

# Low Probability of Initiating *nirS* Transcription Explains Observed Gas Kinetics and Growth of Bacteria Switching from Aerobic Respiration to Denitrification



Junaid Hassan<sup>1</sup>, Linda L. Bergaust<sup>1</sup>, I. David Wheat<sup>2</sup>, Lars R. Bakken<sup>1</sup>\*

1 Department of Environmental Sciences, Norwegian University of Life Sciences, Ås, Norway, 2 Department of Geography, University of Bergen, Bergen, Norway

#### **Abstract**

In response to impending anoxic conditions, denitrifying bacteria sustain respiratory metabolism by producing enzymes for reducing nitrogen oxygnions/-oxides ( $NO_v$ ) to  $N_2$  (denitrification). Since denitrifying bacteria are non-fermentative, the initial production of denitrification proteome depends on energy from aerobic respiration. Thus, if a cell fails to synthesise a minimum of denitrification proteome before O<sub>2</sub> is completely exhausted, it will be unable to produce it later due to energylimitation. Such entrapment in anoxia is recently claimed to be a major phenomenon in batch cultures of the model organism Paracoccus denitrificans on the basis of measured e<sup>-</sup>-flow rates to O<sub>2</sub> and NO<sub>x</sub>. Here we constructed a dynamic model and explicitly simulated actual kinetics of recruitment of the cells to denitrification to directly and more accurately estimate the recruited fraction ( $F_{den}$ ). Transcription of nirS is pivotal for denitrification, for it triggers a cascade of events leading to the synthesis of a full-fledged denitrification proteome. The model is based on the hypothesis that nirS has a low probability ( $\mathbf{r}_{den}$ ,  $h^{-1}$ ) of initial transcription, but once initiated, the transcription is greatly enhanced through positive feedback by NO, resulting in the recruitment of the transcribing cell to denitrification. We assume that the recruitment is initiated as [O<sub>3</sub>] falls below a critical threshold and terminates (assuming energy-limitation) as  $[O_2]$  exhausts. With  $\mathbf{r}_{\text{den}} = 0.005 \text{ h}^{-1}$ , the model robustly simulates observed denitrification kinetics for a range of culture conditions. The resulting  $F_{den}$  (fraction of the cells recruited to denitrification) falls within 0.038-0.161. In contrast, if the recruitment of the entire population is assumed, the simulated denitrification kinetics deviate grossly from those observed. The phenomenon can be understood as a 'bet-hedging strategy': switching to denitrification is a gain if anoxic spell lasts long but is a waste of energy if anoxia turns out to be a 'false alarm'.

Citation: Hassan J, Bergaust LL, Wheat ID, Bakken LR (2014) Low Probability of Initiating *nirS* Transcription Explains Observed Gas Kinetics and Growth of Bacteria Switching from Aerobic Respiration to Denitrification. PLoS Comput Biol 10(11): e1003933. doi:10.1371/journal.pcbi.1003933

Editor: Robinson Fulweiler, Boston University, United States of America

Received April 17, 2014; Accepted September 17, 2014; Published November 6, 2014

**Copyright:** © 2014 Hassan et al. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

**Data Availability:** The authors confirm that all data underlying the findings are fully available without restriction. All relevant data are within the paper and its Supporting Information files.

**Funding:** This work is funded by Norwegian University of Life Sciences, financed by the Ministry of Education and Research, Norway. No institution has, in any way, influenced the outcome of this work. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing Interests: The authors have declared that no competing interests exist.

\* Email: lars.bakken@nmbu.no

### Introduction

A complete denitrification pathway includes the dissimilatory reduction of nitrate ( $NO_3^-$ ) through nitrite ( $NO_2^-$ ), nitric oxide (NO), and nitrous oxide ( $N_2O$ ) to di-nitrogen ( $N_2$ ). Typically, the genes encoding reductases for these nitrogen oxyanions/-oxides ( $NO_x$ ) are not expressed constitutively but only in response to  $O_2$  depletion, making denitrification a facultative trait [1]. Hence, during anoxic spells, the process enables denitrifying bacteria to sustain respiratory metabolism, replacing  $O_2$  by  $NO_x$  as the terminal electron (e<sup>-</sup>) acceptors. Since permanently anoxic environments lack available  $NO_x$ , denitrification is confined to sites where  $O_2$  concentration fluctuates, such as biofilms, surface layers of sediments, and drained soil (which turns anoxic in response to flooding).

### From modelling denitrifying communities as a homogenous unit to a model of regulation of denitrification in an individual strain

Denitrification is a key process in the global nitrogen cycle and is also a major source of atmospheric  $N_2O$  [2]. A plethora of

biogeochemical models have been developed for understanding the ecosystem controls of denitrification and N<sub>2</sub>O emissions [3]. A common feature of these models is that the denitrifying community of the system (primarily soils and sediments) in question is treated as one homogenous unit with certain characteristic responses to O<sub>2</sub> and NO<sub>3</sub><sup>-</sup> concentrations. This simplification is fully legitimate from a pragmatic point of view, but in reality any denitrifying community is composed of a mixture of organisms with widely different denitrification regulatory phenotypes [4]. Modelling has been used to a limited extent to analyse kinetic data for various phenotypes (See [5] and references therein) and for understanding the accumulation of intermediates [6]. To our knowledge, however, no attempts have been made to model the regulation during transition from aerobic to anaerobic respiration in individual strains, despite considerable progress in the understanding of their regulatory networks. It would be well worth the effort, since the regulatory phenomena at the cellular level provide clues as to how denitrification and NO and N2O emissions therefrom are regulated in intact soils [7]. Explicit modelling of the entire denitrification regulatory network,

### **Author Summary**

In response to oxygen-limiting conditions, denitrifying bacteria produce a set of enzymes to convert  $NO_3^-/NO_2^$ to N<sub>2</sub> via NO and N<sub>2</sub>O. The process (denitrification) helps generate energy for survival and growth during anoxia. Denitrification is imperative for the nitrogen cycle and has far-reaching consequences including contribution to global warming and destruction of stratospheric ozone. Recent experiments provide circumstantial evidence for a previously unknown phenomenon in the model denitrifying bacterium Paracoccus denitrificans: as O2 depletes, only a marginal fraction of its population appears to switch to denitrification. We hypothesise that the low success rate is due to a) low probability for the cells to initiate the transcription of genes (nirS) encoding a key denitrification enzyme (NirS), and b) a limited time-window in which NirS must be produced. Based on this hypothesis, we constructed a dynamic model of denitrification in Pa. denitrificans. The simulation results show that, within the limited time available, a probability of 0.005 h<sup>-1</sup> for each cell to initiate nirS transcription (resulting in the recruitment of 3.8-16.1% cells to denitrification) is sufficient to adequately simulate experimental data. The result challenges conventional outlook on the regulation of denitrification in general and that of Pa. denitrificans in particular.

however, would take us beyond available experimental evidence, with numerous parameters for which there are no empirical values. Considering this limitation, here we have constructed a simplified model to investigate if a stochastic transcriptional initiation of key denitrification genes (*nirS*) could possibly explain peculiar kinetics of e<sup>-</sup>-flow as *Paracoccus denitrificans* switch from aerobic to anaerobic respiration [4,8].

Although denitrification is widespread among bacteria, the αproteobacterium Pa. denitrificans is the 'paradigm' model organism in denitrification research. Recent studies [4,8,9] have indicated a previously unknown phenomenon in this species that, in response to O<sub>2</sub> depletion, only a marginal fraction (F<sub>den</sub>) of its entire population appears to successfully switch to denitrification. In these studies, however, Fden is inferred from rates of consumption and production of gases (O2, NOx, and N2), and a clear hypothesis as to the underlying cause of the low F<sub>den</sub> is also lacking. To fill these gaps, we formulated a refined hypothesis addressing the underlying regulatory mechanism of the cell differentiation in response to O2 depletion. On its basis, we constructed a dynamic model and explicitly simulated the actual kinetics of recruitment of the cells from aerobic respiration to denitrification. The model adequately matches batch cultivation data for a range of experimental conditions [4,8] and provides a direct and refined estimation of  $F_{den}$ . The exercise is important for understanding the physiology of denitrification in general and of Pa. denitrificans in particular and carries important implications for correctly interpreting various denitrification experiments.

