left. - \frac{80b\gamma\omega(2\omega + 1)(1 - p_a)\hat{L}^2}{np_a^3} - \frac{16\gamma(p_a - p_{aa})(1 - p_a)\hat{L}^2}{np_a^3} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( 1 - \frac{b}{2} \right) \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

775 Due to  $\frac{p_a}{2} \leq b \leq p_a$ , we have

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{100\gamma\omega(2\omega + 1)\hat{L}^2}{np_a^2} - \frac{24\gamma(p_a - p_{aa})\hat{L}^2}{np_a^3}\right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left(\frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
\end{aligned}$$

776 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
&\mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
&\quad + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left(\frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
\end{aligned}$$

777 Note that  $\gamma \leq \frac{a}{4\mu} \leq \frac{p_a}{4\mu} \leq \frac{b}{2\mu}$ , thus  $1 - \frac{b}{2} \leq 1 - \gamma\mu$  and

$$\begin{aligned}
&\mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
&\quad + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad + (1 - \gamma\mu) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + (1 - \gamma\mu) \left(\frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
\end{aligned}$$

778 In the view of Lemma 11 with

$$\begin{aligned}
\Psi^t &= \frac{2(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{4((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad + \frac{2}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(\frac{40b\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right],
\end{aligned}$$

779 we can conclude the proof of the theorem.  $\square$

#### 780 F.4.4 Proof for DASHA-PP-PAGE under PL-condition

**Theorem 9.** Suppose that Assumption 1, 2, 3, 7, 4, 8, and 9 hold. Let us take  $a = \frac{p_a}{2\omega + 1}$ , probability

$$p_{\text{page}} = \frac{B}{m+B}, b = \frac{p_{\text{page}} p_a}{2-p_a},$$

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega + 1)}{np_a^2} \left( \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right)} + \frac{48}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

781 and  $h_i^0 = g_i^0 = \nabla f_i(x^0)$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-PAGE), then  $\mathbb{E}[f(x^T)] - f^* \leq$   
 782  $(1 - \gamma\mu)^T \Delta_0$ .

783 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 12, Lemma 8,  
 784 and the law of total expectation, we obtain

$$\begin{aligned}
 & \mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
 & + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
 & \leq \mathbb{E}\left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2}\right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2\right] \\
 & + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
 & + \frac{10\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2\right] \\
 & + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
 & \leq \mathbb{E}\left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2}\right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2\right] \\
 & + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
 & + \frac{10\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E}\left[\left(2\hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B}\right) \|x^{t+1} - x^t\|^2 + \frac{2b^2}{p_{\text{page}}} \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
 & + \nu \mathbb{E}\left[\left(\frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} + \frac{(1 - p_{\text{page}})L_{\max}^2}{np_a B}\right) \|x^{t+1} - x^t\|^2\right. \\
 & \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left(p_{\text{page}} \left(1 - \frac{b}{p_{\text{page}}}\right)^2 + (1 - p_{\text{page}})\right) \|h^t - \nabla f(x^t)\|^2\right] \\
 & + \rho \mathbb{E}\left[\left(\frac{2(1 - p_a)\hat{L}^2}{p_a} + \frac{(1 - p_{\text{page}})L_{\max}^2}{p_a B}\right) \|x^{t+1} - x^t\|^2\right. \\
 & \quad \left. + \left(\frac{2(1 - p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left(1 - \frac{b}{p_{\text{page}}}\right)^2 + (1 - p_{\text{page}})\right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right].
 \end{aligned}$$

785 After rearranging the terms, we get

$$\begin{aligned}
 & \mathbb{E}[f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
 & + \nu \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
 & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
 & + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right]
 \end{aligned}$$

$$\begin{aligned}
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right. \\
& \quad \left. - \nu \left( \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} + \frac{(1-p_{\text{page}})L_{\max}^2}{np_a B} \right) - \rho \left( \frac{2(1-p_a)\hat{L}^2}{p_a} + \frac{(1-p_{\text{page}})L_{\max}^2}{p_a B} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( \gamma + \nu \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \right) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{2\nu(p_a - p_{aa})b^2}{np_a^2 p_{\text{page}}} \right. \\
& \quad \left. + \rho \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Due to  $b = \frac{p_{\text{page}} p_a}{2-p_a} \leq p_{\text{page}}$ , one can show that  $\left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \leq 1-b$ . Thus, if we take  $\nu = \frac{2\gamma}{b}$ , then

$$\left( \gamma + \nu \left( p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \right) \leq \gamma + \nu(1-b) = \left( 1 - \frac{b}{2} \right) \nu,$$

786 therefore

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) \right. \\
& \quad \left. - \frac{2\gamma}{bn p_a} \left( 2 \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1-p_{\text{page}})L_{\max}^2}{B} \right) - \rho \left( \frac{2(1-p_a)\hat{L}^2}{p_a} + \frac{(1-p_{\text{page}})L_{\max}^2}{p_a B} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{4\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{page}}} \right. \\
& \quad \left. + \rho \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Next, with the choice of  $b = \frac{p_{\text{page}} p_a}{2-p_a}$ , we ensure that

$$\left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \leq 1-b.$$

If we take  $\rho = \frac{40b^2\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}}$ , then

$$\left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{4\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{page}}} + \rho \left( \frac{2(1-p_a)b^2}{p_a p_{\text{page}}} + p_{\text{page}} \left( 1 - \frac{b}{p_{\text{page}}} \right)^2 + (1-p_{\text{page}}) \right) \right) \leq \left( 1 - \frac{b}{2} \right) \rho,$$



787 therefore

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( 2\hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \right. \\
& \quad \left. - \frac{2\gamma}{bnp_a} \left( 2 \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \right. \\
& \quad \left. - \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^3 p_{\text{page}}} + \frac{8\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2(1 - p_a) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \right] \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( 1 - \frac{b}{2} \right) \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

Let us simplify the inequality. First, due to  $b \geq \frac{p_{\text{page}} p_a}{2}$ , we have

$$\frac{2\gamma}{bnp_a} \left( 2 \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \leq \frac{8\gamma}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right).$$

788 Second, due to  $b \leq p_a p_{\text{page}}$  and  $p_{aa} \leq p_a^2$ , we get

$$\begin{aligned}
& \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^3 p_{\text{page}}} + \frac{8\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2(1 - p_a) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \left( \frac{40\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2 p_{\text{page}}} \right) \left( 2 \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \frac{80\gamma\omega(2\omega+1)}{np_a^2} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \quad + \frac{16\gamma \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \leq \frac{80\gamma\omega(2\omega+1)}{np_a^2} \left( \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right) \\
& \quad + \frac{16\gamma}{np_a^2 p_{\text{page}}} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\text{max}}^2}{B} \right).
\end{aligned}$$

789 Combining all bounds together, we obtain the following inequality:

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2]
\end{aligned}$$

$$\begin{aligned}
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{100\gamma\omega(2\omega + 1)}{np_a^2} \left( \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \right. \\
& \quad \left. - \frac{24\gamma}{np_a^2 p_{\text{page}}} \left( \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 + \frac{(1 - p_{\text{page}})L_{\max}^2}{B} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

790 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

791 Note that  $\gamma \leq \frac{b}{2\mu}$ , thus  $1 - \frac{b}{2} \leq 1 - \gamma\mu$  and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + (1 - \gamma\mu) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + (1 - \gamma\mu) \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right].
\end{aligned}$$

792 It is left to apply Lemma 11 with

$$\begin{aligned}
\Psi^t &= \frac{2(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{4((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{2}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left( \frac{40b\omega(2\omega + 1)}{np_a^2 p_{\text{page}}} + \frac{8(p_a - p_{aa})}{np_a^2 p_{\text{page}}} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]
\end{aligned}$$

793 to conclude the proof.  $\square$

794 **F.4.5 Proof for DASHA-PP-MVR under PL-condition**

**Theorem 10.** Suppose that Assumption 1, 2, 3, 7, 5, 6, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
 $b \in \left(0, \frac{p_a}{2-p_a}\right]$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{200\omega(2\omega+1)}{np_a^2} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \widehat{L}^2 \right) + \frac{40}{np_a b} \left( \frac{(1-b)^2 L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right)} \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

795 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 1 (DASHA-PP-MVR), then

$$\begin{aligned} & \mathbb{E} [f(x^T) - f^*] \\ & \leq (1 - \gamma\mu)^T \left( \Delta_0 + \frac{2\gamma}{b} \|h^0 - \nabla f(x^0)\|^2 + \left( \frac{40\gamma b\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) \\ & \quad + \frac{1}{\mu} \left( \frac{100b^2\omega(2\omega+1)}{p_a^2} + \frac{20b}{p_a} \right) \frac{\sigma^2}{nB}. \end{aligned}$$

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & \quad + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & \quad + \frac{10\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right]. \end{aligned}$$

796 *Proof.* Let us fix constants  $\nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 12, Lemma 10,  
 797 and the law of total expectation, we obtain

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & \quad + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & \quad + \frac{10\gamma(2\omega+1)\omega}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|k_i^{t+1}\|^2 \right] \\ & \quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \|h^t - \nabla f(x^t)\|^2 \right] \\ & \quad + (1 - \gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \end{aligned}$$

$$\begin{aligned}
& + \frac{10\gamma\omega(2\omega+1)}{np_a^2} \mathbb{E} \left[ \frac{2b^2\sigma^2}{B} + \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 + 2b^2 \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \nu \mathbb{E} \left( \frac{2b^2\sigma^2}{np_a B} + \left( \frac{2(1-b)^2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \right) \\
& + \rho \mathbb{E} \left( \frac{2b^2\sigma^2}{p_a B} + \left( \frac{2(1-b)^2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right. \\
& \quad \left. + \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right).
\end{aligned}$$

