
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^{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.¹ Since we work in the nonconvex world, our goal is to find an ε -solution (ε -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^{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 $\hat{L}^2 := \frac{1}{n} \sum_{i=1}^n L_i^2$.²*

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^{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 ξ_{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 ε -solution, and the *communication complexity*, i.e., the number of bits that each node has to send to the server to get ε -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 $\omega \in \mathbb{R}$

$$\text{such that } \mathbb{E}[\mathcal{C}(x)] = x \quad \text{and} \quad \mathbb{E}[\|\mathcal{C}(x) - x\|^2] \leq \omega \|x\|^2 \quad \forall x \in \mathbb{R}^d. \quad (4)$$

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.

¹Note 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.

²Note 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 $\mathcal{O}\left(\frac{\omega}{\sqrt{n\varepsilon}}\right)$?

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

^(a) **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.

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

^(c) 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.

^(d) 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.

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

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*

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

and these events from different communication rounds are independent.

We are not fighting for the full generality and believe that more complex sampling strategies can be considered in the analysis. For simplicity, we settle upon Assumption 8. Standard partial participation strategies, including s -nice sampling, where the server chooses uniformly s nodes without replacement ($p_a = s/n$ and $p_{aa} = s(s-1)/(n(n-1))$), and independent participation, where each

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)	✓	✗ ^(a)	✓ ^(b)	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)	✗	✗ ^(a)	✗	Suboptimal oracle complexity $\mathcal{O}\left(\frac{m}{\varepsilon}\right)$.
DASHA (Tyurin and Richtárik, 2023)	✓	✗	✓	—
DASHA-PP (new)	✓	✓	✓	—

^(a), ^(b) : see Table 1.

node independently participates with probability p_a (due to independence, we have $p_{aa} = p_a^2$), satisfy Assumption 8. In the literature, s -nice sampling is one of the most popular strategies (Zhao et al., 2021a; Richtárik et al., 2021; Reddi et al., 2020; Konečný et al., 2016).

3 Motivation and Related Work

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

1. Variance reduction of stochastic gradients

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

2. Compressed communication

In distributed optimization (Ramesh et al., 2021; Xu et al., 2021), lossy communication compression 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 ε -solution. In the finite-sum and stochastic settings, the current

113 SOTA communication complexity is attained by the **DASHA** method, while maintaining the optimal
 114 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.

115 3. Partial participation

116 From the beginning of federated learning era, the partial participation has been considered to be
 117 the essential feature of distributed optimization methods (McMahan et al., 2017; Konečný et al.,
 118 2016; Kairouz et al., 2021). However, previously proposed methods have limitations: i) methods
 119 **MARINA** and **FedPAGE** from (Gorbunov et al., 2021; Zhao et al., 2021b) still require synchronization
 120 of all nodes with a small probability. ii) in the stochastic settings, methods **FedAvg**, **SCAFFOLD**, and
 121 **FRECON** with the partial participation mechanism (McMahan et al., 2017; Karimireddy et al., 2020b;
 122 Zhao et al., 2021a) provide results without variance reduction techniques from (Fang et al., 2018; Li
 123 et al., 2021a; Cutkosky and Orabona, 2019) and, therefore, get suboptimal oracle complexities. Note
 124 that **FRECON** and **DASHA** reduce the variance *only from compressors* (in the partial participation and
 125 stochastic setting). iii) in the finite-sum setting, the **ZeroSARAH** method by Li et al. (2021b) focuses
 126 on the homogeneous regime only (the functions f_i are equal). iv) The **MIME** method by Karimireddy
 127 et al. (2020a) and the **CE-LSGD** method (for Partial Participation) by the concurrent paper (Patel
 128 et al., 2022) consider the online version of the problem (1). Therefore, **MIME** and **CE-LSGD** (for
 129 Partial Participation) require stricter assumptions, including the bounded inter-client gradient variance
 130 assumption. In the finite-sum setting (2), **MIME** and **CE-LSGD** obtain a suboptimal oracle complexity
 131 $\mathcal{O}(1/\varepsilon^{3/2})$ while, in the full participation setting, it is possible to get the complexity $\mathcal{O}(1/\varepsilon)$.

132 4 Contributions

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

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

145 5 Algorithm Description and Main Challenges Towards Partial Participation

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

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

$$g_i^{t+1} = g_i^t + C_i (\nabla f_i(x^{t+1}) - (1-a)\nabla f_i(x^t) - ag_i^t). \quad (6)$$

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

$$C_i^p := \begin{cases} \frac{1}{p}C_i, & \text{w.p. } p_a, \\ 0, & \text{w.p. } 1 - p_a. \end{cases} \quad \stackrel{(6)}{\Rightarrow} g_i^{t+1} = \begin{cases} g_i^t + \frac{1}{p_a}C_i (\nabla f_i(x^{t+1}) - (1-a)\nabla f_i(x^t) - ag_i^t), & \text{w.p. } p_a \\ g_i^t, & \text{w.p. } 1 - p_a. \end{cases}$$

156 With probability $1 - p_a$, a node does not update g_i^t and does not send anything to the server. The
 157 main 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 .

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*^(a), 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*^(a) nodes
 - 7: **for** $i = 1, \dots, n$ in parallel **do**
 - 8: **if** i^{th} node is *participating*^(a) **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: $k_i^{t+1} = \begin{cases} \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)), \\ \text{with probability } p_{\text{page}} \text{ on all } \textit{participating} \text{ nodes,} \\ \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)), \\ \text{with probability } 1 - p_{\text{page}} \text{ on all } \textit{participating} \text{ nodes} \end{cases}$
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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 However, we focus our attention on partial participation *in the finite-sum and stochastic settings*.
 159 Consider the nodes' steps in **DASHA-MVR** (Tyurin and Richtárik, 2023) (see Algorithm 7) that is
 160 designed for the stochastic setting:

$$h_i^{t+1} = \nabla f_i(x^{t+1}; \xi_i^{t+1}) + (1 - b)(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})), \quad (7)$$

$$g_i^{t+1} = g_i^t + C_i (h_i^{t+1} - h_i^t - a(g_i^t - h_i^t)). \quad (8)$$

161 Even if we use the same trick for (8), we still have to update (7) in every iteration of the algorithm
 162 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^t and
 163 does not send anything to the server, it still has to update h_i^t , what is impossible without the points
 164 x^{t+1} and x^t . One of the main challenges was to “guess” how to generalize (7) and (8) to the partial
 165 participation setting. We now provide a solution (**DASHA-PP-MVR** with the batch size $B = 1$):

$$\begin{aligned} h_i^{t+1} &= h_i^t + \frac{1}{p_a} k_i^{t+1}, \quad k_i^{t+1} = \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})), \\ g_i^{t+1} &= g_i^t + C_i \left(\frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \text{ with probability } p_a, \\ \text{and } h_i^{t+1} &= h_i^t, \quad g_i^{t+1} = g_i^t \text{ with probability } 1 - p_a. \end{aligned} \quad (9)$$

166 Now both control variables g_i^t and h_i^t do not change with the probability $1 - p_a$. When the i^{th} node
 167 participates, the update rule of g_i^{t+1} and h_i^{t+1} was modified to make the proof work. When $p_a = 1$
 168 (no partial participation), the update rules from (9) reduce to (7) and (8).

169 The theoretical analysis of the new algorithm became more complicated: unlike (7) and (8), the
 170 control variables h_i^{t+1} and g_i^{t+1} in (9) (see also main Algorithm 1) are coupled by the randomness
 171 from the partial participation. Going deeper into details, for instance, one can compare Lemma I.2
 172 from (Tyurin and Richtárik, 2023) and Lemma 5 that both bound $\|g_i^{t+1} - h_i^{t+1}\|^2$. The former
 173 lemma does not use the knowledge about the update rule of h_i^{t+1} , works with one expectation $\mathbb{E}_{\mathcal{C}}[\cdot]$,
 174 uses only (4), (13), and (14). The latter lemma additionally requires and uses the structure of the
 175 update rule of h_i^{t+1} , surgically copes with the expectations $\mathbb{E}_{\mathcal{C}}[\cdot]$ and $\mathbb{E}_{p_a}[\cdot]$ (for instance, it is not
 176 trivial in each order one should apply the expectations), and uses the sampling lemma (Lemma 1).
 177 The same reasoning applies to other parts of the analysis and the finite-sum setting: the generalization
 178 of the previous algorithm and the additional randomness from the partial participation required us to
 179 rethink the previous proofs.

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

185 6 Theorems

186 We now present the convergence rates theorems of **DASHA-PP** in different settings. We will compare
 187 the theorems with the results of the current state-of-the-art methods, **MARINA** and **DASHA**, that work
 188 in the full participation setting. Suppose that **MARINA** or **DASHA** converges to ε -solution after T
 189 communication rounds. Then, ideally, we would expect the convergence of the new algorithms to
 190 ε -solution after up to T/p_a communication rounds due to the partial participation constraints³. The
 191 detailed analysis of the algorithms under Polyak-Łojasiewicz condition we provide in Section F. Let
 192 us define $\Delta_0 := f(x^0) - f^*$.

193 6.1 Gradient Setting

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

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

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

³We check this numerically in Section A.

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

6.2 Finite-Sum Setting

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

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

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

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

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

$$T := \mathcal{O}\left(\frac{\Delta_0}{\varepsilon} \left[L + \frac{\omega}{p_a \sqrt{n}} \left(\hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{1}{p_a} \sqrt{\frac{m}{n}} \left(\frac{\mathbb{1}_{p_a} \hat{L}}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right) \quad (10)$$

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

The convergence rate the rate of the current state-of-the-art method **DASHA-PAGE** without partial 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 (10). As expected, we see that the second term w.r.t. ω degenerates up to $1/p_a$. Surprisingly, the third term 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 $1/p_a$, one should take the batch size $B = \mathcal{O}(L_{\max}^2/\hat{L}^2)$. This interesting effect we analyze separately in Section C. The fact that the degeneration is up to $1/p_a$ we check numerically in Section A.