## Regulation of denitrification in terms of relevance to fitness

Generally, the transcription of genes encoding denitrification enzymes is inactivated in the presence of  $O_2$ . A population undertaking denitrification typically responds to full aeration by completely shutting down denitrification and immediately initiating aerobic respiration [10]. Thus,  $O_2$  controls denitrification at transcriptional as well as metabolic level, and both have a plausible

fitness value. The transcriptional control minimises the energy cost of producing denitrification enzymes, and the metabolic control maximises ATP (per mole electrons transferred) because the mole ATP per mole electrons transferred to the terminal e $^-$ -acceptor is  $\sim 50\%$  higher for aerobic respiration than for denitrification [10].

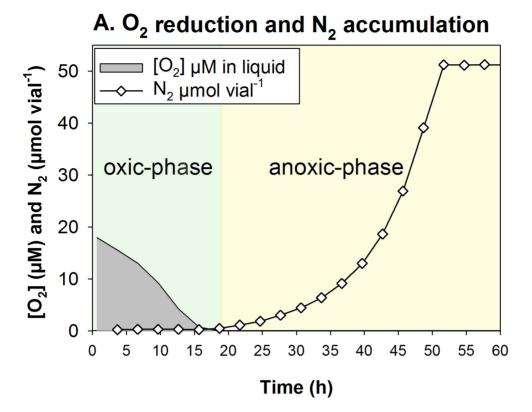
Denitrification enzymes produced in response to an anoxic spell are likely to linger within the cells under subsequent oxic conditions (although, this has not been studied in detail), ready to be used if O<sub>2</sub> should become limiting later on. However, these enzymes will be diluted by aerobic growth, since the transcription of their genes is effectively inactivated by  $O_2$ . Hence, a population growing through many generations under fully oxic conditions will probably be dominated by the cells without intact denitrification proteome. When confronted with  $O_2$  depletion, such a population will have to start from scratch, i.e., transcribe the relevant genes, translate mRNA into peptide chains (protein synthesis by ribosomes) and secure that these chains are correctly folded by the chaperones, transport the enzymes to their correct locations in the cell, and insert necessary co-factors (e.g., Cu, Fe, or Mo). In E. coli grown under optimal conditions, the whole process from the transcriptional activation to a functional enzyme takes ≤20 minutes [11] and costs significant amount of energy (ATP).

Synthesis of denitrification enzymes is rewarding if anoxia lasts long and  $\mathrm{NO}_{\mathrm{x}}$  remains available, but it is a waste of energy if anoxia is brief. Since the organisms cannot sense how long an impending anoxic spell will last, a 'bet-hedging strategy' [12] where one fraction of a population synthesises denitrification enzymes while the other does not may increase overall fitness.

# A delayed response to $O_2$ depletion may lead to entrapment in anoxia

Most, if not all, denitrifying bacteria are non-fermentative and completely rely on respiration to generate energy [13,14]. This implies that their metabolic machinery will run out of energy whenever deprived of terminal e-acceptors. When [O<sub>2</sub>] falls below some critical threshold, the cells will 'sense' this and start synthesising denitrification proteome, utilising energy from aerobic respiration [10]. However, if O2 is suddenly exhausted or removed, the lack of a terminal e-acceptor will create energy limitation, restraining the cells from enzyme synthesis, hence, entrapping them in anoxia. This was clearly demonstrated by Højberg et al. [15], who used silicone immobilised cells to transfer them from a completely oxic to a completely anoxic environment. Such a rapid transition is unlikely to occur in nature; however, the experiment illustrates one of the apparent perils in the regulation of denitrification: the cells that respond too late to O2 depletion will be entrapped in anoxia, unable to utilise alternative electron acceptors for energy conservation and growth.

Højberg et al.'s [15] observations have largely been ignored in the research on the regulation of denitrification, and it is implicitly assumed that, in response to O2 depletion, all cells in cultures of denitrifying bacteria will switch to denitrification. Contrary to this, however, Bergaust et al. [4,8,16] followed by Nadeem et al. [9] proposed that in batch cultures of *Pa. denitrificans*, only a small fraction of all cells is able to switch to denitrification. During transition from oxic to anoxic conditions, they observed a severe depression in the total e<sup>-</sup>-flow rate (i.e., to O<sub>2</sub>+NO<sub>x</sub>, see Fig. 1), which was estimated on the basis of measured gas kinetics. Had all of the cells switched to denitrification as  $O_2$  exhausted, the total e-flow rate would have carried on increasing, without such a depression. The depression was followed by an exponential increase in the e-flow rate, which was tentatively ascribed to anaerobic growth of a small F<sub>den</sub> (fraction recruited to denitrification). It was postulated that this fraction escaped entrapment in



# B. Calculation of relative growth rates ( $\mu_{anoxic}$ : $\mu_{oxic}$ )

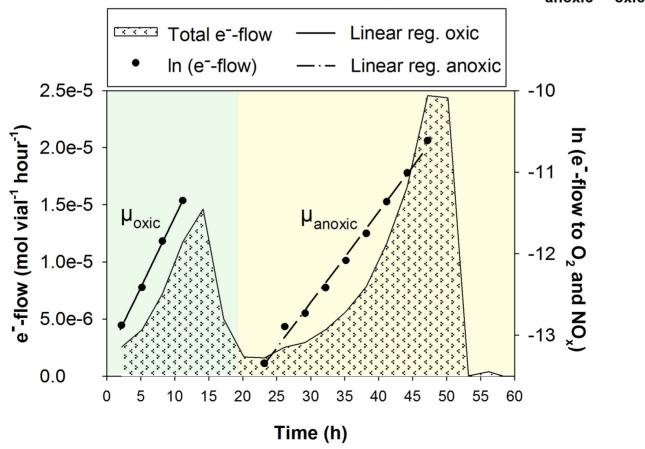


Figure 1. Data generated by batch cultivation of Pa. denitrificans [4] (redrawn). As the cells transited from oxic to anoxic conditions (Panel A), Bergaust et al. [4] observed a severe depression in the total  $e^-$ -flow rate (i.e., to  $O_2+NO_{xv}$ , Panel B), which was taken to indicate that only a fraction of the cells switched to anaerobic respiration (denitrification). Had all of the cells switched, the total  $e^-$ -flow would have carried on increasing without such a depression. The depression was followed by an exponential increase in the  $e^-$ -flow rate, which was ascribed to anaerobic growth of a small fraction ( $F_{den}$ ) of the cells that escaped entrapment in anoxia and carried on growing by denitrification. doi:10.1371/journal.pcbi.1003933.g001

anoxia by synthesising initial denitrification proteins within the time-window when  $\mathrm{O}_2$  was still present, whereas the majority of the cells  $(1-F_{den})$  failed to do so, thus remained unable to utilise  $\mathrm{NO}_x$ .