798 After rearranging the terms, we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right. \\
& \quad \left. - \nu \left( \frac{2(1-b)^2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2(1-b)^2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + (\gamma + \nu(1-b)^2) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\nu(p_a - p_{aa})b^2}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2} + \nu \frac{2b^2}{np_a} + \rho \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

799 By taking  $\nu = \frac{2\gamma}{b}$ , one can show that  $(\gamma + \nu(1-b)^2) \leq (1 - \frac{b}{2})\nu$ , and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right)
\end{aligned}$$

$$\begin{aligned}
& -\frac{2\gamma}{b} \left( \frac{2(1-b)^2 L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa}) \hat{L}^2}{np_a^2} \right) - \rho \left( \frac{2(1-b)^2 L_\sigma^2}{p_a B} + \frac{2(1-p_a) \hat{L}^2}{p_a} \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
& + \left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma (p_a - p_{aa}) b}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma b}{np_a} + \rho \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

800 Note that  $b \leq \frac{p_a}{2-p_a}$ , thus

$$\begin{aligned}
& \left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma (p_a - p_{aa}) b}{np_a^2} + \rho \left( \frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \right) \\
& \leq \left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma (p_a - p_{aa}) b}{np_a^2} + \rho (1-b) \right).
\end{aligned}$$

801 And if we take  $\rho = \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a-p_{aa})}{np_a^2}$ , then

$$\left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma (p_a - p_{aa}) b}{np_a^2} + \rho (1-b) \right) \leq \rho,$$

802 and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a-p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{10\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2\hat{L}^2 \right) \right. \\
& \quad \left. - \frac{2\gamma}{np_a b} \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2 \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^3} + \frac{8\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2} \right) \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 2(1-p_a) \hat{L}^2 \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a-p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{20b^2 \gamma \omega (2\omega + 1)}{np_a^2} + \frac{4\gamma b}{np_a} + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a-p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

803 Let us simplify the inequality. First, due to  $b \leq p_a$  and  $(1-p_a) \leq \left(1 - \frac{p_{aa}}{p_a}\right)$ , we have

$$\left( \frac{40b\gamma\omega(2\omega+1)}{np_a^3} + \frac{2\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2} \right) \left( \frac{2(1-b)^2 L_\sigma^2}{B} + 8(1-p_a) \hat{L}^2 \right)$$

$$\begin{aligned}
&= \frac{40b\gamma\omega(2\omega+1)}{np_a^3} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2(1-p_a)\widehat{L}^2 \right) \\
&\quad + \frac{8\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2(1-p_a)\widehat{L}^2 \right) \\
&\leq \frac{40\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \\
&\quad + \frac{8\gamma}{np_ab} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2 \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right),
\end{aligned}$$

804 therefore

$$\begin{aligned}
&\mathbb{E}[f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\
&\quad + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{50\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2\widehat{L}^2 \right) \right. \\
&\quad \left. - \frac{10\gamma}{np_ab} \left( \frac{2(1-b)^2L_\sigma^2}{B} + 2 \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
&\quad + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{4\gamma b}{np_a} + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B} \\
&\leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
&\quad + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{100\gamma\omega(2\omega+1)}{np_a^2} \left( \frac{(1-b)^2L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \frac{20\gamma}{np_ab} \left( \frac{(1-b)^2L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
&\quad + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\
&\quad + \left( \frac{20b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{4\gamma b}{np_a} + \left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

805 Also, we can simplify the last term:

$$\begin{aligned}
&\left( \frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \frac{2b^2}{p_a} \\
&= \frac{80b^3\gamma\omega(2\omega+1)}{np_a^3} + \frac{16b^2\gamma \left(1 - \frac{p_{aa}}{p_a}\right)}{np_a^2}
\end{aligned}$$

$$\leq \frac{80b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{16b\gamma}{np_a},$$

806 thus

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{100\gamma\omega(2\omega+1)}{np_a^2} \left(\frac{(1-b)^2 L_\sigma^2}{B} + \widehat{L}^2\right)\right. \\ & \quad \left. - \frac{20\gamma}{np_a b} \left(\frac{(1-b)^2 L_\sigma^2}{B} + \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2\right)\right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\ & + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left(\frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \left(\frac{100b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{20\gamma b}{np_a}\right) \frac{\sigma^2}{B}. \end{aligned}$$

807 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \\ & + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \left(\frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2\right] \\ & + \left(\frac{100b^2\gamma\omega(2\omega+1)}{np_a^2} + \frac{20\gamma b}{np_a}\right) \frac{\sigma^2}{B}. \end{aligned}$$

808 Note that  $\gamma \leq \frac{b}{2\mu}$ , thus  $1 - \frac{b}{2} \leq 1 - \gamma\mu$  and

$$\begin{aligned} & \mathbb{E}[f(x^{t+1})] + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2\right] \\ & + \frac{2\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \left(\frac{40b\gamma\omega(2\omega+1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2}\right) \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2\right] \\ & \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\ & + (1-\gamma\mu) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + (1-\gamma\mu) \frac{4\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2\right] \end{aligned}$$

$$\begin{aligned}
& + (1 - \gamma\mu) \frac{2\gamma}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] + (1 - \gamma\mu) \left( \frac{40b\gamma\omega(2\omega + 1)}{np_a^2} + \frac{8\gamma(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{100b^2\gamma\omega(2\omega + 1)}{np_a^2} + \frac{20\gamma b}{np_a} \right) \frac{\sigma^2}{B}.
\end{aligned}$$

809 It is left to apply Lemma 11 with

$$\begin{aligned}
\Psi^t &= \frac{2(2\omega + 1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{4((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&+ \frac{2}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] + \left( \frac{40b\omega(2\omega + 1)}{np_a^2} + \frac{8(p_a - p_{aa})}{np_a^2} \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]
\end{aligned}$$

810 and  $C = \left( \frac{100b^2\omega(2\omega+1)}{p_a^2} + \frac{20b}{p_a} \right) \frac{\sigma^2}{nB}$  to conclude the proof.  $\square$

811 **Corollary 5.** Suppose that assumptions of Theorem 10 hold, batch size  $B \leq \min \left\{ \frac{\sigma}{p_a\sqrt{\mu\varepsilon n}}, \frac{L_\sigma^2}{L^2} \right\}$ ,  
812 we take RandK compressors with  $K = \Theta \left( \frac{Bd\sqrt{\mu\varepsilon n}}{\sigma} \right)$ . Then the communication complexity equals

$$\tilde{\mathcal{O}} \left( \frac{d\sigma}{p_a\sqrt{\mu\varepsilon n}} + \frac{dL_\sigma}{p_a\mu\sqrt{n}} \right),$$

813 and the expected number of stochastic gradient calculations per node equals

$$\tilde{\mathcal{O}} \left( \frac{\sigma^2}{p_a\mu n\varepsilon} + \frac{\sigma L_\sigma}{p_a n\mu^{3/2}\sqrt{\varepsilon}} \right).$$

814 *Proof.* In the view of Theorem 10, DASHA-PP have to run

$$\tilde{\mathcal{O}} \left( K \frac{\omega + 1}{p_a} + \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + \frac{\sigma^2}{p_a \mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right)$$

815 communication rounds in the stochastic settings to get  $\varepsilon$ -solution. Note that  $K = \mathcal{O} \left( \frac{d}{p_a \sqrt{n}} \right)$ .

816 Moreover, we can skip the initialization procedure and initialize  $h_i^0$  and  $g_i^0$ , for instance, with zeros  
817 because the initialization error is under a logarithm. Considering Theorem 6, the communication  
818 complexity equals

$$\begin{aligned}
& \tilde{\mathcal{O}} \left( K \frac{\omega + 1}{p_a} + K \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + K \frac{\sigma^2}{p_a \mu n \varepsilon B} + K \frac{L}{\mu} + K \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + K \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right) \\
&= \tilde{\mathcal{O}} \left( K \frac{\omega + 1}{p_a} + K \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + K \frac{\sigma^2}{p_a \mu n \varepsilon B} + K \frac{L}{\mu} + K \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + K \frac{\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \right) \\
&= \tilde{\mathcal{O}} \left( \frac{d}{p_a} + \frac{d}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + \frac{K\sigma^2}{p_a \mu n \varepsilon B} + \frac{dL}{p_a \mu \sqrt{n}} + \frac{d}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{K\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \right) \\
&= \tilde{\mathcal{O}} \left( \frac{d}{p_a} + \frac{d\sigma}{p_a \sqrt{\mu n \varepsilon B}} + \frac{d\sigma}{p_a \sqrt{\mu \varepsilon n}} + \frac{dL}{p_a \mu \sqrt{n}} + \frac{d}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{dL_\sigma}{p_a \mu \sqrt{n}} \right) \\
&= \tilde{\mathcal{O}} \left( \frac{d\sigma}{p_a \sqrt{\mu \varepsilon n}} + \frac{dL_\sigma}{p_a \mu \sqrt{n}} \right).
\end{aligned}$$