In the following corollary, we consider **RandK** compressors⁴ (see Definition 5) and show that with the particular choice of parameters, up to the Lipschitz constants factors, **DASHA-PP-PAGE** gets the optimal oracle complexity and SOTA communication complexity. Indeed, comparing the following result with (Tyurin and Richtárik, 2023, Corollary 6.6), one can see that we get the degeneration up to $1/p_a$ factor, which is expected in the partial participation setting. Note that the complexities improve with the number of workers n .

Corollary 2. Suppose that assumptions of Corollary 1 hold, $B \leq \min \left\{ \frac{1}{p_a} \sqrt{\frac{m}{n}}, \frac{L_{\max}^2}{\frac{1}{p_a} \hat{L}^2} \right\}$ ⁵, and we use the unbiased compressor **RandK** with $K = \Theta(Bd/\sqrt{m})$. Then the communication complexity of Algorithm 1 is $\mathcal{O}\left(d + \frac{L_{\max} \Delta_0 d}{p_a \varepsilon \sqrt{n}}\right)$, and the expected number of gradient calculations per node equals $\mathcal{O}\left(m + \frac{L_{\max} \Delta_0 \sqrt{m}}{p_a \varepsilon \sqrt{n}}\right)$.

The convergence rate of **DASHA-PP-FINITE-MVR** is provided in Section E.5.

⁴The choice of the compressor is driven by simplicity, and the following analysis can be used for other unbiased compressors.

⁵If $\mathbb{1}_{p_a} = 0$, then $\frac{L_{\max}^2}{\frac{1}{p_a} \hat{L}^2} = +\infty$

6.3 Stochastic Setting

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

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

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

In the next corollary, we choose momentum b and initialize vectors h_i^0 to get ε -solution.

Corollary 3. Suppose that assumptions from Theorem 4 hold, momentum $b = \Theta \left(\min \left\{ \frac{p_a}{\sqrt{\frac{n\varepsilon B}{\sigma^2}}}, \frac{p_a n \varepsilon B}{\sigma^2} \right\} \right)$, $\frac{\sigma^2}{n\varepsilon B} \geq 1$, 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 $i \in [n]$, and batch size $B_{\text{init}} = \Theta \left(\frac{\sqrt{p_a B}}{b} \right)$, then Algorithm 1 (DASHA-PP-MVR) needs

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

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

The convergence rate of the DASHA-SYNC-MVR, the state-of-the-art method without partial participation, 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 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 $B = \mathcal{O}(L_\sigma^2/\hat{L}^2)$, then the degeneration of the third term will be up to $1/p_a$. This effect we analyze in Section C. The fact that the degeneration is up to $1/p_a$ we check numerically in Section A.

In the following corollary, we consider RandK compressors (see Definition 5) and show that with the particular choice of parameters, up to the Lipschitz constants factors, DASHA-PP-MVR gets the optimal oracle complexity and SOTA communication complexity of DASHA-SYNC-MVR method. Indeed, comparing the following result with (Tyurin and Richtárik, 2023, Corollary 6.9), one can see that we get the degeneration up to $1/p_a$ factor, which is expected in the partial participation setting. Note that the complexities improve with the number of workers n .

Corollary 4. Suppose that assumptions of Corollary 3 hold, batch size $B \leq \min \left\{ \frac{\sigma}{p_a \sqrt{\varepsilon n}}, \frac{L_\sigma^2}{1 p_a \hat{L}^2} \right\}$, we take RandK compressors with $K = \Theta \left(\frac{B d \sqrt{\varepsilon n}}{\sigma} \right)$. Then the communication complexity equals $\mathcal{O} \left(\frac{d\sigma}{\sqrt{p_a} \sqrt{n\varepsilon}} + \frac{L_\sigma \Delta_0 d}{p_a \sqrt{n\varepsilon}} \right)$, and the expected number of stochastic gradient calculations per node equals $\mathcal{O} \left(\frac{\sigma^2}{\sqrt{p_a n \varepsilon}} + \frac{L_\sigma \Delta_0 \sigma}{p_a \varepsilon^{3/2} n} \right)$.

We are aware that the initial batch size B_{init} can be suboptimal w.r.t. ω 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).

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401 A Numerical Verification of Theoretical Dependencies

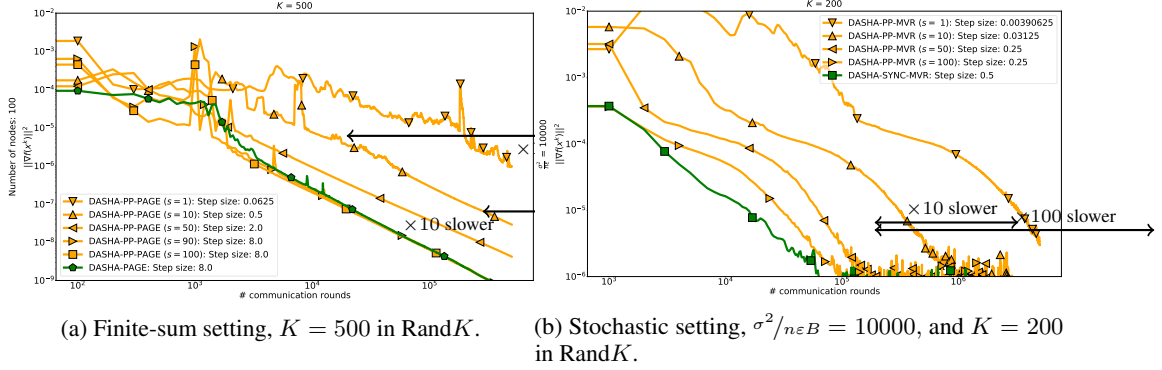


Figure 1: Classification task with the *real-sim* dataset.

402 Our main goal is to verify the dependeces from the theory. We compare **DASHA-PP** with **DASHA**.
 403 Clearly, **DASHA-PP** can not generally perform better than **DASHA**. In different settings, we verify
 404 that the bigger p_a , the closer **DASHA-PP** is to **DASHA**, i.e., **DASHA-PP** converges no slower than $1/p_a$
 405 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

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

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

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

409 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^{th} node,
 410 $y_{ij} \in \{1, 2\}$ is a corresponding label, m is the number of samples located on the i^{th} node, constant
 411 $\lambda = 0.001$ for all $i \in [n]$.

412 The code was written in Python 3.6.8 using PyTorch 1.9 (Paszke et al., 2019). A distributed
 413 environment was emulated on a machine with Intel(R) Xeon(R) Gold 6226R CPU @ 2.90GHz and
 414 64 cores.

415 We use the standard setting in experiments⁶ where all parameters except step sizes are taken as
 416 suggested in theory. Step sizes are finetuned from a set $\{2^i \mid i \in [-10, 10]\}$. We emulate the partial
 417 participation setting using s -nice sampling with the number of nodes $n = 100$. We consider the
 418 $\text{Rand}K$ compressor and take the batch size $B = 1$. We plot the relation between communication
 419 rounds and values of the norm of gradients at each communication round.

420 In the finite-sum (Figure 1a) and in the stochastic setting (Figure 1b), we see that the bigger probability
 421 $p_a = s/n$ to 1, the closer **DASHA-PP** to **DASHA**. Moreover, **DASHA-PP** with $s = 10$ and $s = 1$
 422 converges approximately $\times 10$ ($= 1/p_a$) and $\times 100$ ($= 1/p_a$) times slower, accordingly. Our theory
 423 predicts such behavior.