The core hypothesis: A low probability of initiating *nirS* transcription seems to drive the cell differentiation

**Autocatalytic transcription of denitrification genes.** In Pa. denitrificans, denitrification is driven by four core enzymes: Nar (membrane-bound nitrate reductase), NirS (cytochrome  $cd_1$  nitrite reductase), cNor (nitric oxide reductase), and NosZ (nitrous oxide reductase, see Fig. 2). The transcriptional regulation of genes encoding these enzymes (nar, nirS, nor and nosZ, respectively) involves, at least, three FNR-type proteins acting as sensors for  $O_2$  (FnrP),  $NO_3^-/NO_2^-$  (NarR), and NO (NNR) [10,17,18]. NarR and NNR facilitate product-induced transcription of the nar and nirS genes: When anoxia is imminent, the low  $[O_2]$  is sensed by FnrP, which in interplay with NarR induces nar transcription. NarR is activated by  $NO_2^-$  (and/or probably by

NO<sub>3</sub><sup>-</sup>); thus once a cell starts producing traces of NO<sub>2</sub><sup>-</sup>, *nar* expression becomes autocatalytic. The transcription of *nirS* is induced by NNR, which requires NO for activation; thus once traces of NO are produced, the expression of *nirS* also becomes autocatalytic. In contrast, the transcription of *nor* is substrate (NO) induced via NNR, while *nosZ* is equally but independently induced by NNR and FnrP [19]. Here we are concerned with the dynamics that start with the transcription of *nirS*, since the experimental treatments that we simulated were not supplemented with NO<sub>3</sub><sup>-</sup> but various concentrations of NO<sub>2</sub><sup>-</sup> only (Table 1).

Low probability of initiating *nirS* transcription. The transcription of nirS is known to be suppressed by  $O_2$  [4,8], but the exact mechanism remains unclear. Circumstantial evidence suggests that it is due to  $O_2$  inactivating NNR [20] (dashed link in Fig. 2), but this is not necessary to explain the repression of NirS. There are several mechanisms through which high  $O_2$  concentrations may restrain NirS activity, i.e., through post-transcriptional regulation, direct interaction with the enzyme, or due to competition for electrons. Regardless of the exact mechanism(s),

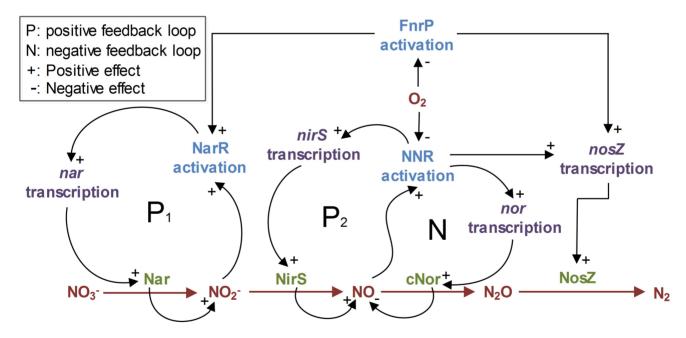


Figure 2. The regulatory network of denitrification in Pa. denitrificans. In Pa. denitrificans, denitrification is driven by four core enzymes: Nar (nitrate reductase encoded by the nar genes), NirS (nitrite reductase encoded by nirS), cNor (NO reductase encoded by nor), and NosZ (N<sub>2</sub>O reductase encoded by nosZ). The transcription of these genes is regulated by, at least, three FNR-type proteins, which are sensors for  $O_2$  (FnrP),  $NO_3^-$  /N $O_2^-$  (NarR), and NO (NNR). NarR and NNR facilitate product-induced transcription of the nar and nirS genes (see positive-feedback loops), where NNR also counteracts the NO accumulation (negative-feedback loop) [10,17,18]. Circumstantial evidence suggests that  $O_2$  inactivates NNR (grey dashed link) [20], and NirS is also unlikely to be functional in the presence of high  $O_2$  concentrations. Hence, for our modelling we hypothesise that the probability of an autocatalytic transcriptional activation of nirS is zero until  $O_2$  falls below a critical concentration  $O_2$  ( $O_2$ ) trigger. When  $O_2$  falls below  $O_2$  f

**Table 1.** The simulated experiment of Bergaust et al [4,8].

| Batch No. | $O_{2\mathrm{HS}}$ $(t_0$ $)$ (vol. %)* | $\mathbf{NO_2}^ (\mathbf{t}_0$ $)$ (mM) |
|-----------|---|---|
| 1         | ~0                                      | 0.2                                     |
| 2         | ~0                                      | 1                                       |
| 3         | ~0                                      | 2                                       |
| 4         | 1                                       | 0.2                                     |
| 5         | 1                                       | 1                                       |
| 6         | 1                                       | 2                                       |
| 7         | 7                                       | 0.2                                     |
| 8         | 7                                       | 1                                       |
| 9         | 7                                       | 2                                       |

\*Targeted values for initial  $O_2$  in the headspace (where the headspace vol. = 70 mL). The actual initial  $O_2$  measured in the 0, 1, and 7% treatments was 0.012–0.19, 1.2–1.66, 6.6–6.8 vol.%, respectively. The  $O_2$  present in the  $\sim$ 0% treatments was due to traces of  $O_2$  left behind despite various cycles of evacuation of the headspace air and subsequent flushing of the vials with helium (He-washing).

doi:10.1371/journal.pcbi.1003933.t001

the ultimate consequence is the elimination of the positive feedback via NO and NNR. When  $O_2$  falls below a critical threshold, facilitating NirS activity, this positive feedback would allow the product of a single transcript of nirS to induce a subsequent burst of nirS transcription in response to NO. Such 'switches' in gene expression by positive-feedback loops are not uncommon in prokaryotes, and they have been found to result in cell differentiation because the initial transcription is stochastic with a relatively low probability [21].

Our model assumes such stochastic recruitment to denitrification, triggered by an initial nirS transcription occurring with a low probability. This initial transcription is possibly mediated by a minute pool of intact NNR and/or through crosstalk with other factors, such as FnrP. A  $NO_2^-$ -supplemented medium contains non-biologically formed traces of NO which, once diffused into the cells while  $O_2$  is low, will activate background levels of NNR and, thereby, may also increase the probability of triggering nirS transcription.

For this modelling exercise, we do not need a full clarification of the mechanisms involved but only to assume that the probability of an autocatalytic transcriptional activation of nirS would be practically zero as long as  $O_2$  concentration is above a certain threshold. This assumption is backed by empirical data indicating that NO is not produced to detectable levels before  $O_2$  concentration falls below a critical threshold [8,22]. For  $O_2$  concentrations below this threshold, the model assumes a low (but unknown) probability for each cell to initiate the autocatalytic transcription of nirS, paving the way for the rest of the denitrification proteome.

 $O_2$  is required for the initial production of NirS. We further assume that the recruitment to denitrification will only be possible as long as a minimum of  $O_2$  is available because the synthesis of first molecules of NirS will depend on energy from aerobic respiration.

Can NO produced within one cell help activate the autocatalytic transcription of nirS in the neighbouring cells? It is perhaps less obvious that the autocatalytic transcriptional activation of nirS takes place only within the NO-producing cell because NO diffuses easily across membranes [23]. However, the average distance between the cells in a culture with  $10^9$  cells mL<sup>-1</sup> (roughly the numbers that we are dealing

with) is  $\sim\!10$   $\mu m,$  which is  $\sim\!10$  times the diameter of a cell. This implies that an NO molecule produced by a cell has a much higher probability to react with and activate the NNR inside the same cell than to do so in another one.