819 The expected number of stochastic gradient calculations per node equals

$$\tilde{\mathcal{O}} \left( B \frac{\omega + 1}{p_a} + B \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + B \frac{\sigma^2}{p_a \mu n \varepsilon B} + B \frac{L}{\mu} + B \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + B \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) \right)$$



$$\begin{aligned}
&= \tilde{\mathcal{O}} \left( B \frac{\omega + 1}{p_a} + B \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + B \frac{\sigma^2}{p_a \mu n \varepsilon B} + B \frac{L}{\mu} + B \frac{\omega}{p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + B \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left( \frac{L_\sigma}{\sqrt{B}} \right) \right) \\
&= \tilde{\mathcal{O}} \left( \frac{Bd}{K p_a} + \frac{Bd}{K p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}} + \frac{\sigma^2}{p_a \mu n \varepsilon} + B \frac{L}{\mu} + \frac{Bd}{K p_a \mu \sqrt{n}} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon}} \right) \\
&= \tilde{\mathcal{O}} \left( \frac{\sigma}{p_a \sqrt{\mu \varepsilon n}} + \frac{\sigma^2}{p_a \mu \varepsilon n \sqrt{B}} + \frac{\sigma^2}{p_a \mu n \varepsilon} + \frac{\sigma L}{p_a \mu^{3/2} \sqrt{\varepsilon} n} + \frac{\sigma}{p_a \mu^{3/2} \sqrt{\varepsilon} n} \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon}} \right) \\
&= \tilde{\mathcal{O}} \left( \frac{\sigma^2}{p_a \mu n \varepsilon} + \frac{\sigma L_\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon}} \right).
\end{aligned}$$

820

□

## 821 G Description of DASHA-PP-SYNC-MVR

822 By analogy to (Tyurin and Richtárik, 2023), we provide a “synchronized” version of the algorithm.  
 823 With a small probability, participating nodes calculate and send a mega batch without compression.  
 824 This helps us to resolve the suboptimality of DASHA-PP-MVR w.r.t.  $\omega$ . Note that this suboptimality is  
 825 not a problem. We show in Corollary 4 that DASHA-PP-MVR can have the optimal oracle complexity  
 826 and SOTA communication complexity with the particular choices of parameters of the compressors.

---

### Algorithm 8 DASHA-PP-SYNC-MVR

---

```

1: Input: starting point  $x^0 \in \mathbb{R}^d$ , stepsize  $\gamma > 0$ , momentum  $a \in (0, 1]$ , momentum  $b \in$ 
   (0, 1], probability  $p_{\text{mega}} \in (0, 1]$ , batch size  $B'$  and  $B$ , probability  $p_a \in (0, 1]$  that a node is
   participating(a), number of iterations  $T \geq 1$ .
2: Initialize  $g_i^0, h_i^0$  on the nodes and  $g^0 = \frac{1}{n} \sum_{i=1}^n g_i^0$  on the server
3: for  $t = 0, 1, \dots, T - 1$  do
4:    $x^{t+1} = x^t - \gamma g^t$ 
5:    $c^{t+1} = \begin{cases} 1, & \text{with probability } p_{\text{mega}}, \\ 0, & \text{with probability } 1 - p_{\text{mega}} \end{cases}$ 
6:   Broadcast  $x^{t+1}, x^t$  to all participating(a) nodes
7:   for  $i = 1, \dots, n$  in parallel do
8:     if  $i^{\text{th}}$  node is participating(a) then
9:       if  $c^{t+1} = 1$  then
10:        Generate i.i.d. samples  $\{\xi_{ik}^{t+1}\}_{k=1}^{B'}$  of size  $B'$  from  $\mathcal{D}_i$ .
11:         $k_i^{t+1} = \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) - \frac{b}{p_{\text{mega}}} \left( h_i^t - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) \right)$ 
12:         $m_i^{t+1} = \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t)$ 
13:       else
14:        Generate i.i.d. samples  $\{\xi_{ij}^{t+1}\}_{j=1}^B$  of size  $B$  from  $\mathcal{D}_i$ .
15:         $k_i^{t+1} = \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1})$ 
16:         $m_i^{t+1} = C_i \left( \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right)$ 
17:       end if
18:        $h_i^{t+1} = h_i^t + \frac{1}{p_a} k_i^{t+1}$ 
19:        $g_i^{t+1} = g_i^t + m_i^{t+1}$ 
20:       Send  $m_i^{t+1}$  to the server
21:     else
22:        $h_i^{t+1} = h_i^t$ 
23:        $m_i^{t+1} = 0$ 
24:        $g_i^{t+1} = g_i^t$ 
25:     end if
26:   end for
27:    $g^{t+1} = g^t + \frac{1}{n} \sum_{i=1}^n m_i^{t+1}$ 
28: end for
29: Output:  $\hat{x}^T$  chosen uniformly at random from  $\{x^t\}_{k=0}^{T-1}$ 
   (a): For the formal description see Section 2.2.

```

---

827 In the following theorem, we provide the convergence rate of DASHA-PP-SYNC-MVR.

**Theorem 11.** Suppose that Assumptions 1, 2, 3, 5, 6, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
 $b = \frac{p_{\text{mega}} p_a}{2-p_a}$ , probability  $p_{\text{mega}} \in (0, 1]$ , batch size  $B' \geq B \geq 1$

$$\gamma \leq \left( L + \sqrt{\frac{8(2\omega+1)\omega}{np_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right) + \frac{16}{np_{\text{mega}} p_a^2} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{L_\sigma^2}{B} \right)} \right)^{-1},$$

828 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 8. Then

$$\begin{aligned} \mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] &\leq \frac{1}{T} \left[ \frac{2\Delta_0}{\gamma} + \frac{4}{p_{\text{mega}} p_a} \|h^0 - \nabla f(x^0)\|^2 + \frac{4 \left(1 - \frac{p_{aa}}{p_a}\right)}{n p_{\text{mega}} p_a} \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right] \\ &\quad + \frac{12\sigma^2}{nB'}. \end{aligned}$$

829 First, we introduce the expected density of compressors (Gorbunov et al., 2021; Tyurin and Richtárik, 2023).

831 **Definition 12.** The expected density of the compressor  $\mathcal{C}_i$  is  $\zeta_{\mathcal{C}_i} := \sup_{x \in \mathbb{R}^d} \mathbb{E} [\|\mathcal{C}_i(x)\|_0]$ , where  
832  $\|x\|_0$  is the number of nonzero components of  $x \in \mathbb{R}^d$ . Let  $\zeta_{\mathcal{C}} = \max_{i \in [n]} \zeta_{\mathcal{C}_i}$ .

833 Note that  $\zeta_{\mathcal{C}}$  is finite and  $\zeta_{\mathcal{C}} \leq d$ .

834 In the next corollary, we choose particular algorithm parameters to reveal the communication and  
835 oracle complexity.

**Corollary 6.** Suppose that assumptions from Theorem 11 hold, probability  $p_{\text{mega}} = \min \left\{ \frac{\zeta_{\mathcal{C}}}{d}, \frac{n\varepsilon B}{\sigma^2} \right\}$ ,  
batch size  $B' = \Theta \left( \frac{\sigma^2}{n\varepsilon} \right)$ , and  $h_i^0 = g_i^0 = \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0)$  for all  $i \in [n]$ , initial batch size  
 $B_{\text{init}} = \Theta \left( \frac{B}{p_{\text{mega}} \sqrt{p_a}} \right) = \Theta \left( \max \left\{ \frac{Bd}{\sqrt{p_a} \zeta_{\mathcal{C}}}, \frac{\sigma^2}{\sqrt{p_a} n\varepsilon} \right\} \right)$ , then DASHA-PP-SYNC-MVR needs

$$T := \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \left( \frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_{\mathcal{C}} n}} \right) \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon} n} \left( \frac{\hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right).$$

836 communication rounds to get an  $\varepsilon$ -solution, the expected communication complexity is equal to  
837  $\mathcal{O}(d + \zeta_{\mathcal{C}} T)$ , and the expected number of stochastic gradient calculations per node equals  $\mathcal{O}(B_{\text{init}} +$   
838  $BT)$ , where  $\zeta_{\mathcal{C}}$  is the expected density from Definition 12.

839 The main improvement of Corollary 6 over Corollary 3 is the size of the initial batch size  $B_{\text{init}}$ .  
840 However, Corollary 4 reveals that we can avoid regimes when DASHA-PP-MVR is suboptimal.