⁶Code: <https://github.com/mysteryresearcher/dasha-partial-participation>

424 B Original DASHA and DASHA-MVR Methods

425 To simplify the discussion and explanation from the main part, we present the algorithms from (Tyurin
426 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  $x^{t+1}$  and  $x^t$ 
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  $x^{t+1}$  and  $x^t$ 
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  $\{x^t\}_{k=0}^{T-1}$ 

```

427 C Problem of Estimating the Mean in the Partial Participation Setting

428 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

429 $B = \mathcal{O}\left(\min\left\{\frac{\sigma}{p_a\sqrt{\varepsilon n}}, \frac{L_{\sigma}^2}{p_a^2 \hat{\mathcal{L}}^2}\right\}\right)$ in **DASHA-PP-PAGE** and **DASHA-PP-MVR**, accordingly, guarantees

430 the degeneration up to $1/p_a$. This is surprising, because in methods with the variance reduction of

431 stochastic gradients (Li et al., 2021a; Tyurin and Richtárik, 2023) we can take the size of batch size

432 $B = \mathcal{O}(\sqrt{\frac{m}{n}})$ and $B = \mathcal{O}\left(\frac{\sigma}{\sqrt{\varepsilon n}}\right)$ and guarantee the optimality. Note that the smaller the batch size

433 B , the more the server and the nodes have to communicate to get ε -solution.

434 Let us consider the task of estimating the mean of vectors in the distributed setting. Suppose that we

435 have n nodes, and each of them contains m vectors $\{x_{ij}\}_{j=1}^m$, where $x_{ij} \in \mathbb{R}^d$ for all $i \in [n], j \in [m]$.

436 First, let us consider that each node samples a mini-batch I^i of size B with replacement and sends it

437 to the server. Then the server calculates the mean of the mini-batches from nodes. One can easily

438 show that the variance of the estimator is

$$\begin{aligned} & \mathbb{E} \left[\left\| \frac{1}{nB} \sum_{i=1}^n \sum_{j \in I^i} x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2 \right] \\ &= \frac{1}{nB} \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \left\| x_{ij} - \frac{1}{m} \sum_{j=1}^m x_{ij} \right\|^2. \end{aligned} \quad (11)$$

439 Next, we consider the same task in the partial participation setting with s -nice sampling, i.e., we

440 sample a random set $S \subset [n]$ of $s \in [n]$ nodes without replacement and receive the mini-batches

441 only from the sampled nodes. Such sampling of nodes satisfy Assumption 8 with $p_a = s/n$ and

442 $p_a = s(s-1)/n(n-1)$. In this case, the variance of the estimator (See Lemma 1 with $r_i = 0$ and

443 $s_i = \sum_{j \in I^i} x_{ij}$) is

$$\begin{aligned} & \mathbb{E} \left[\left\| \frac{1}{sB} \sum_{i \in S} \sum_{j \in I^i} x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2 \right] \\ &= \frac{1}{sB} \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m \underbrace{\left\| x_{ij} - \frac{1}{m} \sum_{j=1}^m x_{ij} \right\|^2}_{\mathcal{L}_{\max}^2} \\ &+ \frac{n-s}{s(n-1)} \frac{1}{n} \sum_{i=1}^n \underbrace{\left\| \frac{1}{m} \sum_{j=1}^m x_{ij} - \frac{1}{nm} \sum_{i=1}^n \sum_{j=1}^m x_{ij} \right\|^2}_{\hat{\mathcal{L}}^2}. \end{aligned} \quad (12)$$

444 Let us assume that $s \leq n/2$. Note that (11) scales with any $B \geq 1$, while (12) only scales when

445 $B = \mathcal{O}(\mathcal{L}_{\max}^2/\hat{\mathcal{L}}^2)$. In other words, for large enough B , the variance in (12) does not significantly

446 improves with the growth of B due to the term $\hat{\mathcal{L}}^2$. In our proof, due to partial participation, the

447 variance from (12) naturally appears, and we get the same effect. As was mentioned in Sections 6.2

448 and 6.3, it can be seen in our convergence rate bounds.

449 D Auxiliary facts

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

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

$$\|x + y\|^2 \leq 2\|x\|^2 + 2\|y\|^2 \quad (13)$$

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

$$\mathbb{E} \left[\|\xi\|^2 \right] = \mathbb{E} \left[\|\xi - \mathbb{E}[\xi]\|^2 \right] + \|\mathbb{E}[\xi]\|^2. \quad (14)$$

453 D.1 Sampling Lemma

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

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

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

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

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

460 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 $i \in [n]$,
461 nonrandom vector $r_i \in \mathbb{R}^d$ for all $i \in [n]$, and random vectors

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

462 then

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

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

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

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

466 Note, that

$$p_{an} = p_{aa} - p_a \quad (15)$$

467 and

$$p_{nn} = 1 - p_{aa} - 2p_{an}. \quad (16)$$

468 Using the law of total expectation and

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

469 we have

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

470 From the independence of random vectors s_i , we obtain

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

471 Using (15) and (16), we have

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

$$\begin{aligned}
& + \frac{p_{aa} - p_a^2}{n^2 p_a^2} \sum_{i \neq j}^n \langle E[s_i], E[s_j] \rangle \\
& \stackrel{(14)}{=} \frac{1}{n^2 p_a} \sum_{i=1}^n E \left[\|s_i - E[s_i]\|^2 \right] \\
& + \frac{1 - p_a}{n^2 p_a} \sum_{i=1}^n \|E[s_i]\|^2 \\
& + \frac{p_{aa} - p_a^2}{n^2 p_a^2} \sum_{i \neq j}^n \langle E[s_i], E[s_j] \rangle \\
& = \frac{1}{n^2 p_a} \sum_{i=1}^n E \left[\|s_i - E[s_i]\|^2 \right] \\
& + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|E[s_i]\|^2 \\
& + \frac{p_{aa} - p_a^2}{p_a^2} \left\| \frac{1}{n} \sum_{i=1}^n E[s_i] \right\|.
\end{aligned}$$

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

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

473

□

474 D.2 Compressors Facts

475 We define the *RandK* compressor that chooses without replacement K coordinates, scales them by a
476 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

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

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

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

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

480 E Proofs of Theorems

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

487 In the case of **DASHA-PP-PAGE**, there are two different sources of randomness from $\{k_i^{t+1}\}_{i=1}^n$.
488 We define $E_{p_{\text{page}}}[\cdot]$ and $E_B[\cdot]$ to be conditional expectations w.r.t. the probabilistic switching and
489 mini-batch indices I_i^t , accordingly, conditioned on all previous randomness. Note, that $E_{t+1}[\cdot] =$
490 $E_B[E_C[E_{p_a}[E_{p_{\text{page}}}[\cdot]]]]$ and $E_{t+1}[\cdot] = E_B[E_{p_{\text{page}}}[E_C[E_{p_a}[\cdot]]]]$.

491 **E.1 Standard Lemmas in the Nonconvex Setting**

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

493 **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
 494 $\gamma > 0$, we have

$$f(x^{t+1}) \leq f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2. \quad (17)$$

495 *Proof.* Using L -smoothness, we have

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

496 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

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

497 □

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

$$\mathbb{E} [f(x^{t+1})] + \gamma \Psi^{t+1} \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] + \gamma \Psi^t + \gamma C, \quad (18)$$

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

$$\mathbb{E} [\|\nabla f(\hat{x}^T)\|^2] \leq \frac{2\Delta_0}{\gamma T} + \frac{2\Psi^0}{T} + 2C, \quad (19)$$

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

502 *Proof.* By unrolling (18) for t from 0 to $T - 1$, we obtain

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

503 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
 504 $\Psi^t \geq 0$ for all $t \in [T]$, to get the following inequality:

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

505 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

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

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

507 **E.2 Generic Lemmas**

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

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g^{t+1} - h^{t+1}\|^2 \right] \right] \\ & \leq \frac{2\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \frac{a^2((2\omega + 1)p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2, \end{aligned} \quad (20)$$

510 and

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \\ & \leq \frac{2\omega}{p_a} \|k_i^{t+1}\|^2 + \left(\frac{a^2(2\omega + 1 - p_a)}{p_a} + (1-a)^2 \right) \|g_i^t - h_i^t\|^2 \quad \forall i \in [n]. \end{aligned} \quad (21)$$

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

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

512 where we used (14). Due to Assumption 8, we have

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

513 and

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

514 Thus, we can get

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

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

516 $p_a \mathcal{C}_i \left(\frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_i^{t+1}$, and obtain

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

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

517 From Assumption 7, we have

$$\begin{aligned}
& \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g^{t+1} - h^{t+1}\|^2 \right] \right] \\
& \leq \frac{\omega p_a}{n^2} \sum_{i=1}^n \left\| \frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right\|^2 + \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \\
& = \frac{\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1} - a (g_i^t - h_i^t)\|^2 + \frac{a^2 (p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2 \\
& \stackrel{(13)}{\leq} \frac{2\omega}{n^2 p_a} \sum_{i=1}^n \|k_i^{t+1}\|^2 + \frac{a^2 ((2\omega + 1)p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g^t - h^t\|^2.
\end{aligned}$$

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

$$\begin{aligned}
& \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \\
& = \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g_i^{t+1} - h_i^{t+1} - \mathbb{E}_{\mathcal{C}} [\mathbb{E}_{p_a} [g_i^{t+1} - h_i^{t+1}]]\|^2 \right] \right] + \|\mathbb{E}_{\mathcal{C}} [\mathbb{E}_{p_a} [g_i^{t+1} - h_i^{t+1}]]\|^2 \\
& = \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g_i^{t+1} - h_i^{t+1} - g_i^t + a (g_i^t - h_i^t) + h_i^t\|^2 \right] \right] + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& = p_a \mathbb{E}_{\mathcal{C}} \left[\left\| \mathcal{C}_i \left(\frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \frac{1}{p_a} k_i^{t+1} + a (g_i^t - h_i^t) \right\|^2 \right] \\
& \quad + a^2 (1-p_a) \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(14)}{=} p_a \mathbb{E}_{\mathcal{C}} \left[\left\| \mathcal{C}_i \left(\frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left(\frac{1}{p_a} k_i^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& \quad + a^2 \frac{(1-p_a)^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& \quad + a^2 (1-p_a) \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \leq \frac{\omega}{p_a} \|k_i^{t+1} - a (g_i^t - h_i^t)\|^2 \\
& \quad + \frac{a^2 (1-p_a)}{p_a} \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(13)}{\leq} \frac{2\omega}{p_a} \|k_i^{t+1}\|^2 + \frac{a^2 (2\omega + 1 - p_a)}{p_a} \|g_i^t - h_i^t\|^2 + (1-a)^2 \|g_i^t - h_i^t\|^2.
\end{aligned}$$

519

□

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

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

521 *Proof.* Due to Lemma 2 and the update step from Line 5 in Algorithm 1, we have

$$\begin{aligned}
& \mathbb{E}_{t+1} [f(x^{t+1})] \\
& \leq \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2 \right] \\
& = \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - h^t + h^t - \nabla f(x^t)\|^2 \right] \\
& \stackrel{(14)}{\leq} \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma (\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2) \right].
\end{aligned}$$

522 Let us fix some constants $\kappa, \eta \in [0, \infty)$ that we will define later. Combining the last inequality,
523 bounds (20), (21) and using the law of total expectation, we get

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

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

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

525 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
 526 $\left(\frac{\gamma a((2\omega+1)p_a - p_{aa})}{np_a^2} + \eta \left(\frac{a^2(2\omega+1-p_a)}{p_a} + (1-a)^2 \right) \right) \leq \eta$. Thus

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

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

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

528 □

529 E.3 Proof for DASHA-PP

530 **Lemma 7.** Suppose that Assumptions 3 and 8 hold. For h_i^{t+1} and k_i^{t+1} from Algorithm 1 (DASHA-PP)
 531 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.

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

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

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

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

533 Using

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

534 and (14), we have

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

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

$$\begin{aligned} & \mathbb{E}_{p_a} \left[\left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \\ & \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left\| k_i^{t+1} - k_i^{t+1} \right\|^2 + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| k_i^{t+1} \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ &= \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ & \stackrel{(13)}{\leq} \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\ & \leq \frac{2(p_a - p_{aa}) \hat{L}^2}{n p_a^2} \left\| x^{t+1} - x^t \right\|^2 + \frac{2b^2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \end{aligned}$$

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

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

537 Finally, the third inequality of the theorem follows from (13) and Assumption 3. \square

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

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

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

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