### Modelling the cell differentiation

To represent the batch cultivation conducted by Bergaust et al. [4,8], the model explicitly simulates growth of two sub-populations, one with denitrification enzymes  $(N_{D+})$  and the other without  $(N_{D-})$ ; both equally consume  $O_2$ , but  $N_{D-}$  cannot reduce  $NO_x$  to  $N_2$ . Once oxygen concentration in the liquid ( $[O_2]_{LP}$ ) falls below a critical level  $([O_2]_{trigger})$  [22], the cells within  $N_{D-}$  are assumed to initiate nirS transcription (and thereby ensure recruitment to N<sub>D+</sub>) with a rate described by a probabilistic function:  $N_{D-} \times r_{den}(O_2)$  (cells  $h^{-1}$ ), where  $r_{den}(O_2)$  is assumed to be an  $[O_2]_{LP}$  dependent probability (h<sup>-1</sup>) for any cell within  $N_{D-}$ to initiate nirS transcription (leading to a full denitrification capacity). When  $[O_2]_{LP}$  falls below  $[O_2]_{trigger}$ ,  $r_{den}(O_2)$  triggers and holds a constant value as long as  $\left[O_2\right]_{LP}$  is above a critical minimum ( $[O_2]_{min}$ ). For  $[O_2]_{LP} > [O_2]_{trigger}$ ,  $r_{den}(O_2)$  is zero (assuming the inactivation of NNR by  $O_2$ );  $r_{den}(O_2)$  is also zero for  $\left[O_{2}\right]_{LP} < \left[O_{2}\right]_{min}$  (assuming the lack of energy for protein synthesis).

The recruitment of  $N_{D-}$  to  $N_{D+}$  is simulated as an instantaneous event; thus, the model does not take into account the time-lag between the initiation of *nirS* transcription and the time when the transcribing cell has become a fully functional denitrifier. This simplification is based on the evidence that this lag is rather short. Experiments with *E. coli* [11] under optimal conditions suggest lags of ~20 minutes between the onset of transcription and the emergence of a functional enzyme. In *Pa. denitrificans* [8,22], the lag observed between the emergence of denitrification gene transcripts and the subsequent gas products suggests that the time required for synthesising the enzymes is within the same range.

# Employing the model to understand 'diauxic lags' between the aerobic and anaerobic growth-phases

In a series of experiments with denitrifying bacteria (Pseudomonas denitrificans, Pseudomonas fluorescens, Alcaligenes eutrophus and Paracoccus pantotrophus) [24–26], oxic cultures were sparged with  $N_2$  to remove  $O_2$  and were monitored by measuring optical density (OD<sub>550</sub>). All the strains except Ps. fluorescens went through a conspicuous 'diauxic lag: a period of little or no growth' [26]; the OD remained practically constant during the lag period, lasting 4–30 hours, which was eventually followed by anaerobic growth.

To understand the diauxic lag, Liu et al. [24] used the common assumption that all cells would eventually switch to denitrification. They constructed a simulation model based on the assumption that all the cells contained a minimum of denitrification proteome (even after many generations under oxic conditions). This minimum would allow them to produce more denitrification enzymes when deprived of  $O_2$ , albeit very slowly due to energy limitation. The time taken to effectively produce adequate amounts of denitrification enzymes (= the diauxic lag) was taken to be a function of the initial amounts of these enzymes per cell. Although their model may possibly explain short time-lags, it appears unrealistic for lag phases as long as 10-30 hours [25] because to produce such long lags, conceivably, the initial enzyme concentration would be less than one enzyme molecule per cell, which is mathematically possible but biologically meaningless.

The model presented in this paper provides an alternative explanation for the apparent diauxic lags: a sudden shift from fully oxic to near anoxic conditions (by sparging with N<sub>2</sub>) would leave the medium with only traces of O2, which would be quickly depleted due to aerobic respiration. As a consequence, the available time for initiating the synthesis of denitrification proteome would be marginal, allowing only a tiny fraction (F<sub>den</sub>) of the cells to switch to denitrification. This marginal fraction would grow exponentially from the very onset of anoxic conditions, but it would remain practically undetectable as measured (OD) for a long time, creating the apparent 4-30 h lag. The length of the lag depends on the fraction of the cells switching to denitrification. To demonstrate this alternative explanation, we adjusted our model to the reported conditions and simulated the experiment of Liu et al [24]. The model produced qualitatively similar 'diauxic lags' in the simulated cell density (OD), although the time length of the lag could be anything (depending on assumptions regarding the residual  $O_2$  after sparging, which was not measured).

#### **Materials and Methods**

# An overview of the modelled experiment: Batch incubations in gas-tight vials

Bergaust *et al.* [4,8] studied aerobic and anaerobic respiration rates in *Paracoccus denitrificans* (DSM413). The cells were incubated (at  $20^{\circ}$ C) as stirred batches in 120 mL gastight vials, containing 50 mL Sistrom's medium [27] (Fig. 3). The medium was supplemented with various concentrations of KNO<sub>3</sub> or KNO<sub>2</sub>. Prior to inoculation, air in the headspace was replaced with He to remove O<sub>2</sub> and N<sub>2</sub> (He-washing), followed by the injection of no, 1, or 7 headspace-vol.% O<sub>2</sub>. Finally, each vial was inoculated with  $\sim 3 \times 10^8$  aerobically grown cells.

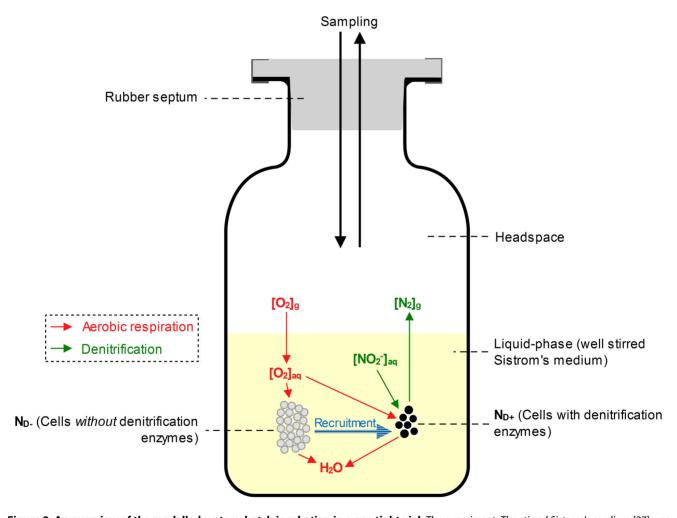


Figure 3. An overview of the modelled system: batch incubation in a gas-tight vial. The experiment: The stirred Sistrom's medium [27] was inoculated with aerobically grown Pa. denitrificans cells, which were provided with different concentrations of  $O_2$  and  $NO_2^-$  (g or aq with a chemical species-name represents gaseous or aqueous, respectively).  $O_2$  is consumed by respiration, driving its transport from the headspace to the liquid. Once the aerobic respiration becomes limited, the cells may switch to denitrification (recruitment), reducing  $NO_2^-$  to  $N_2$  via the intermediates NO and  $N_2O$  (not shown). For monitoring  $O_2$ ,  $CO_2$ ,  $N_2$ , NO and  $N_2O$ , a robotised incubation system [28] was used, which automatically takes samples from the headspace by piercing the rubber septum. Each sampling removes a fraction (3–3.4%) of all gases in the headspace, but it also involves a marginal leakage of  $O_2$  and  $N_2$  into the vial (as indicated by the two-way arrows at the top of the figure). The model: The model operates with two sub-populations: one without and the other with denitrification enzymes ( $N_{D-}$  and  $N_{D+}$ , respectively). Both consume  $O_2$  if present, but  $N_{D-}$  cannot reduce  $NO_x$ . The  $N_{D-}$  cells may be recruited to the  $N_{D+}$  pool as  $[O_2]_{aq}$  falls below a critical threshold. The rate of recruitment ( $R_{rec}$ ) is modelled as a probabilistic function:  $R_{rec} = N_{D-} \times r_{den}(O_2)$  (cells  $h^{-1}$ ), where  $r_{den}(O_2)$  represents an  $O_2$  dependent specific-probability ( $h^{-1}$ ) for any  $N_{D-}$  cell to initiate nirS transcription (leading to the synthesis of a full-fledged denitrification proteome).