841 We also provide a theorem under PL-condition (see Assumption 9).

**Theorem 13.** Suppose that Assumptions 1, 2, 3, 5, 6, 7, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  
 $b = \frac{p_{\text{mega}} p_a}{2-p_a}$ , probability  $p_{\text{mega}} \in (0, 1]$ , batch size  $B' \geq B \geq 1$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{16(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \hat{L}^2 \right) + \left( \frac{48L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{24 \left(1 - \frac{p_{aa}}{p_a}\right) \hat{L}^2}{np_{\text{mega}} p_a^2} \right)} \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

842 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 8. Then

$$\begin{aligned} &\mathbb{E} [f(x^T) - f^*] \\ &\leq (1 - \gamma\mu)^T \left( \Delta_0 + \frac{2\gamma}{b} \|h^0 - \nabla f(x^0)\|^2 + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) + \frac{20\sigma^2}{\mu n B'}. \end{aligned}$$

843 Let us provide bounds up to logarithmic factors and use  $\tilde{\mathcal{O}}(\cdot)$  notation.

**Corollary 7.** Suppose that assumptions from Theorem 13 hold, probability  $p_{\text{mega}} =$   
 $\min \left\{ \frac{\zeta_{\mathcal{C}}}{d}, \frac{\mu n \varepsilon B}{\sigma^2} \right\}$ , batch size  $B' = \Theta \left( \frac{\sigma^2}{\mu n \varepsilon} \right)$  then DASHA-PP-SYNC-MVR needs

$$T := \tilde{\mathcal{O}} \left( \frac{\omega+1}{p_a} + \frac{d}{p_a \zeta_{\mathcal{C}}} + \frac{\sigma^2}{p_a \mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left( \frac{L_\sigma}{\sqrt{B}} + \hat{L} \right) + \left( \frac{\sqrt{d}}{p_a \mu \sqrt{\zeta_{\mathcal{C}} n}} + \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \right) \left( \frac{L_\sigma}{\sqrt{B}} + \hat{L} \right) \right).$$

844 communication rounds to get an  $\varepsilon$ -solution, the expected communication complexity is equal to  
 845  $\tilde{\mathcal{O}}(\zeta_c T)$ , and the expected number of stochastic gradient calculations per node equals  $\tilde{\mathcal{O}}(BT)$ ,  
 846 where  $\zeta_c$  is the expected density from Definition 12.

847 The proof of this corollary almost repeats the proof of Corollary 6. Note that we can skip the  
 848 initialization procedure and initialize  $h_i^0$  and  $g_i^0$ , for instance, with zeros because the initialization  
 849 error is under a logarithm.

850 Let us assume that  $\frac{d}{\zeta_c} = \Theta(\omega)$  (holds for the RandK compressor), then the convergence rate of  
 851 DASHA-PP-SYNC-MVR is

$$\tilde{\mathcal{O}}\left(\frac{\omega+1}{p_a} + \frac{\sigma^2}{p_a \mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left(\frac{L_\sigma}{\sqrt{B}} + \hat{L}\right) + \frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left(\frac{L_\sigma}{\sqrt{B}} + \hat{L}\right)\right). \quad (35)$$

852 Comparing (35) with the rate of DASHA-PP-MVR (32), one can see that DASHA-PP-SYNC-MVR  
 853 improves the suboptimal term  $\mathcal{P}_2$  from (32). However, Corollary 5 reveals that we can escape these  
 854 suboptimal regimes by choosing the parameter  $K$  of RandK compressors in a particular way.

### 855 G.1 Proof for DASHA-PP-SYNC-MVR

856 In this section, we provide the proof of the convergence rate for DASHA-PP-SYNC-MVR. There are  
 857 four different sources of randomness in Algorithm 8: the first one from random samples  $\xi_i^{t+1}$ , the  
 858 second one from compressors  $\{\mathcal{C}_i\}_{i=1}^n$ , the third one from availability of nodes, and the fourth one  
 859 from  $c^{t+1}$ . We define  $\mathbb{E}_k[\cdot]$ ,  $\mathbb{E}_c[\cdot]$ ,  $\mathbb{E}_{p_a}[\cdot]$  and  $\mathbb{E}_{p_{\text{mega}}}[\cdot]$  to be conditional expectations w.r.t.  $\xi_i^{t+1}$ ,  
 860  $\{\mathcal{C}_i\}_{i=1}^n$ , availability, and  $c^{t+1}$ , accordingly, conditioned on all previous randomness. Moreover, we  
 861 define  $\mathbb{E}_{t+1}[\cdot]$  to be a conditional expectation w.r.t. all randomness in iteration  $t+1$  conditioned on  
 862 all previous randomness.

863 Let us denote

$$\begin{aligned} k_{i,1}^{t+1} &:= \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) - \frac{b}{p_{\text{mega}}} \left( h_i^t - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) \right), \\ k_{i,2}^{t+1} &:= \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1}), \\ h_{i,1}^{t+1} &:= \begin{cases} h_i^t + \frac{1}{p_a} k_{i,1}^{t+1}, & i^{\text{th}} \text{ node is participating,} \\ h_i^t, & \text{otherwise,} \end{cases} \\ h_{i,2}^{t+1} &:= \begin{cases} h_i^t + \frac{1}{p_a} k_{i,2}^{t+1}, & i^{\text{th}} \text{ node is participating,} \\ h_i^t, & \text{otherwise,} \end{cases} \\ g_{i,1}^{t+1} &:= \begin{cases} g_i^t + \frac{1}{p_a} k_{i,1}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t), & i^{\text{th}} \text{ node is participating,} \\ g_i^t, & \text{otherwise,} \end{cases} \\ g_{i,2}^{t+1} &:= \begin{cases} g_i^t + \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right), & i^{\text{th}} \text{ node is participating,} \\ g_i^t, & \text{otherwise,} \end{cases} \end{aligned}$$

864  $h_1^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,1}^{t+1}$ ,  $h_2^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,2}^{t+1}$ ,  $g_1^{t+1} := \frac{1}{n} \sum_{i=1}^n g_{i,1}^{t+1}$ , and  $g_2^{t+1} := \frac{1}{n} \sum_{i=1}^n g_{i,2}^{t+1}$ .  
 865 Note, that

$$h^{t+1} = \begin{cases} h_1^{t+1}, & c^{t+1} = 1, \\ h_2^{t+1}, & c^{t+1} = 0, \end{cases}$$

866 and

$$g^{t+1} = \begin{cases} g_1^{t+1}, & c^{t+1} = 1, \\ g_2^{t+1}, & c^{t+1} = 0 \end{cases}$$

867 First, we will prove two lemmas.

868 **Lemma 13.** Suppose that Assumptions 3, 5, 7 and 8 hold and let us consider sequences  $\{g_i^{t+1}\}_{i=1}^n$   
869 and  $\{h_i^{t+1}\}_{i=1}^n$  from Algorithm 8, then

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\ & \leq \frac{2(1-p_{\text{mega}})\omega}{n^2 p_a} \sum_{i=1}^n \|k_{i,2}^{t+1}\|^2 + \left( \frac{(p_a - p_{aa})a^2}{n^2 p_a^2} + \frac{2(1-p_{\text{mega}})a^2\omega}{n^2 p_a} \right) \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \\ & \quad + (1-a)^2 \|g^t - h^t\|^2, \end{aligned}$$

870 and

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \right] \\ & \leq \frac{2(1-p_{\text{mega}})\omega}{p_a} \|k_{i,2}^{t+1}\|^2 + \left( \frac{(1-p_a)a^2}{p_a} + \frac{2(1-p_{\text{mega}})a^2\omega}{p_a} \right) \|g_i^t - h_i^t\|^2 \\ & \quad + (1-a)^2 \|g_i^t - h_i^t\|^2, \quad \forall i \in [n]. \end{aligned}$$

871 *Proof.* First, we get the bound for  $\mathbb{E}_{t+1} [\|g^{t+1} - h^{t+1}\|^2]$ :

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\ & = p_{\text{mega}} \mathbb{E}_{p_a} \left[ \|g_1^{t+1} - h_1^{t+1}\|^2 \right] + (1-p_{\text{mega}}) \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g_2^{t+1} - h_2^{t+1}\|^2 \right] \right]. \end{aligned}$$

872 Using

$$\mathbb{E}_{p_a} [g_{i,1}^{t+1} - h_{i,1}^{t+1}] = g_i^t + k_{i,1}^{t+1} - a(g_i^t - h_i^t) - h_i^t - k_{i,1}^{t+1} = (1-a)(g_i^t - h_i^t)$$

873 and

$$\mathbb{E}_{\mathcal{C}} [\mathbb{E}_{p_a} [g_{i,2}^{t+1} - h_{i,2}^{t+1}]] = g_i^t + k_{i,2}^{t+1} - a(g_i^t - h_i^t) - h_i^t - k_{i,2}^{t+1} = (1-a)(g_i^t - h_i^t),$$

874 we have

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\ & \stackrel{(17)}{=} p_{\text{mega}} \mathbb{E}_{p_a} \left[ \|g_1^{t+1} - h_1^{t+1} - \mathbb{E}_{p_a} [g_1^{t+1} - h_1^{t+1}]\|^2 \right] \\ & \quad + (1-p_{\text{mega}}) \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \|g_2^{t+1} - h_2^{t+1} - \mathbb{E}_{p_a} [g_2^{t+1} - h_2^{t+1}]\|^2 \right] \right] \\ & \quad + (1-a)^2 \|g^t - h^t\|^2. \end{aligned}$$

875 We can use Lemma 1 two times with i)  $r_i = g_i^t - h_i^t$  and  $s_i = -a(g_i^t - h_i^t)$  and ii)  $r_i = g_i^t - h_i^t$  and

876  $s_i = p_a \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_{i,2}^{t+1}$ , to obtain

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\ & \leq \frac{p_{\text{mega}} a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \\ & \quad + (1-p_{\text{mega}}) \left( \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_{\mathcal{C}} \left[ \left\| p_a \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - (k_{i,2}^{t+1} - a(g_i^t - h_i^t)) \right\|^2 \right] \right) \\ & \quad + (1-p_{\text{mega}}) \left( \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right) \\ & \quad + (1-a)^2 \|g^t - h^t\|^2 \\ & = \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \end{aligned}$$

$$\begin{aligned}
& + (1 - p_{\text{mega}}) \left( \frac{p_a}{n^2} \sum_{i=1}^n \mathbb{E}_C \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \right) \\
& + (1 - a)^2 \|g^t - h^t\|^2 \\
& \leq \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \\
& + \frac{(1 - p_{\text{mega}}) p_a \omega}{n^2} \sum_{i=1}^n \left\| \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right\|^2 \\
& + (1 - a)^2 \|g^t - h^t\|^2 \\
& = \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \\
& + \frac{(1 - p_{\text{mega}}) \omega}{n^2 p_a} \sum_{i=1}^n \|k_{i,2}^{t+1} - a (g_i^t - h_i^t)\|^2 \\
& + (1 - a)^2 \|g^t - h^t\|^2.
\end{aligned}$$