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

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

542 After rearranging the terms, we get

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

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

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

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

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

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

545 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

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

546 and

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

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

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

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

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

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

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

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

550 Using Lemma 4 and the assumption about γ , we get

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

551 It is left to apply Lemma 3 with

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

552 to conclude the proof. \square

553 **E.4 Proof for DASHA-PP-PAGE**

554 Let us denote

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

555 $h_1^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,1}^{t+1}$, and $h_2^{t+1} := \frac{1}{n} \sum_{i=1}^n h_{i,2}^{t+1}$. Note, that

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

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

1.

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

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.

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

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

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

559 Using

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

560 and

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

561 we obtain

$$\begin{aligned} & \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\mathbb{E}_{p_{\text{page}}} \left[\|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \right] \\ & \stackrel{(14)}{=} p_{\text{page}} \mathbb{E}_{p_a} \left[\|h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}]\|^2 \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}]\|^2 \right] \right] \end{aligned}$$

$$\begin{aligned}
& + p_{\text{page}} \left\| \mathbb{E}_{p_a} [h_1^{t+1}] - \nabla f(x^{t+1}) \right\|^2 + (1 - p_{\text{page}}) \left\| \mathbb{E}_B [\mathbb{E}_{p_a} [h_2^{t+1}]] - \nabla f(x^{t+1}) \right\|^2 \\
& = p_{\text{page}} \mathbb{E}_{p_a} \left[\left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[\left\| h_2^{t+1} - \mathbb{E}_{p_a} [h_2^{t+1}] \right\|^2 \right] \\
& + \left(p_{\text{page}} \left(1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \left\| h^t - \nabla f(x^t) \right\|^2. \tag{22}
\end{aligned}$$

562 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 $s_i = k_{i,1}^{t+1}$
563 to obtain

$$\begin{aligned}
& \mathbb{E}_{p_a} \left[\left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] \\
& \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left\| k_{i,1}^{t+1} - k_{i,1}^{t+1} \right\|^2 + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| k_{i,1}^{t+1} \right\|^2 \\
& = \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
& \stackrel{(13)}{\leq} \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2.
\end{aligned}$$

564 From Assumption 3, we have

$$\begin{aligned}
& \mathbb{E}_{p_a} \left[\left\| h_1^{t+1} - \mathbb{E}_{p_a} [h_1^{t+1}] \right\|^2 \right] \\
& \leq \frac{2(p_a - p_{aa})\hat{L}^2}{n p_a^2} \left\| x^{t+1} - x^t \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{page}}^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2. \tag{23}
\end{aligned}$$

565 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-
566 batches in the algorithm are independent, we can use Lemma 1 with $r_i = h_i^t$ and $s_i = k_{i,2}^{t+1}$
567 to obtain

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

568 Next, we use Assumptions 3 and 4 to get

$$\mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_2^{t+1} - \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_2^{t+1} \right] \right] \right\|^2 \right] \right] \leq \left(\frac{L_{\max}^2}{np_a B} + \frac{(p_a - p_{aa}) \hat{L}^2}{np_a^2} \right) \|x^{t+1} - x^t\|^2. \quad (24)$$

569 Applying (23) and (24) into (22), we get

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

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

$$\begin{aligned} & \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\mathbb{E}_{p_{\text{page}}} \left[\left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \right] \\ & = p_{\text{page}} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,1}^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,2}^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \\ & \stackrel{(14)}{=} p_{\text{page}} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & \quad + p_{\text{page}} \left\| \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,1}^{t+1} \right] \right] - \nabla f_i(x^{t+1}) \right\|^2 + (1 - p_{\text{page}}) \left\| \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,2}^{t+1} \right] \right] - \nabla f_i(x^{t+1}) \right\|^2 \\ & = p_{\text{page}} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,1}^{t+1} - \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,1}^{t+1} \right] \right] \right\|^2 \right] \right] + (1 - p_{\text{page}}) \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,2}^{t+1} - \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[h_{i,2}^{t+1} \right] \right] \right\|^2 \right] \right] \\ & \quad + \left(p_{\text{page}} \left(1 - \frac{b}{p_{\text{page}}} \right)^2 + (1 - p_{\text{page}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2. \quad (25) \end{aligned}$$

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

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

572 Considering (13) and Assumption 3, we obtain

$$\begin{aligned} & \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,1}^{t+1} - \mathbb{E}_{p_a} [h_{i,1}^{t+1}] \right\|^2 \right] \right] \\ & \leq \frac{2(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2 + \frac{2(1-p_a)b^2}{p_a p_{\text{page}}^2} \|h_i^t - \nabla f_i(x^t)\|^2. \end{aligned} \quad (26)$$

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

$$\begin{aligned} & \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_{i,2}^{t+1} - \mathbb{E}_{p_a} [h_{i,2}^{t+1}] \right\|^2 \right] \right] \\ & = p_a \mathbb{E}_B \left[\left\| h_i^t + \frac{1}{p_a} k_{i,2}^{t+1} - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + (1-p_a) \mathbb{E}_B \left[\left\| h_i^t - (h_i^t + \nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & = p_a \mathbb{E}_B \left[\left\| \frac{1}{p_a} k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\ & \quad + (1-p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \stackrel{(14)}{=} \frac{1}{p_a} \mathbb{E}_B \left[\left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{(1-p_a)^2}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \quad + (1-p_a) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & = \frac{1}{p_a} \mathbb{E}_B \left[\left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{1-p_a}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \leq \frac{1}{p_a} \mathbb{E}_B \left[\left\| k_{i,2}^{t+1} - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] + \frac{(1-p_a)L_i^2}{p_a} \|x^{t+1} - x^t\|^2, \end{aligned} \quad (27)$$

574 where we used Assumption 3. By plugging (26) and (27) into (25), we get

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

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

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

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

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

$$\begin{aligned}
&\mathbb{E}_B \left[\mathbb{E}_{p_{\text{page}}} \left[\|k_i^{t+1}\|^2 \right] \right] \\
&= p_{\text{page}} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[\left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) \right\|^2 \right] \\
&\stackrel{(14)}{=} p_{\text{page}} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{page}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[\left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + (1-p_{\text{page}}) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\stackrel{(13)}{\leq} 2p_{\text{page}} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\quad + (1-p_{\text{page}}) \mathbb{E}_B \left[\left\| \frac{1}{B} \sum_{j \in I_i^t} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + (1-p_{\text{page}}) \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\leq 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \frac{2b^2}{p_{\text{page}}} \|h_i^t - \nabla f_i(x^t)\|^2
\end{aligned}$$

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

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

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

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

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

579

□

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

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

582 and $g_i^0 = h_i^0 = \nabla f_i(x^0)$ for all $i \in [n]$ in Algorithm 1 (DASHA-PP-PAGE) then $\mathbb{E} [\|\nabla f(\hat{x}^T)\|^2] \leq$
583 $\frac{2\Delta_0}{\gamma T}$.

584 *Proof.* Let us fix constants $\nu, \rho \in [0, \infty)$ that we will define later. Considering Lemma 6, Lemma 8,
585 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}$$

586 After rearranging the terms, we get

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

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

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

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

587 therefore

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

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

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

588 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

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

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

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

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

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

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

591 Using Lemma 4 and the assumption about γ , we get

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

592 It is left to apply Lemma 3 with

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

593 to conclude the proof. \square

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

$$T := \mathcal{O} \left(\frac{\Delta_0}{\varepsilon} \left[L + \frac{\omega}{p_a \sqrt{n}} \left(\hat{L} + \frac{L_{\max}}{\sqrt{B}} \right) + \frac{1}{p_a} \sqrt{\frac{m}{n}} \left(\frac{1}{\sqrt{B}} + \frac{L_{\max}}{B} \right) \right] \right) \quad (10)$$

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

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

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

599 steps to get ε -solution. Using the choice of p_{mega} and the definition of \mathbb{I}_{p_a} , we can get (10).

600 Note that the expected number of gradients calculations at each communication round equals $p_{\text{mega}}m +$
601 $(1 - p_{\text{mega}})B = \frac{2mB}{m+B} \leq 2B$. \square

602 **Corollary 2.** Suppose that assumptions of Corollary 1 hold, $B \leq \min \left\{ \frac{1}{p_a} \sqrt{\frac{m}{n}}, \frac{L_{\max}^2}{\frac{1}{p_a} L^2} \right\}$ ⁷, and we
603 use the unbiased compressor RandK with $K = \Theta(Bd/\sqrt{m})$. Then the communication complexity of
604 Algorithm 1 is $\mathcal{O} \left(d + \frac{L_{\max} \Delta_0 d}{p_a \varepsilon \sqrt{n}} \right)$, and the expected number of gradient calculations per node equals
605 $\mathcal{O} \left(m + \frac{L_{\max} \Delta_0 \sqrt{m}}{p_a \varepsilon \sqrt{n}} \right)$.

⁷If $\mathbb{I}_{p_a} = 0$, then $\frac{L_{\max}^2}{\frac{1}{p_a} L^2} = +\infty$

606 *Proof.* The communication complexity equals

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

607 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

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

608 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

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

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

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

610 □

611 E.5 Proof for DASHA-PP-FINITE-MVR

612 **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
613 (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.

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

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

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

615 thus

$$\begin{aligned}
& \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\
& \stackrel{(14)}{=} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h^{t+1} - \mathbb{E}_B [\mathbb{E}_{p_a} [h^{t+1}]]\|^2 \right] \right] + (1-b)^2 \|h^t - \nabla f(x^t)\|^2.
\end{aligned}$$

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

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

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

618 $p_a = \frac{B}{m}$, and $p_{aa} = \frac{B(B-1)}{m(m-1)}$:

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

$$\begin{aligned}
&\leq \frac{1}{n^2 p_a} \sum_{i=1}^n \left(\frac{m-B}{Bm(m-1)} \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \right) \\
&\quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
&\leq \frac{1}{n^2 p_a Bm} \sum_{i=1}^n \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \\
&\quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
&\stackrel{(13)}{\leq} \frac{2}{n^2 p_a Bm} \sum_{i=1}^n \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) \right\|^2 + \frac{2b^2}{n^2 p_a Bm} \sum_{i=1}^n \sum_{j=1}^m \left\| h_{ij}^t - \nabla f_{ij}(x^t) \right\|^2 \\
&\quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2.
\end{aligned}$$

619 Due to Assumptions 3 and 4, we have

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

620 Let us get the bound for the second inequality:

$$\begin{aligned}
&\mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right] \\
&\stackrel{(14)}{=} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_i^{t+1} - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \right] \\
&\quad + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