**Treatments selected for simulation.** Only  $NO_2^-$ -supplemented treatments (Table 1) were selected for this modelling exercise for two reasons. First,  $NO_2^-$  was not monitored; hence, results of the  $NO_3^-$ -supplemented treatments could not provide exact estimates of anaerobic respiration rates (due to an unknown transient accumulation of  $NO_2^-$ ). Second, by excluding the treatments requiring Nar, we could single out and focus on the regulation of the other key enzyme NirS.

Aerobic respiration followed by denitrification.  $O_2$  diffused from the headspace to the liquid (Fig. 3), where the cells consumed it before switching to denitrification: the stepwise reduction of  $NO_2^-$  to  $N_2$  via the intermediates NO and  $N_2O$  (not shown). Headspace concentrations of gases were monitored by frequent sampling (every 3 hours). A typical result is shown in Fig. 1A, illustrating the increasing rate of  $O_2$  consumption until depletion, followed by transition to denitrification. The denitrification rate increased exponentially till all the  $NO_2^-$  present in the medium was recovered as  $N_2$ . The medium contained ample amounts of carbon substrate (34 mM succinate) to support the consumption of all available electron acceptors.

**Sampling procedure.** To monitor  $O_2$ ,  $CO_2$ , NO,  $N_2O$ , and  $N_2$  in the headspace for respiring cultures, Bergaust *et al.* [4,8] used a robotised incubation system, which automatically takes samples from the headspace by piercing the rubber septum (Fig. 3). The auto-sampler is connected to a gas chromatograph (GC) and an NO analyser (For details, see [28]). The system uses peristaltic pumping, which removes a fraction (3–3.4%) of all the gases in the headspace and then reverses the pumping to inject an equal amount of He into the headspace, thus maintaining  $\sim$ 1 atmosphere pressure inside the vial. Sampling also involves a marginal leakage of  $O_2$  and  $N_2$  into the headspace ( $\sim$ 22 and  $\sim$ 60 nmol per sampling, respectively) through tubing and membranes of the injection system.

**Calculation of gases in the liquid.** Concentrations of gases in the liquid were calculated using solubility of each gas at the given temperature  $(20^{\circ}\text{C})$ , assuming equilibrium between the headspace and the liquid. However, the  $O_2$  consumption rate was so high that to calculate  $[O_2]$  in the liquid, its transport rate (from the headspace to the liquid) had to be taken into account.

### An overview of the model

The model effectively represents the physical phenomena mentioned above, so as to ensure that the simulation results match the measured data for the right reasons. Net effect of sampling (dilution and leakage) is included in the simulation of O<sub>2</sub> kinetics at the reported sampling times. Transport of O<sub>2</sub> between the headspace and the liquid is modelled using an empirically determined transport coefficient and the solubility of O<sub>2</sub> in water at 20°C. To simulate the metabolic activity (O<sub>2</sub> consumption and  $N_2$  production) and growth, the model divides the cells into two sub-populations: one without and the other with denitrification enzymes ( $N_{D-}$  and  $N_{D+}$  pools, respectively, see Fig. 3). Both equally consume  $O_2$  if present, but  $N_{D-}$  cannot reduce  $NO_2^-$  to  $N_2$ . Those  $N_{D-}$  cells that, in response to  $O_2$  depletion, are able to initiate nirS transcription (see Fig. 2) are recruited to the  $N_{D+}$ pool, where  $N_{D+} = 0$  prior to the recruitment. The recruitment rate (R<sub>rec</sub>) is modelled according to a probabilistic function described below (Eqs. 7-8).

The model ignores sampling effect on  $N_2$  (leakage and loss), thus calculating the cumulative  $N_2$  production as if no sampling took place. That is because the experimentally determined  $N_2$  accumulation (which is to be compared with the model predictions) was already corrected for the net sampling effect.

The model is developed in Vensim DSS 6.2 Double Precision (Ventana Systems, Inc. http://vensim.com/) using techniques from the field of system dynamics [29]. The model is divided into three sectors: I.  $O_2$  kinetics, II. Population dynamics of  $N_{D-}$  and  $N_{D+}$ , and III. Denitrification kinetics (Fig. 4).

### Sector I: O<sub>2</sub> kinetics

Structural-basis for the  $O_2$  kinetics is mapped in Fig. 4A: the squares represent the state variables, the circles the rate of change in the state variables, the shaded ovals the auxiliary variables, the arrows mutual dependencies between the variables, and the edges represent flows into or out of the state variables. Briefly, Fig. 4A (left to right) shows that  $O_2$  in the vial's headspace  $(O_{2HS})$  is transported  $(Tr_{O_2})$  to the liquid-phase  $(O_{2LP})$ , where it is consumed  $(Cr_{O_2})$  by both the  $N_{D-}$  and  $N_{D+}$  populations (lacking and carrying denitrification enzymes, respectively) in proportion to an identical cell-specific velocity of  $O_2$  consumption  $(\nu_{O_2})$ .  $\Delta O_{2(S)}$  represents net marginal changes in  $O_{2HS}$  due to sampling. Below we present equations and a detailed explanation of the structural components shown for this sector.

 $\mathbf{O_2}$  in the headspace. (O<sub>2HS</sub>, mol vial<sup>-1</sup>) is initialised by measured initial concentrations (Table 1) and modelled as a function of transport (Tr<sub>O2</sub>) between the headspace and the liquid [28]:

$$Tr_{O_2} = k_t \times \left(k_{H(O_2)} \times P_{O_2} - [O_2]_{LP}\right)$$
Units:  $mol\ vial^{-1}\ h^{-1}$ 

where  $k_t$  (L vial $^{-1}$  h $^{-1}$ ) is the empirically determined coefficient for the transport of  $O_2$  between the headspace and the liquid (See Table 2 for parametric values and their sources),  $k_{H(O_2)}$  (mol  $L^{-1}$  atm $^{-1}$ ) is the solubility of  $O_2$  in water at  $20^{\circ}C,\ P_{O_2}$  (atm) is the partial pressure of  $O_2$  in the headspace, and  $[O_2]_{LP}$  (mol  $L^{-1}$ ) is the  $O_2$  concentration in the liquid-phase  $\left([O_2]_{LP} = \frac{O_{2LP}}{Vol_{LP}}\right)$ .

In addition, changes in  $O_{2HS}$  due to sampling are included at the reported sampling times. The robotised incubation system [28] used in the experiment monitors gas concentrations by sampling the headspace, where each sampling alters the concentrations in a predictable manner: a fraction of  $O_{2HS}$  is removed and replaced by He (dilution), but the sampling also results in a marginal leakage of  $O_2$  through the tubing and membranes of the injection system. Eq. 2 shows how the model calculates the *net* change in  $O_{2HS}$  ( $\Delta O_{2(S)}$ ) as a result of each sampling:

$$\Delta O_{2(S)} = \frac{(O_{2leak} - O_{2HS} \times D)}{t_s}$$

$$mol \ rial^{-1} \ h^{-1}$$
(2)

where  $O_{2leak}$  (mol vial $^{-1}$ ) is the  $O_2$  leakage into the headspace, D (dilution) is the fraction of  $O_{2HS}$  replaced by He, and  $t_s$  (h) is the time taken to complete each sampling.  $\Delta O_{2(S)}$  is negative if  $O_{2HS}$  is greater than  $0.58~\mu mol\ vial^{-1}$  and marginally positive if it is less than that.