877 In the last inequality, we use Assumption 7. Next, using (16), we have

$$\begin{aligned}
& \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\
& \leq \frac{2(1 - p_{\text{mega}}) \omega}{n^2 p_a} \sum_{i=1}^n \|k_{i,2}^{t+1}\|^2 + \left( \frac{(p_a - p_{aa}) a^2}{n^2 p_a^2} + \frac{2(1 - p_{\text{mega}}) \omega a^2}{n^2 p_a} \right) \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \\
& + (1 - a)^2 \|g^t - h^t\|^2.
\end{aligned}$$

878 The second inequality can be proved almost in the same way:

$$\begin{aligned}
& \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \right] \\
& = p_{\text{mega}} \mathbb{E}_{p_a} \left[ \|g_{i,1}^{t+1} - h_{i,1}^{t+1}\|^2 \right] + (1 - p_{\text{mega}}) \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& \stackrel{(17)}{=} p_{\text{mega}} \mathbb{E}_{p_a} \left[ \|g_{i,1}^{t+1} - h_{i,1}^{t+1} - (1 - a)(g_i^t - h_i^t)\|^2 \right] + (1 - p_{\text{mega}}) \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& + p_{\text{mega}} (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 + (1 - p_{\text{mega}}) \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& + p_{\text{mega}} (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(17)}{=} \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 + (1 - p_{\text{mega}}) \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \|g_{i,2}^{t+1} - h_{i,2}^{t+1} - (1 - a)(g_i^t - h_i^t)\|^2 \right] \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[ \left\| g_i^t + \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( h_i^t + \frac{1}{p_a} k_{i,2}^{t+1} \right) - (1 - a)(g_i^t - h_i^t) \right\|^2 \right] \\
& + (1 - p_{\text{mega}}) (1 - p_a) \|g_i^t - h_i^t - (1 - a)(g_i^t - h_i^t)\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_{i,2}^{t+1} - a(g_i^t - h_i^t) \right) \right\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& + (1 - p_{\text{mega}}) (1 - p_a) a^2 \|g_i^t - h_i^t\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(17)}{=} \left( \frac{p_{\text{mega}}(1 - p_a)a^2}{p_a} + \frac{(1 - p_{\text{mega}})(1 - p_a)a^2}{p_a} \right) \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E} \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{(1 - p_a)a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E} \left[ \left\| \mathcal{C}_i \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left( \frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \leq \frac{(1 - p_a)a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + \frac{(1 - p_{\text{mega}})\omega}{p_a} \|k_{i,2}^{t+1} - a(g_i^t - h_i^t)\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(16)}{\leq} \frac{2(1 - p_{\text{mega}})\omega}{p_a} \|k_{i,2}^{t+1}\|^2 + \left( \frac{(1 - p_a)a^2}{p_a} + \frac{2(1 - p_{\text{mega}})a^2\omega}{p_a} \right) \|g_i^t - h_i^t\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2.
\end{aligned}$$

879

□

880 **Lemma 14.** Suppose that Assumptions 3, 5, 6 and 8 hold and let us consider sequence  $\{h_i^{t+1}\}_{i=1}^n$   
881 from Algorithm 8, then

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\
& \leq \frac{2b^2\sigma^2}{np_{\text{mega}}p_a B'} + \left( \frac{2p_{\text{mega}}L_\sigma^2}{np_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1 - p_{\text{mega}})L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\hat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \\
& + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2,
\end{aligned}$$

882

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\
& \leq \frac{2b^2\sigma^2}{p_a p_{\text{mega}} B'} + \left( \frac{2p_{\text{mega}}L_\sigma^2}{p_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1 - p_{\text{mega}})L_\sigma^2}{p_a B} + \frac{2(1 - p_a)L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
& + \frac{2(1 - p_a)b^2}{p_{\text{mega}}p_a} \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2, \quad \forall i \in [n],
\end{aligned}$$

883 and

$$\mathbb{E}_k \left[ \|k_{i,2}^{t+1}\|^2 \right] \leq \left( \frac{L_\sigma^2}{B} + L_i^2 \right) \|x^{t+1} - x^t\|^2, \quad \forall i \in [n],$$

884 *Proof.* First, we prove the bound for  $\mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right]$ . Using

$$\mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ h_{i,1}^{t+1} \right] \right]$$

$$\begin{aligned}
&= h_i^t + E_k \left[ \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) - \frac{b}{p_{\text{mega}}} \left( h_i^t - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) \right) \right] \\
&= h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t))
\end{aligned}$$

885 and

$$\begin{aligned}
&E_k [E_{p_a} [h_{i,2}^{t+1}]] \\
&= h_i^t + E_k \left[ \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1}) \right] \\
&= h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t),
\end{aligned}$$

886 we have

$$\begin{aligned}
&E_k [E_{p_a} [E_{p_{\text{mega}}} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]]] \\
&= p_{\text{mega}} E_k [E_{p_a} [\|h_1^{t+1} - \nabla f(x^{t+1})\|^2]] + (1 - p_{\text{mega}}) E_k [E_{p_a} [\|h_2^{t+1} - \nabla f(x^{t+1})\|^2]] \\
&\stackrel{(17)}{=} p_{\text{mega}} E_k [E_{p_a} [\|h_1^{t+1} - E_k [E_{p_a} [h_1^{t+1}]]\|^2]] + (1 - p_{\text{mega}}) E_k [E_{p_a} [\|h_2^{t+1} - E_k [E_{p_a} [h_2^{t+1}]]\|^2]] \\
&\quad + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2.
\end{aligned}$$

887 We can use Lemma 1 two times with i)  $r_i = h_i^t$  and  $s_i = k_{i,1}^{t+1}$  and ii)  $r_i = h_i^t$  and  $s_i = k_{i,2}^{t+1}$ , to  
888 obtain

$$\begin{aligned}
&E_k [E_{p_a} [E_{p_{\text{mega}}} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]]] \\
&\leq p_{\text{mega}} \left( \frac{1}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \right) \\
&\quad + (1 - p_{\text{mega}}) \left( \frac{1}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,2}^{t+1} - E_k [k_{i,2}^{t+1}]\|^2] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \right) \\
&\quad + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2 \\
&\stackrel{(16)}{\leq} \frac{p_{\text{mega}}}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2] \\
&\quad + \frac{1 - p_{\text{mega}}}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,2}^{t+1} - E_k [k_{i,2}^{t+1}]\|^2] \\
&\quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2.
\end{aligned} \tag{36}$$

889 Let us consider  $E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2]$ .

$$\begin{aligned}
&E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2] \\
&= E_k \left[ \left\| \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) - \frac{b}{p_{\text{mega}}} \left( h_i^t - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) \right) \right\|^2 \right]
\end{aligned}$$



$$\begin{aligned}
& - \left( \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t)) \right) \Big\|^2 \Big] \\
& = \mathbb{E}_k \left[ \left\| \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) + \frac{b}{p_{\text{mega}}} \left( \frac{1}{B'} \sum_{k=1}^{B'} \nabla f_i(x^t; \xi_{ik}^{t+1}) \right) \right. \right. \\
& \quad \left. \left. - \left( \nabla f_i(x^{t+1}) - \nabla f_i(x^t) + \frac{b}{p_{\text{mega}}} (\nabla f_i(x^t)) \right) \right\|^2 \right] \\
& = \frac{1}{B'^2} \sum_{k=1}^{B'} \mathbb{E}_k \left[ \left\| \frac{b}{p_{\text{mega}}} (\nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \nabla f_i(x^{t+1})) \right. \right. \\
& \quad \left. \left. + \left( 1 - \frac{b}{p_{\text{mega}}} \right) (\nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \nabla f_i(x^t; \xi_{ik}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))) \right\|^2 \right],
\end{aligned}$$

890 where we used independence of the mini-batch samples. Using (16), we get

$$\begin{aligned}
& \mathbb{E}_k \left[ \left\| k_{i,1}^{t+1} - \mathbb{E}_k [k_{i,1}^{t+1}] \right\|^2 \right] \\
& \leq \frac{2b^2}{B'^2 p_{\text{mega}}^2} \sum_{k=1}^{B'} \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \nabla f_i(x^{t+1}) \right\|^2 \right] \\
& \quad + \frac{2}{B'^2} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 \sum_{k=1}^{B'} \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_{ik}^{t+1}) - \nabla f_i(x^t; \xi_{ik}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right].
\end{aligned}$$

891 Due to Assumptions 5 and 6, we have

$$\mathbb{E}_k \left[ \left\| k_{i,1}^{t+1} - \mathbb{E}_k [k_{i,1}^{t+1}] \right\|^2 \right] \leq \frac{2b^2 \sigma^2}{B' p_{\text{mega}}^2} + \frac{2L_\sigma^2}{B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 \|x^{t+1} - x^t\|^2. \quad (37)$$