&= p_a \mathbb{E}_B \left[\left\| h_i^t + \frac{1}{p_a} k_i^{t+1} - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \right] \\
&\quad + (1-p_a) \left\| h_i^t - (\nabla f_i(x^{t+1}) + (1-b)(h_i^t - \nabla f_i(x^t))) \right\|^2 \\
&\quad + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
&\stackrel{(14)}{=} \frac{1}{p_a} \mathbb{E}_B \left[\left\| k_i^{t+1} - \mathbb{E}_B[k_i^{t+1}] \right\|^2 \right] \\
&\quad + \frac{1-p_a}{p_a} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-b)^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2.
\end{aligned}$$

621 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

622 $p_{aa} = \frac{B(B-1)}{m(m-1)}$:

$$\mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\left\| h_i^{t+1} - \nabla f_i(x^{t+1}) \right\|^2 \right] \right]$$

$$\begin{aligned}
&\leq \frac{1}{p_a} \left(\frac{m-B}{Bm(m-1)} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \right) \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \frac{1}{p_a B m} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\stackrel{(13)}{\leq} \frac{2}{p_a B m} \sum_{j=1}^m \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 + \frac{2(1-p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\quad + \frac{2b^2}{p_a B m} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left(\frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
&\leq \left(\frac{2L_{\max}^2}{p_a B} + \frac{2(1-p_a)L_i^2}{p_a} \right) \|x^{t+1} - x^t\|^2 \\
&\quad + \frac{2b^2}{p_a B m} \sum_{j=1}^m \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 + \left(\frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2,
\end{aligned}$$

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

$$\begin{aligned}
624 \quad &\mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] : \\
&\mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h_{ij}^{t+1} - \nabla f_{ij}(x^{t+1})\|^2 \right] \right] \\
&\stackrel{(14)}{=} \mathbb{E}_B \left[\mathbb{E}_{p_a} \left[\|h_{ij}^{t+1} - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t)))\|^2 \right] \right] \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&= \frac{p_a B}{m} \mathbb{E}_B \left[\left\| h_{ij}^t + \frac{m}{B p_a} (\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))) - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t))) \right\|^2 \right] \\
&\quad + \left(1 - \frac{p_a B}{m} \right) \|h_{ij}^t - (\nabla f_{ij}(x^{t+1}) + (1-b)(h_{ij}^t - \nabla f_{ij}(x^t)))\|^2 \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&= \frac{\left(1 - \frac{p_a B}{m} \right)^2}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + \left(1 - \frac{p_a B}{m} \right) \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&= \frac{\left(1 - \frac{p_a B}{m} \right)}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2 \\
&\stackrel{(13)}{\leq} \frac{2 \left(1 - \frac{p_a B}{m} \right)}{\frac{p_a B}{m}} \|\nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t)\|^2 + \left(\frac{2 \left(1 - \frac{p_a B}{m} \right) b^2}{\frac{p_a B}{m}} + (1-b)^2 \right) \|h_{ij}^t - \nabla f_{ij}(x^t)\|^2.
\end{aligned}$$

625 It is left to consider Assumption 4:

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

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

$$\begin{aligned} & \mathbb{E}_B \left[\left\| k_i^{t+1} \right\|^2 \right] \\ & \stackrel{(14)}{=} \mathbb{E}_B \left[\left\| k_i^{t+1} - \mathbb{E}_B \left[k_i^{t+1} \right] \right\|^2 \right] \\ & \quad + \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2. \end{aligned}$$

627 Using Lemma 1, we get

$$\begin{aligned} & \mathbb{E}_B \left[\left\| k_i^{t+1} \right\|^2 \right] \\ & \leq \frac{m-B}{Bm(m-1)} \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \\ & \quad + \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \leq \frac{1}{Bm} \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) - b(h_{ij}^t - \nabla f_{ij}(x^t)) \right\|^2 \\ & \quad + \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\ & \stackrel{(13)}{\leq} \frac{2}{Bm} \sum_{j=1}^m \left\| \nabla f_{ij}(x^{t+1}) - \nabla f_{ij}(x^t) \right\|^2 + 2 \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \\ & \quad + \frac{2b^2}{Bm} \sum_{j=1}^m \left\| h_{ij}^t - \nabla f_{ij}(x^t) \right\|^2 + 2b^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\ & \leq \left(\frac{2L_{\max}^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 \\ & \quad + \frac{2b^2}{Bm} \sum_{j=1}^m \left\| h_{ij}^t - \nabla f_{ij}(x^t) \right\|^2 + 2b^2 \left\| h_i^t - \nabla f_i(x^t) \right\|^2, \end{aligned}$$

628 where we used Assumptions 3 and 4. □

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

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

629 $g_i^0 = h_i^0 = \nabla f_i(x^0)$ 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

630 (DASHA-PP-FINITE-MVR) then $\mathbb{E} \left[\left\| \nabla f(\hat{x}^T) \right\|^2 \right] \leq \frac{2\Delta_0}{\gamma^T}$.

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

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

$$\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] \\
& + \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} \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] \\
& + \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} \left[\|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \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] \\
& + \nu \mathbb{E} \left(\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 \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)
\end{aligned}$$

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

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

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

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

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

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

634 After rearranging the terms, we get

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

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

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

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

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

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

636 therefore

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

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

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

$$= \frac{24b^2\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma b}{np_aB}.$$

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

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

639 and

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

640 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

641 have

$$\begin{aligned} & \frac{4\gamma\omega(2\omega+1)}{np_a^2} \left(\frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \\ & + \left(\frac{2\gamma L_{\max}^2}{bnp_aB} + \frac{2\gamma(p_a - p_{aa})\hat{L}^2}{bnp_a^2} \right) \\ & + \left(\frac{8b\gamma\omega(2\omega+1)}{np_a^2} + \frac{2\gamma(p_a - p_{aa})}{np_a^2} \right) \left(\frac{2L_{\max}^2}{p_aB} + \frac{2(1-p_a)\hat{L}^2}{p_a} \right) \\ & + \left(\frac{24b\gamma\omega(2\omega+1)}{np_a^2B} + \frac{6\gamma}{np_aB} \right) \frac{2mL_{\max}^2}{p_aB} \\ & \leq \frac{12\gamma\omega(2\omega+1)}{np_a^2} \left(\frac{2L_{\max}^2}{B} + 2\hat{L}^2 \right) \end{aligned}$$

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

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

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

643 All in all, we have

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

644 Using Lemma 4 and the assumption about γ , we get

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

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

645 It is left to apply Lemma 3 with

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

646 to conclude the proof. \square

647 E.6 Proof for DASHA-PP-MVR

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

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

1.

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

2.

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

3.

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

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

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

652 Using

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

653 and (14), we have

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

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

$$\begin{aligned} & \mathbb{E}_k \left[\mathbb{E}_{p_a} \left[\|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] \right] \\ & \leq \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|k_i^{t+1} - \mathbb{E}_k [k_i^{t+1}]\|^2 \right] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\mathbb{E}_k [k_i^{t+1}]\|^2 + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ &= \frac{1}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})) \right. \\ & \quad \left. - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\ & \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ & \stackrel{(13)}{\leq} \frac{2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|b(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1}))\|^2 \right] \\ & \quad + \frac{2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|(1-b)(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)))\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\ & \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2 \\ &= \frac{2b^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\ & \quad + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\ & \quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\ & \quad + (1-b)^2 \|h^t - \nabla f(x^t)\|^2. \end{aligned}$$

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

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

$$\begin{aligned}
&\mathbb{E}_k \left[\mathbb{E}_{p_a} \left[\left\| h^{t+1} - \nabla f(x^{t+1}) \right\|^2 \right] \right] \\
&\leq \frac{2b^2 \sigma^2}{n p_a B} \\
&\quad + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2 \\
&\stackrel{(13)}{\leq} \frac{2b^2 \sigma^2}{n p_a B} \\
&\quad + \frac{2(1-b)^2}{n^2 p_a} \sum_{i=1}^n \mathbb{E}_k \left[\left\| \nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2. \\
&= \frac{2b^2 \sigma^2}{n p_a B} \\
&\quad + \frac{2(1-b)^2}{n^2 p_a B^2} \sum_{i=1}^n \sum_{j=1}^B \mathbb{E}_k \left[\left\| \nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)) \right\|^2 \right] \\
&\quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2} \sum_{i=1}^n \left\| h_i^t - \nabla f_i(x^t) \right\|^2 \\
&\quad + (1-b)^2 \left\| h^t - \nabla f(x^t) \right\|^2,
\end{aligned}$$

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

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

659 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{(14)}{=} \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{(13)}{\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] \\
&\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}$$

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

$$\begin{aligned}
& \mathbb{E}_k \left[\mathbb{E}_{p_a} \left[\|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \right] \\
&= \frac{2b^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
&\quad + \frac{2(1-b)^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
&\quad + \frac{1-p_a}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\quad + (1-b)^2 \|h_i^t - \nabla f_i(x^t)\|^2. \\
&\stackrel{(13)}{\leq} \frac{2b^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
&\quad + \frac{2(1-b)^2}{p_a B^2} \sum_{j=1}^B \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_{ij}^{t+1}) - \nabla f_i(x^t; \xi_{ij}^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
&\quad + \frac{2(1-p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + \left(\frac{2(1-p_a)b^2}{p_a} + (1-b)^2 \right) \|h_i^t - \nabla f_i(x^t)\|^2
\end{aligned}$$

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

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

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

$$\begin{aligned}
& \mathbb{E}_k \left[\|k_i^{t+1}\|^2 \right] \\
&= \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1}))\|^2 \right] \\
&\stackrel{(14)}{=} \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - b(h_i^t - \nabla f_i(x^t; \xi_i^{t+1})) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t)))\|^2 \right] \\
&\quad + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&= \mathbb{E}_k \left[\|b(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})) + (1-b)(\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t)))\|^2 \right] \\
&\quad + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t) - b(h_i^t - \nabla f_i(x^t))\|^2 \\
&\stackrel{(13)}{\leq} 2b^2 \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^{t+1})\|^2 \right] \\
&\quad + 2(1-b)^2 \mathbb{E}_k \left[\|\nabla f_i(x^{t+1}; \xi_i^{t+1}) - \nabla f_i(x^t; \xi_i^{t+1}) - (\nabla f_i(x^{t+1}) - \nabla f_i(x^t))\|^2 \right] \\
&\quad + 2 \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

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

$$\mathbb{E}_k \left[\|k_i^{t+1}\|^2 \right]$$

$$\leq \frac{2b^2\sigma^2}{B} + \left(\frac{2(1-b)^2L_\sigma^2}{B} + 2L_i^2 \right) \|x^{t+1} - x^t\|^2 + 2b^2 \|h_i^t - \nabla f_i(x^t)\|^2.$$

664

□

665 **Theorem 4.** Suppose that Assumptions 1, 2, 3, 5, 6, 7 and 8 hold. Let us take $a = \frac{p_a}{2\omega+1}$,
 666 $b \in \left(0, \frac{p_a}{2-p_a}\right]$, $\gamma \leq \left(L + \left[\frac{48\omega(2\omega+1)}{np_a^2} \left(\widehat{L}^2 + \frac{(1-b)^2L_\sigma^2}{B}\right) + \frac{12}{np_ab} \left(\left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2 + \frac{(1-b)^2L_\sigma^2}{B}\right)\right]^{1/2}\right)^{-1}$, and
 667 $g_i^0 = h_i^0$ for all $i \in [n]$ in Algorithm 1 (DASHA-PP-MVR). Then

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

668 *Proof.* Let us fix constants $\nu, \rho \in [0, \infty)$ that we will define later. Considering Lemma 6, Lemma 10,
 669 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] \\ &+ \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] \end{aligned}$$

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

670 After rearranging the terms, we get

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

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

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

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

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

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

673 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