 $\textbf{O_2}$  in the liquid-phase.  $(O_{2LP}, \ \text{mol vial}^{-1}, \ \text{see Fig. 4A})$  is initialised by assuming equilibrium with  $O_{2HS}$  at the time of inoculation  $\left(O_{2LP}(t_0) = P_{O_2} \times k_{H(O_2)} \times Vol_{LP}\right).$   $O_{2LP}$  is modelled as a function of its transport into the liquid  $(Tr_{O_2}, \ \text{Eq. 1})$  and consumption rate  $(Cr_{O_2}, \ \text{mol vial}^{-1} \ h^{-1}),$  where the latter is

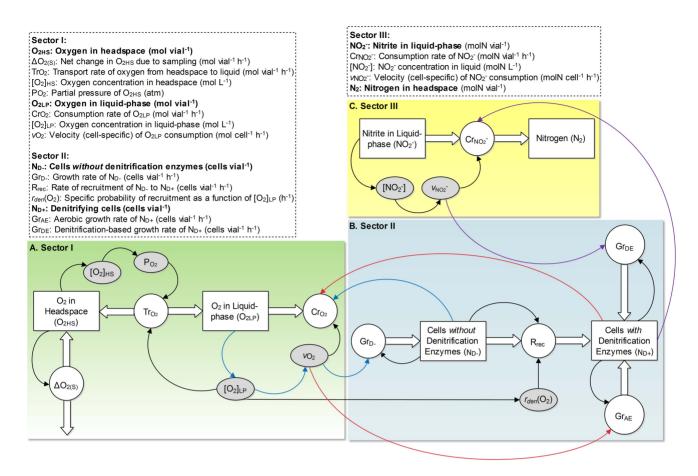


Figure 4. A stock and flow diagram of the model's structure. The squares represent the state variables, the circles the rate of change in the state variables, the shaded ovals the auxiliary variables, the arrows dependencies between the variables, and the edges represent flows into or out of the state variables. A. The panel represents the structure that governs the  $O_2$  kinetics. Briefly, it shows that  $O_2$  in the vial's headspace  $(O_{2HS})$  is transported  $(Tr_{O_2})$  to the liquid-phase  $(O_{2LP})$ , where it is consumed  $(Cr_{O_2})$  by both  $N_{D-}$  and  $N_{D+}$  populations with an identical cell-specific velocity of  $O_2$  consumption  $(\nu_{O_2})$ .  $\Delta O_{2(S)}$  represents net marginal changes in  $O_{2HS}$  due to sampling. B. The panel represents the structural basis for population dynamics of the cells without  $(N_{D-})$  and with  $(N_{D+})$  denitrification enzymes. Briefly, it shows that both the populations are able to grow by aerobic respiration  $(Gr_{D-})$  and  $Gr_{AE}$ , respectively). The growth rate of  $N_{D+}$ , nowever, is primarily based on denitrification  $(Gr_{DE})$ . Initially,  $N_{D+}=0$  and is populated through recruitment  $(R_{rec})$  of the cells from  $N_{D-}$ , where the recruitment is a function of  $N_{D-}$  and an  $[O_2]$  dependent specific-probability of the recruitment  $(r_{den}(O_2))$  for any  $N_{D-}$  cell. C. The panel represents the structural basis for the  $NO_2^-/N_2$  kinetics. Briefly, it illustrates that  $N_{D+}$  control the consumption rate of  $NO_2^-$  ( $Cr_{NO_2^-}$ ), recovered as  $N_2$ , in proportion to a cell-specific velocity of  $NO_2^-$  consumption  $(v_{NO_2^-})$ .

modelled as a function of total cell numbers and the cell-specific velocity of  $O_2$  consumption:

$$\frac{d(O_{2LP})}{dt} = Tr_{O_2} - Cr_{O_2} = Tr_{O_2} - (N_{D-} + N_{D+}) \times v_{O_2}$$
(3)
$$mol\ vial^{-1}\ h^{-1}$$

where  $N_{D-}$  and  $N_{D+}$  (cells vial<sup>-1</sup>, see Sector II for details) are the cells without and with denitrification enzymes, respectively, and  $\nu_{O_2}$  (mol cell<sup>-1</sup> h<sup>-1</sup>) is the cell-specific velocity of  $O_2$  consumption. Thus, we assume that the  $N_{D+}$  and  $N_{D-}$  cells have the same potential to consume  $O_2$ .

 $\nu_{\rm O_2}$  is modelled as a Michaelis-Menten function of  ${\rm O_2}$  concentration:

$$v_{O_2} = \frac{v_{max(O_2)} \times [O_2]_{LP}}{\left(K_{m(O_2)} + [O_2]_{LP}\right)}$$

$$mol \ cell^{-1} \ h^{-1}$$
(4)

where  $v_{max(O_2)}$  (mol cell<sup>-1</sup> h<sup>-1</sup>) is the maximum cell-specific velocity of  $O_2$  consumption (determined under the actual experimental conditions),  $[O_2]_{LP}$  (mol  $L^{-1}$ ) is the  $O_2$  concentration in the liquid-phase, and  $K_{m(O_2)}$  (mol  $L^{-1}$ ) is the half saturation constant for  $O_2$  reduction.

# Sector II: Population dynamics of the cells without ( $N_{D-}$ ) and with ( $N_{D+}$ ) denitrification proteome

Fig. 4B represents the structure governing the population dynamics of  $N_{D-}$  and  $N_{D+}.$  Briefly, the figure shows that both the populations are able to grow by aerobic respiration ( $Gr_{D-}$  and  $Gr_{AE}$ , respectively). Initially,  $N_{D+}=0$  and is populated through recruitment ( $R_{rec}$ ) of the cells from the  $N_{D-}$  pool, where the recruitment is a product of  $N_{D-}$  and an  $[{\rm O_2}]$  dependent specific-probability ( $h^{-1}$ ) of the recruitment ( $r_{den}({\it O_2})$ , see Eqs. 7–8). The growth rate of  $N_{D+}$  is primarily based on denitrification ( $Gr_{DE}$ ), but the  $N_{D+}$  cells that are recruited before  ${\rm O_2}$  is completely exhausted also grow by consuming the remaining traces of  ${\rm O_2}.$  Below we present equations and a detailed explanation of the structural components shown for this sector.

Table 2. Model parameters.

|                                     | Description   | Value                  | Units   | Reference              |
|-------------------------------------|---|------------------------|---|------------------------|
| Sector I: O <sub>2</sub> Kineti     | cs  |                        |   |                        |
| )                                   | Dilution: the fraction of O <sub>2</sub> replaced by He during sampling                 | 0.035                  | Unitless  | [28]                   |
| H(O <sub>2</sub> )                  | Solubility of O <sub>2</sub> in water (20°C)  | 0.00139                | $\mathrm{mol}\ \mathrm{L}^{-1}\ \mathrm{atm}^{-1}$  | [37]                   |
| t                                   | The O <sub>2</sub> transport coefficient between headspace and liquid                   | 1.62                   | $L vial^{-1} h^{-1}$                                | [28]                   |
| ) <sub>2leak</sub>                  | O <sub>2</sub> leakage into the vial during each sampling                               | 2.04×10 <sup>-8</sup>  | mol vial <sup>-1</sup>                              | [28]                   |
|                                     | The time taken to complete each sampling  | 0.017                  | h   | [28]                   |
| -m(O <sub>2</sub> )                 | The half saturation constant for $\mathrm{O}_2$ consumption                             | 2.5×10 <sup>-7</sup>   | mol L <sup>-1</sup>                                 | Model-based estimation |
| $max(O_2)$                          | The maximum cell-specific velocity of O <sub>2</sub> consumption                        | 1.33×10 <sup>-15</sup> | $\mathrm{mol}\;\mathrm{cell}^{-1}\;\mathrm{h}^{-1}$ | [4,8]                  |
| ector II: Populati                  | on dynamics of the cells without ( $N_{D-}$ ) and with ( $N_{D+}$ ) denitrification     | on proteome            |   |                        |
| $[O_2]_{\min}$                      | $\left[\text{O}_{2}\right]$ in the liquid below which the recruitment to $N_{D+}$ halts | 1×10 <sup>-9</sup>     | $ m mol~L^{-1}$                                     | Assumption             |
| O <sub>2</sub> ] <sub>trigger</sub> | $\left[ \text{O}_{2} \right]$ below which the recruitment to $N_{D+}$ triggers          | $9.75 \times 10^{-6}$  | $mol L^{-1}$  | [22]                   |
| den                                 | The specific-probability of recruitment of a cell to $N_{\rm D+}$                       | 0.0052                 | $h^{-1}$  | Model-based estimation |
| NO <sub>2</sub>                     | The growth yield per molN $\mathrm{NO}_2^-$   | 5.79×10 <sup>13</sup>  | cells molN <sup>-1</sup>                            | [4,8]                  |
| O <sub>2</sub>                      | The growth yield per mol O <sub>2</sub>   | 15×10 <sup>13</sup>    | cells mol <sup>-1</sup>                             | [4,8]                  |
| ector III: Denitrifi                | cation Kinetics   |                        |   |                        |
| $m(NO_2^-)$                         | The half saturation constant for $NO_2^-$ reduction                                     | 4×10 <sup>-6</sup>     | $molN L^{-1}$                                       | [33,34]                |
| $max(NO_2^-)$                       | The maximum cell-specific velocity of $NO_2^-\ \mbox{reduction}$                        | 1.83×10 <sup>-15</sup> | ${\rm molN~cell^{-1}~h^{-1}}$                       | [4,8]                  |
| General                             |   |                        |   |                        |
| Ł                                   | Universal gas constant  | 0.083                  | L atm $K^{-1} \text{ mol}^{-1}$                     | -                      |
|                                     | Temperature   | 293.1                  | K   | [4,8]                  |
| ol <sub>HS</sub>                    | Headspace volume  | 0.07                   | L vial <sup>-1</sup>                                | [4,8]                  |
| Vol <sub>LP</sub>                   | Liquid-phase volume   | 0.05                   | L vial <sup>-1</sup>                                | [4,8]                  |