892 Next, we estimate the bound for  $\mathbb{E}_k \left[ \left\| k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}] \right\|^2 \right]$ .

$$\begin{aligned}
& \mathbb{E}_k \left[ \left\| k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}] \right\|^2 \right] \\
& = \mathbb{E}_k \left[ \left\| \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \frac{1}{B} \sum_{j=1}^B \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
& = \frac{1}{B^2} \sum_{j=1}^B \mathbb{E}_k \left[ \left\| \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right].
\end{aligned}$$

893 Due to Assumptions 6, we have

$$\mathbb{E}_k \left[ \left\| k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}] \right\|^2 \right] \leq \frac{L_\sigma^2}{B} \|x^{t+1} - x^t\|^2. \quad (38)$$

894 Plugging (37) and (38) into (36), we obtain

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\
& \leq \frac{p_{\text{mega}}}{np_a} \left( \frac{2b^2 \sigma^2}{B' p_{\text{mega}}^2} + \frac{2L_\sigma^2}{B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 \|x^{t+1} - x^t\|^2 \right) \\
& \quad + \frac{(1 - p_{\text{mega}}) L_\sigma^2}{np_a B} \|x^{t+1} - x^t\|^2 \\
& \quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2
\end{aligned}$$

$$+ \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2.$$

895 Using Assumption 3, we get

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\ & \leq \frac{2b^2 \sigma^2}{n p_{\text{mega}} p_a B'} + \left( \frac{2p_{\text{mega}} L_\sigma^2}{n p_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1 - p_{\text{mega}}) L_\sigma^2}{n p_a B} + \frac{2(p_a - p_{aa}) \widehat{L}^2}{n p_a^2} \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

896 Using almost the same derivations, we can prove the second inequality:

$$\begin{aligned} & \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\ & = p_{\text{mega}} \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_{i,1}^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] + (1 - p_{\text{mega}}) \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_{i,2}^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\ & \stackrel{(17)}{=} p_{\text{mega}} \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_{i,1}^{t+1} - \mathbb{E}_k [h_{i,1}^{t+1}]\|^2 \right] \right] + (1 - p_{\text{mega}}) \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \|h_{i,2}^{t+1} - \mathbb{E}_k [h_{i,2}^{t+1}]\|^2 \right] \right] \\ & \quad + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\ & = p_{\text{mega}} p_a \mathbb{E}_k \left[ \left\| h_i^t + \frac{1}{p_a} k_{i,1}^{t+1} - (h_i^t + \mathbb{E}_k [k_{i,1}^{t+1}]) \right\|^2 \right] \\ & \quad + p_{\text{mega}} (1 - p_a) \|h_i^t - (h_i^t + \mathbb{E}_k [k_{i,1}^{t+1}])\|^2 \\ & \quad + (1 - p_{\text{mega}}) p_a \mathbb{E}_k \left[ \left\| h_i^t + \frac{1}{p_a} k_{i,2}^{t+1} - (h_i^t + \mathbb{E}_k [k_{i,2}^{t+1}]) \right\|^2 \right] \\ & \quad + (1 - p_{\text{mega}}) (1 - p_a) \|h_i^t - (h_i^t + \mathbb{E}_k [k_{i,2}^{t+1}])\|^2 \\ & \quad + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\ & = p_{\text{mega}} p_a \mathbb{E}_k \left[ \left\| \frac{1}{p_a} k_{i,1}^{t+1} - \mathbb{E}_k [k_{i,1}^{t+1}] \right\|^2 \right] \\ & \quad + p_{\text{mega}} (1 - p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \quad + (1 - p_{\text{mega}}) p_a \mathbb{E}_k \left[ \left\| \frac{1}{p_a} k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}] \right\|^2 \right] \\ & \quad + (1 - p_{\text{mega}}) (1 - p_a) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\ & \quad + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\ & \stackrel{(17)}{=} \frac{p_{\text{mega}}}{p_a} \mathbb{E}_k \left[ \|k_{i,1}^{t+1} - \mathbb{E}_k [k_{i,1}^{t+1}]\|^2 \right] \\ & \quad + \frac{(1 - p_{\text{mega}})}{p_a} \mathbb{E}_k \left[ \|k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}]\|^2 \right] \\ & \quad + \frac{p_{\text{mega}} (1 - p_a)}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \quad + \frac{(1 - p_{\text{mega}}) (1 - p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \end{aligned}$$

$$\begin{aligned}
& + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
& \stackrel{(16)}{\leq} \frac{p_{\text{mega}}}{p_a} \mathbb{E}_k \left[ \|k_{i,1}^{t+1} - \mathbb{E}_k[k_{i,1}^{t+1}]\|^2 \right] \\
& + \frac{(1 - p_{\text{mega}})}{p_a} \mathbb{E}_k \left[ \|k_{i,2}^{t+1} - \mathbb{E}_k[k_{i,2}^{t+1}]\|^2 \right] \\
& + \frac{2(1 - p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& + \frac{2(1 - p_a)b^2}{p_{\text{mega}}p_a} \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

897 Using (37) and (38), we get

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\
& \leq \frac{2b^2\sigma^2}{p_a p_{\text{mega}} B'} + \frac{2p_{\text{mega}} L_\sigma^2}{p_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 \|x^{t+1} - x^t\|^2 \\
& + \frac{(1 - p_{\text{mega}}) L_\sigma^2}{p_a B} \|x^{t+1} - x^t\|^2 \\
& + \frac{2(1 - p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& + \frac{2(1 - p_a)b^2}{p_{\text{mega}}p_a} \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

898 Next, due to Assumption 3, we obtain

$$\begin{aligned}
& \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \\
& \leq \frac{2b^2\sigma^2}{p_a p_{\text{mega}} B'} + \left( \frac{2p_{\text{mega}} L_\sigma^2}{p_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1 - p_{\text{mega}}) L_\sigma^2}{p_a B} + \frac{2(1 - p_a) L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
& + \frac{2(1 - p_a)b^2}{p_{\text{mega}}p_a} \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

899 The third inequality can be proved with the help of (38) and Assumption 3.

$$\begin{aligned}
& \mathbb{E}_k \left[ \|k_{i,2}^{t+1}\|^2 \right] \\
& \stackrel{(17)}{=} \mathbb{E}_k \left[ \|k_{i,2}^{t+1} - \mathbb{E}_k[k_{i,2}^{t+1}]\|^2 \right] + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& \leq \frac{L_\sigma^2}{B} \|x^{t+1} - x^t\|^2 + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& \leq \left( \frac{L_\sigma^2}{B} + L_i^2 \right) \|x^{t+1} - x^t\|^2.
\end{aligned}$$

900

□

**Theorem 11.** Suppose that Assumptions 1, 2, 3, 5, 6, 7 and 8 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_{\text{mega}}p_a}{2-p_a}$ , probability  $p_{\text{mega}} \in (0, 1]$ , batch size  $B' \geq B \geq 1$

$$\gamma \leq \left( L + \sqrt{\frac{8(2\omega+1)\omega}{np_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right) + \frac{16}{np_{\text{mega}}p_a^2} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{L_\sigma^2}{B} \right)} \right)^{-1},$$

901 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 8. Then

$$\mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] \leq \frac{1}{T} \left[ \frac{2\Delta_0}{\gamma} + \frac{4}{p_{\text{mega}}p_a} \|h^0 - \nabla f(x^0)\|^2 + \frac{4 \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_{\text{mega}}p_a} \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right]$$

$$+ \frac{12\sigma^2}{nB'}.$$

902 *Proof.* Due to Lemma 2 and the update step from Line 4 in Algorithm 8, we have

$$\begin{aligned} & \mathbb{E}_{t+1} [f(x^{t+1})] \\ & \leq \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2 \right] \\ & = \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - h^t + h^t - \nabla f(x^t)\|^2 \right] \\ & \stackrel{(17)}{\leq} \mathbb{E}_{t+1} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma (\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2) \right]. \end{aligned}$$

903 Let us fix constants  $\kappa, \eta, \nu, \rho \in [0, \infty)$  that we will define later. Considering Lemma 13, Lemma 14,  
904 and the law of total expectation, we obtain

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & \quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma (\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2) \right] \\ & \quad + \kappa \mathbb{E} \left[ \mathbb{E}_k \left[ \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} [\|g^{t+1} - h^{t+1}\|^2] \right] \right] \right] \right] \\ & \quad + \eta \mathbb{E} \left[ \mathbb{E}_k \left[ \mathbb{E}_C \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \right] \right] \right] \\ & \quad + \nu \mathbb{E} \left[ \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] \right] \right] \right] \\ & \quad + \rho \mathbb{E} \left[ \mathbb{E}_k \left[ \mathbb{E}_{p_a} \left[ \mathbb{E}_{p_{\text{mega}}} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \right] \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma (\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2) \right] \\ & \quad + \kappa \mathbb{E} \left( \frac{2(1-p_{\text{mega}})\omega}{np_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \left( \frac{(p_a - p_{aa})a^2}{n^2 p_a^2} + \frac{2(1-p_{\text{mega}})a^2\omega}{n^2 p_a} \right) \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \right) \\ & \quad + \eta \mathbb{E} \left( \frac{2(1-p_{\text{mega}})\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \left( \frac{(1-p_a)a^2}{p_a} + \frac{2(1-p_{\text{mega}})a^2\omega}{p_a} \right) \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \right) \\ & \quad + \nu \mathbb{E} \left( \frac{2b^2\sigma^2}{np_{\text{mega}}p_a B'} + \left( \frac{2p_{\text{mega}}L_\sigma^2}{np_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1-p_{\text{mega}})L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1-p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2 \right) \\ & \quad + \rho \mathbb{E} \left( \frac{2b^2\sigma^2}{p_a p_{\text{mega}} B'} + \left( \frac{2p_{\text{mega}}L_\sigma^2}{p_a B'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + \frac{(1-p_{\text{mega}})L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right) \end{aligned}$$

$$+ \frac{2(1-p_a)b^2}{np_{\text{mega}}p_a} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1-p_{\text{mega}}) \right) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \Bigg).$$