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

674 and

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

675 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

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

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

676 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] + \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{12\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{3\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) \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} \\ & = \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{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}$$

677 Also, we can simplify the last term:

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

678 thus

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

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

679 Using Lemma 4 and the assumption about γ , we get

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

680 It is left to apply Lemma 3 with

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

681 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

682 **Corollary 3.** Suppose that assumptions from Theorem 4 hold, momentum $b =$
683 $\Theta \left(\min \left\{ \frac{p_a}{\omega} \sqrt{\frac{n\varepsilon B}{\sigma^2}}, \frac{p_a n \varepsilon B}{\sigma^2} \right\} \right)$, $\frac{\sigma^2}{n\varepsilon B} \geq 1$, 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 $i \in [n]$,
684 and batch size $B_{\text{init}} = \Theta \left(\frac{\sqrt{p_a B}}{b} \right)$, then Algorithm 1 (DASHA-PP-MVR) needs

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

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

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

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

688 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 $\frac{1}{b} =$

689 $\Theta \left(\max \left\{ \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{n\varepsilon B}}, \frac{\sigma^2}{p_a n \varepsilon B} \right\} \right) \leq \Theta \left(\max \left\{ \frac{\omega^2}{p_a}, \frac{\sigma^2}{p_a n \varepsilon B} \right\} \right)$, thus

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

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

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

691 Let us bound the norms:

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

692 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

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

694 Using the choice of B_{init} and b , we obtain

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

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

697 **Corollary 4.** Suppose that assumptions of Corollary 3 hold, batch size $B \leq \min \left\{ \frac{\sigma}{p_a \sqrt{\varepsilon n}}, \frac{L_\sigma^2}{\mathbb{1}_{p_a}^2 \hat{L}^2} \right\}$,
698 we take RandK compressors with $K = \Theta \left(\frac{Bd\sqrt{\varepsilon n}}{\sigma} \right)$. Then the communication complexity equals
699 $\mathcal{O} \left(\frac{d\sigma}{\sqrt{p_a} \sqrt{n\varepsilon}} + \frac{L_\sigma \Delta_0 d}{p_a \sqrt{n\varepsilon}} \right)$, and the expected number of stochastic gradient calculations per node equals
700 $\mathcal{O} \left(\frac{\sigma^2}{\sqrt{p_a} n \varepsilon} + \frac{L_\sigma \Delta_0 \sigma}{p_a \varepsilon^{3/2} n} \right)$.

701 *Proof.* The communication complexity equals

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

702 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

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

703 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
704 complexity equals

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

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

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

706

□

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:

$$\|\nabla f(x)\|^2 \geq 2\mu(f(x) - f^*), \quad \forall x \in \mathbb{R}, \quad (28)$$

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

Under Polyak-Łojasiewicz condition, a (random) point \hat{x} is ε -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 ε -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 ε -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 ε -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 ε -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.

730 F.3 Stochastic Setting

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

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

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

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

732 The provided theorems states that to get ε -solution DASHA-PP have to run

$$\tilde{\mathcal{O}} \left(\underbrace{\frac{\omega+1}{p_a} + \frac{\omega}{p_a} \sqrt{\frac{\sigma^2}{\mu n \varepsilon B}}}_{\mathcal{P}_2} + \frac{\sigma^2}{p_a \mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{p_a \mu \sqrt{n}} \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \underbrace{\frac{\sigma}{p_a n \mu^{3/2} \sqrt{\varepsilon B}} \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}} \right)}_{\mathcal{P}_1} \right) \quad (29)$$

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

734 $\Theta \left(\min \left\{ \frac{p_a}{\omega^2}, \frac{p_a \mu n \varepsilon B}{\sigma^2} \right\} \right).$

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

$$\tilde{\mathcal{O}} \left(\omega + \frac{\sigma^2}{\mu n \varepsilon B} + \frac{L}{\mu} + \frac{\omega}{\mu \sqrt{n}} \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{n \mu^{3/2} \sqrt{\varepsilon B}} \left(\frac{L_\sigma}{\sqrt{B}} \right) \right) \quad (30)$$

736 communication rounds to get ε -solution⁸.

737 In the stochastic setting, the comparison is a little bit more complicated. As in the finite-sum setting,
 738 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 (29).

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

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

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

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

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

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

747 Up to Lipschitz constants, DASHA-PP-MVR has the state-of-the-art oracle complexity under PL-
 748 condition (see (Li et al., 2021a)). Moreover, DASHA-PP-MVR has the state-of-the-art communication
 749 complexity of DASHA for a small enough μ .

⁸For simplicity, we omitted $\frac{d}{\zeta_C}$ term from the complexity in the stochastic setting, where ζ_C is defined in Definition 12. For instance, for the RandK compressor (see Definition 5 and Theorem 6), $\zeta_C = K$ and $\frac{d}{\zeta_C} = \Theta(\omega)$.

750 **F.4 Proofs of Theorems**

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

753 **F.4.1 Standard Lemma under Polyak-Łojasiewicz Condition**

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

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

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

$$\mathbb{E} [f(x^T) - f^*] \leq (1 - \gamma\mu)^T ((f(x^0) - f^*) + \gamma \Psi^0) + \frac{C}{\mu}. \quad (31)$$

757 *Proof.* We subtract f^* and use PŁ-condition (28) to get

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

758 Unrolling the inequality, we have

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

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

760 **F.4.2 Generic Lemma**

761 We now provide the counterpart of Lemma **6**.

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

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

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

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

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

765 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

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

766 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
767 $\eta = \frac{4\gamma((2\omega+1)p_a-p_{aa})}{np_a^2}$, 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

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

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

769 **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 $a = \frac{p_a}{2\omega+1}$, $b = \frac{p_a}{2-p_a}$,

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

770 and $h_i^0 = g_i^0 = \nabla f_i(x^0)$ for all $i \in [n]$ in Algorithm 1 (DASHA-PP), then $\mathbb{E}[f(x^T)] - f^* \leq$
 771 $(1 - \gamma\mu)^T \Delta_0$.

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

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

774 After rearranging the terms, we get

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

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

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

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

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

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

777 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

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

778 and

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

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

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

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

780 Using Lemma 4 and the assumption about γ , we get

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

781 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

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

782 In the view of Lemma 11 with

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

783 we can conclude the proof of the theorem. \square

784 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

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

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

785 and $h_i^0 = g_i^0 = \nabla f_i(x^0)$ for all $i \in [n]$ in Algorithm 1 (DASHA-PP-PAGE), then $\mathbb{E}[f(x^T)] - f^* \leq$
 786 $(1 - \gamma\mu)^T \Delta_0$.

787 *Proof.* Let us fix constants $\nu, \rho \in [0, \infty)$ that we will define later. Considering Lemma 12, Lemma 8,
 788 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}$$

789 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

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

790 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

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

791 therefore

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

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

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

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

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

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

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

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

794 Using Lemma 4 and the assumption about γ , we get

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

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

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

796 It is left to apply Lemma 11 with

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

797 to conclude the proof. \square

798 **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 $a = \frac{p_a}{2\omega+1}$,
 $b \in \left(0, \frac{p_a}{2-p_a}\right]$,

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

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

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

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

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

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

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

802 After rearranging the terms, we get

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

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

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

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

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

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

805 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

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

806 and

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

807 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

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

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

808 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}$$

809 Also, we can simplify the last term:

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

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

810 thus

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

811 Using Lemma 4 and the assumption about γ , we get

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

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

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

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

813 It is left to apply Lemma 11 with

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

814 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

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

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

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

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

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

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

819 communication rounds in the stochastic settings to get ε -solution. Note that $K = \mathcal{O} \left(\frac{d}{p_a\sqrt{n}} \right)$.
820 Moreover, we can skip the initialization procedure and initialize h_i^0 and g_i^0 , for instance, with zeros
821 because the initialization error is under a logarithm. Considering Theorem 6, the communication
822 complexity equals

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

823 The expected number of stochastic gradient calculations per node equals

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

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

824

□

825 **G Description of DASHA-PP-SYNC-MVR**

826 By analogy to (Tyurin and Richtárik, 2023), we provide a “synchronized” version of the algorithm.
 827 With a small probability, participating nodes calculate and send a mega batch without compression.
 828 This helps us to resolve the suboptimality of DASHA-PP-MVR w.r.t. ω . Note that this suboptimality is
 829 not a problem. We show in Corollary 4 that DASHA-PP-MVR can have the optimal oracle complexity
 830 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.