doi:10.1371/journal.pcbi.1003933.t002

The pool of the cells lacking denitrification proteome. The pool of the cells lacking denitrification proteome  $(N_{D-})$  is initialised with  $3\times10^8$  cells vial<sup>-1</sup>. The population dynamics of  $N_{D-}$  are modelled as:

$$\frac{d(N_{D-})}{dt} = Gr_{D-} - R_{rec}$$
 (5)

$$cells \ vial^{-1} \ h^{-1}$$

where  $Gr_{D-}$  (cells  $vial^{-1}\ h^{-1}$ ) is the (aerobic) growth rate, and  $R_{rec}$  (cells  $vial^{-1}\ h^{-1}$ , Eq. 7) is the rate of recruitment of  $N_{D-}$  to the  $N_{D+}$  pool.

 $Gr_{D-}$  is modelled as:

$$Gr_{D-} = N_{D-} \times v_{O_2} \times Y_{O_2}$$
 (6)

$$cells \ vial^{-1} \ h^{-1}$$

where  $v_{O_2}$  (mol cell<sup>-1</sup> h<sup>-1</sup>, Eq. 4) is the cell-specific velocity of  $O_2$  consumption, and  $Y_{O_2}$  (cells mol<sup>-1</sup>) is the cell yield per mole of  $O_2$  (determined under the actual experimental conditions).

The rate of recruitment. The rate of recruitment ( $R_{rec}$ , see Fig. 4B) of the cells from  $N_{D-}$  to  $N_{D+}$  is modelled as:

$$R_{rec} = N_{D-} \times r_{den}(O_2)$$

$$cells \ vial^{-1} \ h^{-1}$$

$$(7)$$

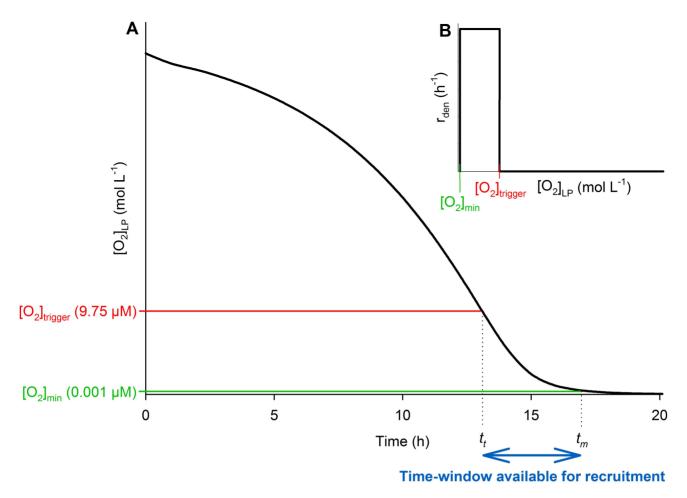
where  $r_{den}(O_2)$  (h<sup>-1</sup>) represents the conditional specificprobability for any N<sub>D-</sub> cell to be recruited to denitrification, modelled as a function of O<sub>2</sub> concentration in the liquid-phase ([O<sub>2</sub>]<sub>LP</sub>, see Fig. 5):

$$r_{den}(O_2) = \begin{cases} 0 & for \quad [O_2]_{LP} > [O_2]_{trigger} \\ r_{den} & for \quad [O_2]_{min} < [O_2]_{LP} < [O_2]_{trigger} \\ 0 & for \quad [O_2]_{LP} < [O_2]_{min} \end{cases}$$
(8)

$$h^{-1}$$

where  $r_{den}$  (h<sup>-1</sup>) is a constant representing the specific-probability of the recruitment,  $[O_2]_{trigger}$  is the  $O_2$  concentration above which the transcription of nirS is effectively suppressed by  $O_2$ , and  $[O_2]_{min}$  is the  $O_2$  concentration assumed to provide minimum energy for the initial transcription to result in functional NirS. Once the first molecules of NirS are produced while  $[O_2]_{min} < [O_2]_{LP} < [O_2]_{trigger}$ , the transcription of nirS will be greatly enhanced through positive feedback by NO, paving the way for a full-scale production of denitrification proteome [10] (See Introduction and Fig. 2 for details).

 $[O_2]_{trigger}$  (= 9.75×10<sup>-6</sup> mol L<sup>-1</sup>) is the empirically determined  $[O_2]_{LP}$  at the outset of NO accumulation: Bergaust *et al.* [8] estimated  $[O_2]_{trigger}$  between 0.1–12  $\mu$ M, but recent *Pa. denitrificans* batch incubation data have provided a more precise estimate between 8.8–10.7  $\mu$ M (average = 9.75  $\mu$ M) [22].



**Figure 5. Modelling of**  $r_{den}(h^{-1})$  **as a function of**  $[O_2]_{LP}$ **. A.** The panel shows the  $O_2$  concentration in the liquid-phase  $([O_2]_{LP})$  falling as a result of aerobic respiration. **B.** The panel shows the probability for a cell to switch to denitrification  $(r_{den}, h^{-1})$  modelled as a function of  $[O_2]_{LP}$ .  $[O_2]_{trigger}$  (Panels A & B) is the concentration below which  $r_{den}$  is assumed to trigger (due to withdrawal of the transcriptional control of  $O_2$  on denitrification [22]), whereas  $[O_2]_{min}$  is assumed to be the concentration below which  $r_{den}$  terminates (due to lack of energy for enzyme synthesis). The double-headed arrow (at the bottom of Panel A) illustrates the limited time-window  $(t_m - t_t)$  available for the cells to switch to denitrification. doi:10.1371/journal.pcbi.1003933.g005

As for  $[O_2]_{min}$ , we lack empirical basis for determining the parameter value, but sensitivity of the model to this parameter was tested (See Results/Discussion). Our simulations were run with  $[O_2]_{min} = 1 \times 10^{-9} \text{ mol L}^{-1}$ , which would sustain an aerobic respiration rate equivalent to 0.4% of the empirically determined  $v_{max(O_2)}$  (assuming our estimated  $K_{m(O_2)} = 2.5 \times 10^{-7} \text{ mol L}^{-1}$ , Table 2).