Let us simplify the last inequality. Since  $B' \geq B$  and  $b = \frac{p_{\text{mega}}p_a}{2-p_a} \leq p_{\text{mega}}$ , we have  $1 - p_{\text{mega}} \leq 1$ ,

$$\frac{2p_{\text{mega}}L_\sigma^2}{p_aB'} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 \leq \frac{2p_{\text{mega}}L_\sigma^2}{p_aB},$$

$$\left( p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1-p_{\text{mega}}) \right) \leq 1 - b,$$

and

$$\left( \frac{2(1-p_a)b^2}{p_{\text{mega}}p_a} + p_{\text{mega}} \left( 1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1-p_{\text{mega}}) \right) \leq 1 - b.$$

905 Thus

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} \left[ f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left( \frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma (\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2) \right] \\ & + \kappa \mathbb{E} \left( \frac{2\omega}{np_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \frac{((2\omega+1)p_a - p_{aa})a^2}{n^2p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \right) \\ & + \eta \mathbb{E} \left( \frac{2\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \frac{(2\omega+1-p_a)a^2}{p_a} \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \right) \\ & + \nu \mathbb{E} \left( \frac{2b^2\sigma^2}{np_{\text{mega}}p_aB'} + \left( \frac{2L_\sigma^2}{np_aB} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + \frac{2(p_a - p_{aa})b^2}{n^2p_a^2p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + (1-b) \|h^t - \nabla f(x^t)\|^2 \right) \\ & + \rho \mathbb{E} \left( \frac{2b^2\sigma^2}{p_ap_{\text{mega}}B'} + \left( \frac{2L_\sigma^2}{p_aB} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \right. \\ & \quad \left. + (1-b) \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right). \end{aligned}$$

906 After rearranging the terms, we get

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \end{aligned}$$

$$\begin{aligned}
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{2\kappa\omega}{np_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \frac{2\eta\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + (\gamma + \kappa(1-a)^2) \mathbb{E} [\|g^t - h^t\|^2] \\
&\quad + \left( \kappa \frac{((2\omega+1)p_a - p_{aa})a^2}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad + (\gamma + \nu(1-b)) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'}.
\end{aligned}$$

907 Let us take  $\kappa = \frac{\gamma}{a}$ , thus  $\gamma + \kappa(1-a)^2 \leq \kappa$  and

$$\begin{aligned}
&\mathbb{E} [f(x^{t+1})] + \frac{\gamma}{a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
&\quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{2\gamma\omega}{anp_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \frac{2\eta\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma}{a} \mathbb{E} [\|g^t - h^t\|^2] \\
&\quad + \left( \frac{\gamma((2\omega+1)p_a - p_{aa})a}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad + (\gamma + \nu(1-b)) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'}.
\end{aligned}$$

908 Next, since  $a = \frac{p_a}{2\omega+1}$ , we have  $\left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \leq 1-a$ . We the choice  $\eta =$

909  $\frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2}$ , we guarantee  $\frac{\gamma((2\omega+1)p_a - p_{aa})a}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \leq \eta$  and

$$\begin{aligned}
&\mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
&\quad + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{2\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \frac{2\gamma((2\omega+1)p_a - p_{aa})\omega}{np_a^3} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad + (\gamma + \nu(1-b)) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'} \\
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad + (\gamma + \nu(1-b)) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\
&\quad + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
&\quad + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'},
\end{aligned}$$

910 where simplified the term using  $p_{aa} \geq 0$ . Let us take  $\nu = \frac{\gamma}{b}$  to obtain

$$\begin{aligned}
&\mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
&\quad + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
&\leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
&\quad - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
&\quad \left. - \left( \frac{2\gamma L_\sigma^2}{bn p_a B} + \frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bn p_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
&\quad + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&\quad + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2]
\end{aligned}$$

$$\begin{aligned}
& + \left( \frac{2\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{2\gamma b}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'}.
\end{aligned}$$

911 Next, we take  $\rho = \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}}$ , thus

$$\begin{aligned}
& \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma(2\omega + 1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{2\gamma L_\sigma^2}{bn p_a B} + \frac{2\gamma(p_a - p_{aa})\widehat{L}^2}{bn p_a^2} \right) - \left( \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \right) \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{2\gamma b}{np_{\text{mega}} p_a} + \frac{4\gamma(p_a - p_{aa})b^2}{np_a^3 p_{\text{mega}}} \right) \frac{\sigma^2}{B'}.
\end{aligned}$$

912 Since  $\frac{p_{\text{mega}} p_a}{2} \leq b \leq p_{\text{mega}} p_a$  and  $1 - p_a \leq 1 - \frac{p_{aa}}{p_a} \leq 1$ , we get

$$\begin{aligned}
& \mathbb{E}[f(x^{t+1})] + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma(2\omega + 1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{4\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{4\gamma(p_a - p_{aa})\widehat{L}^2}{np_{\text{mega}} p_a^3} \right) - \left( \frac{4\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{4\gamma(1-p_a)\widehat{L}^2}{np_{\text{mega}} p_a^2} \right) \right) \mathbb{E}[\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma(2\omega + 1)}{p_a} \mathbb{E}[\|g^t - h^t\|^2] + \frac{\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E}[\|h^t - \nabla f(x^t)\|^2] + \frac{2\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{6\gamma\sigma^2}{nB'} \\
& \leq \mathbb{E}[f(x^t)] - \frac{\gamma}{2} \mathbb{E}[\|\nabla f(x^t)\|^2]
\end{aligned}$$



$$\begin{aligned}
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \left( \frac{8\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{8\gamma \left(1 - \frac{p_{\text{aa}}}{p_a}\right) \widehat{L}^2}{np_{\text{mega}} p_a^2} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{6\gamma\sigma^2}{nB'}.
\end{aligned}$$

913 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \frac{\gamma(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{\gamma((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \frac{\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \frac{2\gamma(p_a - p_{\text{aa}})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{6\gamma\sigma^2}{nB'}.
\end{aligned}$$

914 It is left to apply Lemma 3 with

$$\begin{aligned}
\Psi^t &= \frac{(2\omega+1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \frac{((2\omega+1)p_a - p_{\text{aa}})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
&+ \frac{1}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \frac{2 \left(1 - \frac{p_{\text{aa}}}{p_a}\right)}{np_a p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]
\end{aligned}$$

915 and  $C = \frac{6\sigma^2}{nB'}$  to conclude the proof.  $\square$

**Corollary 6.** Suppose that assumptions from Theorem 11 hold, probability  $p_{\text{mega}} = \min \left\{ \frac{\zeta_c}{d}, \frac{n\varepsilon B}{\sigma^2} \right\}$ , batch size  $B' = \Theta \left( \frac{\sigma^2}{n\varepsilon} \right)$ , and  $h_i^0 = g_i^0 = \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0)$  for all  $i \in [n]$ , initial batch size  $B_{\text{init}} = \Theta \left( \frac{B}{p_{\text{mega}} \sqrt{p_a}} \right) = \Theta \left( \max \left\{ \frac{Bd}{\sqrt{p_a} \zeta_c}, \frac{\sigma^2}{\sqrt{p_a} n \varepsilon} \right\} \right)$ , then DASHA-PP-SYNC-MVR needs

$$T := \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \left( \frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_c n}} \right) \left( \widehat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon} n} \left( \frac{\widehat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right).$$

916 communication rounds to get an  $\varepsilon$ -solution, the expected communication complexity is equal to  
917  $\mathcal{O}(d + \zeta_c T)$ , and the expected number of stochastic gradient calculations per node equals  $\mathcal{O}(B_{\text{init}} +$   
918  $BT)$ , where  $\zeta_c$  is the expected density from Definition 12.