```

831 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 $B' \geq B \geq 1$

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

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

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

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

835 **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
836 $\|x\|_0$ is the number of nonzero components of $x \in \mathbb{R}^d$. Let $\zeta_{\mathcal{C}} = \max_{i \in [n]} \zeta_{\mathcal{C}_i}$.

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

838 In the next corollary, we choose particular algorithm parameters to reveal the communication and
839 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

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

840 communication rounds to get an ε -solution, the expected communication complexity is equal to
841 $\mathcal{O}(d + \zeta_{\mathcal{C}} T)$, and the expected number of stochastic gradient calculations per node equals $\mathcal{O}(B_{\text{init}} +$
842 $BT)$, where $\zeta_{\mathcal{C}}$ is the expected density from Definition 12.

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

845 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 $B' \geq B \geq 1$,

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

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

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

847 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

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

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

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

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

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

Comparing (32) with the rate of DASHA-PP-MVR (29), one can see that DASHA-PP-SYNC-MVR improves the suboptimal term \mathcal{P}_2 from (29). However, Corollary 5 reveals that we can escape these suboptimal regimes by choosing the parameter K of RandK compressors in a particular way.

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

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

Let us denote

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

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

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

and

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

First, we will prove two lemmas.

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

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

874 and

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

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

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

876 Using

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

877 and

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

878 we have

$$\begin{aligned} & \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\mathbb{E}_{p_{\text{mega}}} \left[\|g^{t+1} - h^{t+1}\|^2 \right] \right] \right] \\ & \stackrel{(14)}{=} p_{\text{mega}} \mathbb{E}_{p_a} \left[\|g_1^{t+1} - h_1^{t+1} - \mathbb{E}_{p_a} [g_1^{t+1} - h_1^{t+1}]\|^2 \right] \\ & \quad + (1-p_{\text{mega}}) \mathbb{E}_{\mathcal{C}} \left[\mathbb{E}_{p_a} \left[\|g_2^{t+1} - h_2^{t+1} - \mathbb{E}_{p_a} [g_2^{t+1} - h_2^{t+1}]\|^2 \right] \right] \\ & \quad + (1-a)^2 \|g^t - h^t\|^2. \end{aligned}$$

879 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
880 $s_i = p_a \mathcal{C}_i \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - k_{i,2}^{t+1}$, to obtain

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

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

881 In the last inequality, we use Assumption 7. Next, using (13), we have

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

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

$$\begin{aligned}
& \mathbb{E}_C \left[\mathbb{E}_{p_a} \left[\mathbb{E}_{p_{\text{mega}}} \left[\|g_i^{t+1} - h_i^{t+1}\|^2 \right] \right] \right] \\
& = p_{\text{mega}} \mathbb{E}_{p_a} \left[\|g_{i,1}^{t+1} - h_{i,1}^{t+1}\|^2 \right] + (1 - p_{\text{mega}}) \mathbb{E}_C \left[\mathbb{E}_{p_a} \left[\|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& \stackrel{(14)}{=} p_{\text{mega}} \mathbb{E}_{p_a} \left[\|g_{i,1}^{t+1} - h_{i,1}^{t+1} - (1 - a) (g_i^t - h_i^t)\|^2 \right] + (1 - p_{\text{mega}}) \mathbb{E}_C \left[\mathbb{E}_{p_a} \left[\|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& + p_{\text{mega}} (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 + (1 - p_{\text{mega}}) \mathbb{E}_C \left[\mathbb{E}_{p_a} \left[\|g_{i,2}^{t+1} - h_{i,2}^{t+1}\|^2 \right] \right] \\
& + p_{\text{mega}} (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(14)}{=} \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 + (1 - p_{\text{mega}}) \mathbb{E}_C \left[\mathbb{E}_{p_a} \left[\|g_{i,2}^{t+1} - h_{i,2}^{t+1} - (1 - a) (g_i^t - h_i^t)\|^2 \right] \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[\left\| g_i^t + \mathcal{C}_i \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left(h_i^t + \frac{1}{p_a} k_{i,2}^{t+1} \right) - (1 - a) (g_i^t - h_i^t) \right\|^2 \right] \\
& + (1 - p_{\text{mega}}) (1 - p_a) \|g_i^t - h_i^t - (1 - a) (g_i^t - h_i^t)\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{p_{\text{mega}} (1 - p_a) a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[\left\| \mathcal{C}_i \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left(\frac{1}{p_a} k_{i,2}^{t+1} - a (g_i^t - h_i^t) \right) \right\|^2 \right]
\end{aligned}$$

$$\begin{aligned}
& + (1 - p_{\text{mega}}) (1 - p_a) a^2 \|g_i^t - h_i^t\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(14)}{=} \left(\frac{p_{\text{mega}}(1 - p_a)a^2}{p_a} + \frac{(1 - p_{\text{mega}})(1 - p_a)a^2}{p_a} \right) \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[\left\| \mathcal{C}_i \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& = \frac{(1 - p_a)a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + (1 - p_{\text{mega}}) p_a \mathbb{E}_C \left[\left\| \mathcal{C}_i \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) - \left(\frac{1}{p_a} k_{i,2}^{t+1} - \frac{a}{p_a} (g_i^t - h_i^t) \right) \right\|^2 \right] \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \leq \frac{(1 - p_a)a^2}{p_a} \|g_i^t - h_i^t\|^2 \\
& + \frac{(1 - p_{\text{mega}})\omega}{p_a} \|k_{i,2}^{t+1} - a(g_i^t - h_i^t)\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2 \\
& \stackrel{(13)}{\leq} \frac{2(1 - p_{\text{mega}})\omega}{p_a} \|k_{i,2}^{t+1}\|^2 + \left(\frac{(1 - p_a)a^2}{p_a} + \frac{2(1 - p_{\text{mega}})a^2\omega}{p_a} \right) \|g_i^t - h_i^t\|^2 \\
& + (1 - a)^2 \|g_i^t - h_i^t\|^2.
\end{aligned}$$

883

□

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

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

886

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

887 and

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

888 *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

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

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

889 and

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

890 we have

$$\begin{aligned}
&E_k [E_{p_a} [E_{p_{\text{mega}}} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]]] \\
&= p_{\text{mega}} E_k [E_{p_a} [\|h_1^{t+1} - \nabla f(x^{t+1})\|^2]] + (1 - p_{\text{mega}}) E_k [E_{p_a} [\|h_2^{t+1} - \nabla f(x^{t+1})\|^2]] \\
&\stackrel{(14)}{=} p_{\text{mega}} E_k [E_{p_a} [\|h_1^{t+1} - E_k [E_{p_a} [h_1^{t+1}]]\|^2]] + (1 - p_{\text{mega}}) E_k [E_{p_a} [\|h_2^{t+1} - E_k [E_{p_a} [h_2^{t+1}]]\|^2]] \\
&\quad + \left(p_{\text{mega}} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2.
\end{aligned}$$

891 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
892 obtain

$$\begin{aligned}
&E_k [E_{p_a} [E_{p_{\text{mega}}} [\|h^{t+1} - \nabla f(x^{t+1})\|^2]]] \\
&\leq p_{\text{mega}} \left(\frac{1}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) - \frac{b}{p_{\text{mega}}} (h_i^t - \nabla f_i(x^t)) \right\|^2 \right) \\
&\quad + (1 - p_{\text{mega}}) \left(\frac{1}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,2}^{t+1} - E_k [k_{i,2}^{t+1}]\|^2] + \frac{p_a - p_{aa}}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \right) \\
&\quad + \left(p_{\text{mega}} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2 \\
&\stackrel{(13)}{\leq} \frac{p_{\text{mega}}}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,1}^{t+1} - E_k [k_{i,1}^{t+1}]\|^2] \\
&\quad + \frac{1 - p_{\text{mega}}}{n^2 p_a} \sum_{i=1}^n E_k [\|k_{i,2}^{t+1} - E_k [k_{i,2}^{t+1}]\|^2] \\
&\quad + \frac{2(p_a - p_{aa})}{n^2 p_a^2} \sum_{i=1}^n \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
&\quad + \frac{2(p_a - p_{aa})b^2}{n^2 p_a^2 p_{\text{mega}}} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 + \left(p_{\text{mega}} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h^t - \nabla f(x^t)\|^2.
\end{aligned} \tag{33}$$

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

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

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

894 where we used independence of the mini-batch samples. Using (13), we get

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

895 Due to Assumptions 5 and 6, we have

$$\mathbb{E}_k \left[\left\| k_{i,1}^{t+1} - \mathbb{E}_k [k_{i,1}^{t+1}] \right\|^2 \right] \leq \frac{2b^2 \sigma^2}{B' p_{\text{mega}}^2} + \frac{2L_\sigma^2}{B'} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 \|x^{t+1} - x^t\|^2. \quad (34)$$

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

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

897 Due to Assumptions 6, we have

$$\mathbb{E}_k \left[\left\| k_{i,2}^{t+1} - \mathbb{E}_k [k_{i,2}^{t+1}] \right\|^2 \right] \leq \frac{L_\sigma^2}{B} \|x^{t+1} - x^t\|^2. \quad (35)$$