919 *Proof.* Due to the choice of  $B'$ , we have

$$\begin{aligned} \mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] &\leq \frac{1}{T} \left[ 2\Delta_0 \left( L + \sqrt{\frac{8(2\omega+1)\omega}{np_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right) + \frac{16}{np_{\text{mega}}p_a^2} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{L_\sigma^2}{B} \right)} \right) \right. \\ &\quad \left. + \frac{4}{p_{\text{mega}}p_a} \|h^0 - \nabla f(x^0)\|^2 + \frac{4 \left( 1 - \frac{p_{aa}}{p_a} \right)}{np_{\text{mega}}p_a} \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right] \\ &\quad + \frac{2\varepsilon}{3}. \end{aligned}$$

920 Using

$$\mathbb{E} \left[ \|h^0 - \nabla f(x^0)\|^2 \right] = \mathbb{E} \left[ \left\| \frac{1}{n} \sum_{i=1}^n \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0) - \nabla f(x^0) \right\|^2 \right] \leq \frac{\sigma^2}{nB_{\text{init}}}$$

921 and

$$\frac{1}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \|h_i^0 - \nabla f_i(x^0)\|^2 \right] = \frac{1}{n^2} \sum_{i=1}^n \mathbb{E} \left[ \left\| \frac{1}{B_{\text{init}}} \sum_{k=1}^{B_{\text{init}}} \nabla f_i(x^0; \xi_{ik}^0) - \nabla f_i(x^0) \right\|^2 \right] \leq \frac{\sigma^2}{nB_{\text{init}}},$$

922 we have

$$\begin{aligned} \mathbb{E} \left[ \|\nabla f(\hat{x}^T)\|^2 \right] &\leq \frac{1}{T} \left[ 2\Delta_0 \left( L + \sqrt{\frac{8(2\omega+1)\omega}{np_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right) + \frac{16}{np_{\text{mega}}p_a^2} \left( \left( 1 - \frac{p_{aa}}{p_a} \right) \hat{L}^2 + \frac{L_\sigma^2}{B} \right)} \right) \right. \\ &\quad \left. + \frac{8\sigma^2}{np_{\text{mega}}p_a B_{\text{init}}} \right] \\ &\quad + \frac{2\varepsilon}{3}. \end{aligned}$$

923 Therefore, we can take the following  $T$  to get  $\varepsilon$ -solution.

$$T = \mathcal{O} \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \sqrt{\frac{\omega^2}{np_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right) + \frac{1}{np_{\text{mega}}p_a^2} \left( \hat{L}^2 + \frac{L_\sigma^2}{B} \right)} \right) + \frac{\sigma^2}{np_{\text{mega}}p_a B_{\text{init}}} \right] \right)$$

924 Considering the choice of  $p_{\text{mega}}$  and  $B_{\text{init}}$ , we obtain

$$\begin{aligned} T &= \mathcal{O} \left( \frac{1}{\varepsilon} \left[ \Delta_0 \left( L + \left( \frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_C n}} \right) \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left( \frac{\hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right) + \frac{\sigma^2}{np_{\text{mega}}p_a B_{\text{init}}} \right] \right) \\ &= \mathcal{O} \left( \frac{\Delta_0}{\varepsilon} \left[ L + \left( \frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_C n}} \right) \left( \hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left( \frac{\hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right). \end{aligned}$$

925 The expected communication complexity equals  $\mathcal{O}(d + p_{\text{mega}}d + (1 - p_{\text{mega}})\zeta_C) =$   
 926  $\mathcal{O}(d + \zeta_C)$  and the expected number of stochastic gradient calculations per node equals  
 927  $\mathcal{O}(B_{\text{init}} + p_{\text{mega}}B' + (1 - p_{\text{mega}})B) = \mathcal{O}(B_{\text{init}} + B)$ .  $\square$

**Theorem 13.** Suppose that Assumptions 1, 2, 3, 5, 6, 7, 8 and 9 hold. Let us take  $a = \frac{p_a}{2\omega+1}$ ,  $b = \frac{p_{\text{mega}}p_a}{2-p_a}$ , probability  $p_{\text{mega}} \in (0, 1]$ , batch size  $B' \geq B \geq 1$ ,

$$\gamma \leq \min \left\{ \left( L + \sqrt{\frac{16(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) + \left( \frac{48L_\sigma^2}{np_{\text{mega}}p_a^2B} + \frac{24 \left( 1 - \frac{p_{aa}}{p_a} \right) \widehat{L}^2}{np_{\text{mega}}p_a^2} \right)} \right)^{-1}, \frac{a}{2\mu}, \frac{b}{2\mu} \right\},$$

928 and  $h_i^0 = g_i^0$  for all  $i \in [n]$  in Algorithm 8. Then

$$\begin{aligned} & \mathbb{E} [f(x^T) - f^*] \\ & \leq (1 - \gamma\mu)^T \left( \Delta_0 + \frac{2\gamma}{b} \|h^0 - \nabla f(x^0)\|^2 + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \frac{1}{n} \sum_{i=1}^n \|h_i^0 - \nabla f_i(x^0)\|^2 \right) + \frac{20\sigma^2}{\mu n B'}. \end{aligned}$$

929 *Proof.* Let us fix constants  $\kappa, \eta, \nu, \rho \in [0, \infty)$  that we will define later. As in the proof of Theorem 11,  
930 we can get

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \kappa \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\ & - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{2\kappa\omega}{np_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \frac{2\eta\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\ & \quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\ & + (\gamma + \kappa(1-a)^2) \mathbb{E} [\|g^t - h^t\|^2] \\ & + \left( \kappa \frac{((2\omega+1)p_a - p_{aa})a^2}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ & + (\gamma + \nu(1-b)) \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] \\ & + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\ & + \left( \frac{2\nu b^2}{np_{\text{mega}}p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'}. \end{aligned}$$

931 Let us take  $\kappa = \frac{2\gamma}{a}$ , thus  $\gamma + \kappa(1-a)^2 \leq (1 - \frac{a}{2})\kappa$  and

$$\begin{aligned} & \mathbb{E} [f(x^{t+1})] + \frac{2\gamma}{a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \eta \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\ & + \nu \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\ & \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\ & - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{4\gamma\omega}{anp_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \frac{2\eta\omega}{p_a} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \end{aligned}$$

$$\begin{aligned}
& -\nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma}{a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] \\
& + \left( \frac{2\gamma((2\omega+1)p_a - p_{aa})a}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + (\gamma + \nu(1-b)) \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'}.
\end{aligned}$$

932 Next, since  $a = \frac{p_a}{2\omega+1}$ , we have  $\left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \leq 1-a$ . We the choice  $\eta =$   
933  $\frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2}$ , we guarantee  $\frac{\gamma((2\omega+1)p_a - p_{aa})a}{np_a^2} + \eta \left( \frac{(2\omega+1-p_a)a^2}{p_a} + (1-a)^2 \right) \leq (1-\frac{a}{2})\eta$  and

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \nu \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \nu \left( \frac{2L_\sigma^2}{np_a B} + \frac{2(p_a - p_{aa})\widehat{L}^2}{np_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + (\gamma + \nu(1-b)) \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( \nu \frac{2(p_a - p_{aa})b^2}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{2\nu b^2}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'},
\end{aligned}$$

934 where simplified the term using  $p_{aa} \geq 0$ . Let us take  $\nu = \frac{2\gamma}{b}$  to obtain

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \rho \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{4\gamma L_\sigma^2}{bnp_a B} + \frac{4\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right) - \rho \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \left( 1 - \frac{a}{2} \right) \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] \\
& + \left( \frac{4\gamma(p_a - p_{aa})b}{np_a^2 p_{\text{mega}}} + \rho(1-b) \right) \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{4\gamma b}{np_{\text{mega}} p_a} + \frac{2\rho b^2}{p_a p_{\text{mega}}} \right) \frac{\sigma^2}{B'},
\end{aligned}$$

935 Next, we take  $\rho = \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}}$ , thus

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{4\gamma L_\sigma^2}{bnp_a B} + \frac{4\gamma(p_a - p_{aa})\widehat{L}^2}{bnp_a^2} \right) - \left( \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \right) \left( \frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right] \\
& + \left( 1 - \frac{a}{2} \right) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \left( 1 - \frac{a}{2} \right) \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left( 1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] + \left( 1 - \frac{b}{2} \right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left( \frac{4\gamma b}{np_{\text{mega}} p_a} + \frac{16\gamma(p_a - p_{aa})b^2}{np_a^3 p_{\text{mega}}^2} \right) \frac{\sigma^2}{B'},
\end{aligned}$$

936 Since  $\frac{p_{\text{mega}} p_a}{2} \leq b \leq p_{\text{mega}} p_a$  and  $1 - p_a \leq 1 - \frac{p_{aa}}{p_a} \leq 1$ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[ \|g^{t+1} - h^{t+1}\|^2 \right] + \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} \left[ \|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[ \|\nabla f(x^t)\|^2 \right] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega+1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left( \frac{8\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{8\gamma(p_a - p_{aa})\widehat{L}^2}{np_{\text{mega}} p_a^3} \right) - \left( \frac{16\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{16\gamma(1-p_a)\widehat{L}^2}{np_{\text{mega}} p_a^2} \right) \right) \mathbb{E} \left[ \|x^{t+1} - x^t\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'} \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& - \left( \frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega + 1)\omega}{np_a^2} \left( \frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \left( \frac{24\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{24\gamma \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2}{np_{\text{mega}} p_a^2} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'}.
\end{aligned}$$

937 Using Lemma 4 and the assumption about  $\gamma$ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'}.
\end{aligned}$$

938 Due to  $\gamma \leq \frac{a}{2\mu}$  and  $\gamma \leq \frac{b}{2\mu}$ , we have

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + (1 - \gamma\mu) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + (1 - \gamma\mu) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + (1 - \gamma\mu) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + (1 - \gamma\mu) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'}.
\end{aligned}$$

939 It is left to apply Lemma 11 with

$$\begin{aligned}\Psi^t &= \frac{2(2\omega + 1)}{p_a} \mathbb{E} \left[ \|g^t - h^t\|^2 \right] + \frac{2((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\ &+ \frac{2}{b} \mathbb{E} \left[ \|h^t - \nabla f(x^t)\|^2 \right] + \frac{8(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[ \frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right]\end{aligned}$$

940 and  $C = \frac{20\sigma^2}{nB'}$  to conclude the proof. □