898 Plugging (34) and (35) into (33), we obtain

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

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

899 Using Assumption 3, we get

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

900 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{(14)}{=} 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) \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^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{(14)}{=} \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} \left\| \nabla f_i(x^{t+1}) - \nabla f_i(x^t) \right\|^2 \end{aligned}$$

$$\begin{aligned}
& + \left(p_{\text{mega}} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2 \\
& \stackrel{(13)}{\leq} \frac{p_{\text{mega}}}{p_a} \mathbb{E}_k \left[\|k_{i,1}^{t+1} - \mathbb{E}_k[k_{i,1}^{t+1}]\|^2 \right] \\
& + \frac{(1 - p_{\text{mega}})}{p_a} \mathbb{E}_k \left[\|k_{i,2}^{t+1} - \mathbb{E}_k[k_{i,2}^{t+1}]\|^2 \right] \\
& + \frac{2(1 - p_a)}{p_a} \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& + \frac{2(1 - p_a)b^2}{p_{\text{mega}}p_a} \|h_i^t - \nabla f_i(x^t)\|^2 + \left(p_{\text{mega}} \left(1 - \frac{b}{p_{\text{mega}}} \right)^2 + (1 - p_{\text{mega}}) \right) \|h_i^t - \nabla f_i(x^t)\|^2.
\end{aligned}$$

901 Using (34) and (35), we get

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

902 Next, due to Assumption 3, we obtain

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

903 The third inequality can be proved with the help of (35) and Assumption 3.

$$\begin{aligned}
& \mathbb{E}_k \left[\|k_{i,2}^{t+1}\|^2 \right] \\
& \stackrel{(14)}{=} \mathbb{E}_k \left[\|k_{i,2}^{t+1} - \mathbb{E}_k[k_{i,2}^{t+1}]\|^2 \right] + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& \leq \frac{L_\sigma^2}{B} \|x^{t+1} - x^t\|^2 + \|\nabla f_i(x^{t+1}) - \nabla f_i(x^t)\|^2 \\
& \leq \left(\frac{L_\sigma^2}{B} + L_i^2 \right) \|x^{t+1} - x^t\|^2.
\end{aligned}$$

904

□

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 $B' \geq B \geq 1$

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

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

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

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

906 *Proof.* Due to Lemma 2 and the update step from Line 5 in Algorithm 8, we have

$$\begin{aligned} & \mathbb{E}_{t+1} [f(x^{t+1})] \\ & \leq \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - \nabla f(x^t)\|^2 \right] \\ & = \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \frac{\gamma}{2} \|g^t - h^t + h^t - \nabla f(x^t)\|^2 \right] \\ & \stackrel{(14)}{\leq} \mathbb{E}_{t+1} \left[f(x^t) - \frac{\gamma}{2} \|\nabla f(x^t)\|^2 - \left(\frac{1}{2\gamma} - \frac{L}{2} \right) \|x^{t+1} - x^t\|^2 + \gamma \left(\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2 \right) \right]. \end{aligned}$$

907 Let us fix constants $\kappa, \eta, \nu, \rho \in [0, \infty)$ that we will define later. Considering Lemma 13, Lemma 14,
908 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 \left(\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2 \right) \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 \left(\|g^t - h^t\|^2 + \|h^t - \nabla f(x^t)\|^2 \right) \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

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

909 Thus

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

910 After rearranging the terms, we get

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

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

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

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

912 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 $\eta =$

913 $\frac{\gamma((2\omega+1)p_a - p_{aa})}{np_a^2}$, 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}$$

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

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

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

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

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

916 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

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

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

917 Using Lemma 4 and the assumption about γ , we get

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

918 It is left to apply Lemma 3 with

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

919 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

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

920 communication rounds to get an ε -solution, the expected communication complexity is equal to
921 $\mathcal{O}(d + \zeta_c T)$, and the expected number of stochastic gradient calculations per node equals $\mathcal{O}(B_{\text{init}} +$
922 $BT)$, where ζ_c is the expected density from Definition 12.

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

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

924 Using

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

925 and

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

926 we have

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

927 Therefore, we can take the following T to get ε -solution.

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

928 Considering the choice of p_{mega} and B_{init} , we obtain

$$\begin{aligned} T &= \mathcal{O} \left(\frac{1}{\varepsilon} \left[\Delta_0 \left(L + \left(\frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_C n}} \right) \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left(\frac{\hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right) + \frac{\sigma^2}{np_{\text{mega}}p_a B_{\text{init}}} \right] \right) \\ &= \mathcal{O} \left(\frac{\Delta_0}{\varepsilon} \left[L + \left(\frac{\omega}{p_a \sqrt{n}} + \sqrt{\frac{d}{p_a^2 \zeta_C n}} \right) \left(\hat{L} + \frac{L_\sigma}{\sqrt{B}} \right) + \frac{\sigma}{p_a \sqrt{\varepsilon n}} \left(\frac{\hat{L}}{\sqrt{B}} + \frac{L_\sigma}{B} \right) \right] + \frac{\sigma^2}{\sqrt{p_a} n \varepsilon B} \right). \end{aligned}$$

929 The expected communication complexity equals $\mathcal{O}(d + p_{\text{mega}}d + (1 - p_{\text{mega}})\zeta_C) =$
 930 $\mathcal{O}(d + \zeta_C)$ and the expected number of stochastic gradient calculations per node equals
 931 $\mathcal{O}(B_{\text{init}} + p_{\text{mega}}B' + (1 - p_{\text{mega}})B) = \mathcal{O}(B_{\text{init}} + B)$. \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 $B' \geq B \geq 1$,

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

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

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

933 *Proof.* Let us fix constants $\kappa, \eta, \nu, \rho \in [0, \infty)$ that we will define later. As in the proof of Theorem 11,
934 we can get

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

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

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

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

936 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 $\eta =$
937 $\frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2}$, 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

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

938 where simplified the term using $p_{aa} \geq 0$. Let us take $\nu = \frac{2\gamma}{b}$ to obtain

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

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

939 Next, we take $\rho = \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}}$, thus

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[\|g^{t+1} - h^{t+1}\|^2 \right] + \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} \left[\|h^{t+1} - \nabla f(x^{t+1})\|^2 \right] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} \left[\|\nabla f(x^t)\|^2 \right] \\
& - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega+1)\omega}{np_a^2} \left(\frac{L_\sigma^2}{B} + \widehat{L}^2 \right) \right. \\
& \quad \left. - \left(\frac{4\gamma L_\sigma^2}{bn p_a B} + \frac{4\gamma(p_a - p_{aa})\widehat{L}^2}{bn p_a^2} \right) - \left(\frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \right) \left(\frac{2L_\sigma^2}{p_a B} + \frac{2(1-p_a)\widehat{L}^2}{p_a} \right) \right) \mathbb{E} \left[\|x^{t+1} - x^t\|^2 \right] \\
& + \left(1 - \frac{a}{2} \right) \frac{2\gamma(2\omega+1)}{p_a} \mathbb{E} \left[\|g^t - h^t\|^2 \right] + \left(1 - \frac{a}{2} \right) \frac{2\gamma((2\omega+1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2} \right) \frac{2\gamma}{b} \mathbb{E} \left[\|h^t - \nabla f(x^t)\|^2 \right] + \left(1 - \frac{b}{2} \right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \left(\frac{4\gamma b}{np_{\text{mega}} p_a} + \frac{16\gamma(p_a - p_{aa})b^2}{np_a^3 p_{\text{mega}}^2} \right) \frac{\sigma^2}{B'},
\end{aligned}$$

940 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

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

$$\begin{aligned}
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'} \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& - \left(\frac{1}{2\gamma} - \frac{L}{2} - \frac{8\gamma(2\omega + 1)\omega}{np_a^2} \left(\frac{L_\sigma^2}{B} + \widehat{L}^2 \right) - \left(\frac{24\gamma L_\sigma^2}{np_{\text{mega}} p_a^2 B} + \frac{24\gamma \left(1 - \frac{p_{aa}}{p_a}\right) \widehat{L}^2}{np_{\text{mega}} p_a^2} \right) \right) \mathbb{E} [\|x^{t+1} - x^t\|^2] \\
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'}.
\end{aligned}$$

941 Using Lemma 4 and the assumption about γ , we get

$$\begin{aligned}
& \mathbb{E} [f(x^{t+1})] + \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^{t+1} - h^{t+1}\|^2] + \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^{t+1} - h_i^{t+1}\|^2 \right] \\
& + \frac{2\gamma}{b} \mathbb{E} [\|h^{t+1} - \nabla f(x^{t+1})\|^2] + \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^{t+1} - \nabla f_i(x^{t+1})\|^2 \right] \\
& \leq \mathbb{E} [f(x^t)] - \frac{\gamma}{2} \mathbb{E} [\|\nabla f(x^t)\|^2] \\
& + \left(1 - \frac{a}{2}\right) \frac{2\gamma(2\omega + 1)}{p_a} \mathbb{E} [\|g^t - h^t\|^2] + \left(1 - \frac{a}{2}\right) \frac{2\gamma((2\omega + 1)p_a - p_{aa})}{np_a^2} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|g_i^t - h_i^t\|^2 \right] \\
& + \left(1 - \frac{b}{2}\right) \frac{2\gamma}{b} \mathbb{E} [\|h^t - \nabla f(x^t)\|^2] + \left(1 - \frac{b}{2}\right) \frac{8\gamma(p_a - p_{aa})}{np_a^2 p_{\text{mega}}} \mathbb{E} \left[\frac{1}{n} \sum_{i=1}^n \|h_i^t - \nabla f_i(x^t)\|^2 \right] \\
& + \frac{20\gamma\sigma^2}{nB'}.
\end{aligned}$$

942 Due to $\gamma \leq \frac{a}{2\mu}$ and $\gamma \leq \frac{b}{2\mu}$, we have

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

943 It is left to apply Lemma 11 with

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

944 and $C = \frac{20\sigma^2}{nB'}$ to conclude the proof. □