As modelled, the time-window for the recruitment to denitrification depends on the time taken to deplete  $\left[O_2\right]_{LP}$  from  $\left[O_2\right]_{trigger}$  to  $\left[O_2\right]_{min}$  (Fig. 5); for obvious reasons, the length of this time-window depends on the cell density.

The lag observed between the emergence of denitrification gene transcripts and the subsequent gas products is as short as 20 minutes [8,22], which is insignificant in the sense that the estimations of  $r_{\text{den}}$  and  $F_{\text{den}}$  will not be affected by including it in the model. Therefore, the recruitment (Eq. 7) is modelled as an instantaneous event.

Calculation of  $F_{den}$ : The fraction of the cells recruited to denitrification.  $F_{den}$  is calculated based on the integral of the recruitment (Eq. 7):

$$F_{\text{den}} = 1 - e^{-r_{\text{den}} \times (t_m - t_t)}$$

$$Dimensionless$$
(9)

where  $\mathbf{r}_{\text{den}}$  (h<sup>-1</sup>, see Eqs. 7–8 and Fig. 5) is the specific-probability for the recruitment of a cell to denitrification,  $t_t$  is the time when  $[O_2]$  in the liquid falls below  $[O_2]_{\text{trigger}}$  (the concentration below which  $\mathbf{r}_{\text{den}}$  triggers), and  $t_m$  is the time when  $[O_2]$  in the liquid falls below  $[O_2]_{\text{min}}$  (the concentration below which  $\mathbf{r}_{\text{den}}$  is assumed to be zero). Hence, effectively,  $\mathbf{F}_{\text{den}}$  expresses the probability for any cell to switch to denitrification within the time-frame  $t_m - t_t$ .

The pool of the cells carrying denitrification proteome. The pool of the cells carrying denitrification proteome  $(N_{D+}$ , see Fig. 4B) is initialised with zero cells, and its population dynamics are modelled as:

$$\frac{d(N_{D+})}{dt} = R_{rec} + Gr_{DE} + Gr_{AE}$$
 (10)

$$cells vial^{-1} h^{-1}$$

where  $R_{rec}$  (cells vial $^{-1}$   $h^{-1},$  Eq. 7) is the recruitment rate,  $Gr_{DE}$  (cells vial $^{-1}$   $h^{-1}$ ) the denitrification-based growth and  $Gr_{AE}$  (cells vial $^{-1}$   $h^{-1}$ ) the aerobic growth rate.

GrDE is modelled as:

$$Gr_{DE} = N_{D+} \times \nu_{NO_2^-} \times Y_{NO_2^-}$$

$$cells \ vial^{-1} \ h^{-1}$$
(11)

where  $\nu_{NO_2^-}$  (molN cell<sup>-1</sup> h<sup>-1</sup>, see Eq. 15) is the cell-specific velocity of  $NO_2^-$  reduction, and  $Y_{NO_2^-}$  (cells molN<sup>-1</sup>) is the growth yield per molN of  $NO_2^-$  as the e<sup>-</sup>-acceptor (determined under the actual experimental conditions).

The  $N_{D+}$  cells are assumed to have the same ability as  $N_{D-}$  to grow by aerobic respiration; their aerobic growth rate is formulated as:

$$Gr_{AE} = N_{D+} \times v_{O_2} \times Y_{O_2}$$

$$cells vial^{-1} h^{-1}$$
(12)

where  $\nu_{O_2}$  (mol cell<sup>-1</sup> h<sup>-1</sup>, see Eq. 4) is the cell-specific velocity of  $O_2$  consumption, and  $Y_{O_2}$  (cells mol<sup>-1</sup>) is the growth yield per mole of  $O_2$  as the e<sup>-</sup>-acceptor.

#### Sector III: Denitrification kinetics

The structure controlling the denitrification kinetics is mapped in Fig. 4C. Briefly, the figure shows that the cells with denitrification proteome  $(N_{D+})$  control the consumption rate of  $NO_2^-$  ( $Cr_{NO_2^-}$ ), recovered as  $N_2$ , in proportion to a cell-specific velocity of  $NO_2^-$  consumption  $(\nu_{NO_2^-})$ . The denitrification intermediates NO and  $N_2O$  are not explicitly modelled, as they accumulated to miniscule concentrations only [4,8].

 $NO_2^-$  and  $N_2$ . The  $NO_2^-$  pool (molN vial<sup>-1</sup>) is initialised by measured initial concentrations (Table 1), and the  $N_2$  pool is initialised with zero molN vial<sup>-1</sup>.  $NO_2^-$  and  $N_2$  kinetics are modelled as:

$$\frac{d(NO_2^-)}{dt} = -Cr_{NO_2^-} \tag{13}$$

 $molN \ vial^{-1} \ h^{-1}$ 

where  $Cr_{NO_2^-}$  is the consumption rate of  $NO_2^-$ :

$$Cr_{NO_2^-} = N_{D+} \times v_{NO_2^-}$$
 (14)

 $molN \ vial^{-1} \ h^{-1}$ 

where  $N_{D+}$  (cells vial<sup>-1</sup>) represents the denitrifying cells, and  $\nu_{NO_2^-}$  (molN cell<sup>-1</sup> h<sup>-1</sup>) is the cell-specific velocity of  $NO_2^-$  reduction, which is modelled as a function of  $NO_2^-$  using the Michaelis-Menten equation:

$$v_{\text{NO}_{2}^{-}} = \frac{v_{\max\left(\text{NO}_{2}^{-}\right)} \times \left[\text{NO}_{2}^{-}\right]}{\left(K_{m\left(\text{NO}_{2}^{-}\right)} + \left[\text{NO}_{2}^{-}\right]\right)}$$
(15)

$$molN \ cell^{-1} \ h^{-1}$$

where  $v_{max\left(NO_{2}^{-}\right)}$  (molN cell<sup>-1</sup> h<sup>-1</sup>) is the maximum cell-specific velocity of  $NO_{2}^{-}$  consumption (determined under the actual experimental conditions),  $\left[NO_{2}^{-}\right]$  (molN L<sup>-1</sup>) is the  $NO_{2}^{-}$  concentration in the liquid-phase, and  $K_{m\left(NO_{2}^{-}\right)}$  (molN L<sup>-1</sup>) is the half saturation constant for  $NO_{2}^{-}$  reduction.

See Table 2 for a summary of the parametric values and their sources and Table 3 for the initial values assigned to the state variables.

#### **Parameterisation**

Most of the parameter values used in the model are well established in the literature (See Table 2). However, somewhat uncertain parameters include  $K_{m(O_2)}$ ,  $K_{m(NO_2^-)}$ ,  $[O_2]_{trigger}$ , and the assumed parameter  $[O_2]_{min}$ :

 $\mathbf{K}_{m(\mathbf{O}_2)}$ . Pa. denitrificans has three alternative terminal oxidases [30] with  $\mathbf{K}_{m(\mathbf{O}_2)}$  ranging from nM to  $\mu$ M [31,32], so we decided to estimate  $\mathbf{K}_{m(\mathbf{O}_2)}$  by fitting our model to the data. Unfortunately, Bergaust et al.'s [4,8]  $\sim$ 0%  $\mathbf{O}_2$  treatments data, for which  $\mathbf{K}_{m(\mathbf{O}_2)}$  is relevant, has technical problems (needle clogging and/or high  $\mathbf{O}_2$  leakage during sampling). Therefore, we estimated  $\mathbf{K}_{m(\mathbf{O}_2)}$  (=  $2.5 \times 10^{-7}$  mol  $\mathbf{L}^{-1}$ ) by aptly simulating our model against another  $\sim$ 0%  $\mathbf{O}_2$  data-set produced by batch cultivations of Pa. denitrificans under similar experimental conditions [22].

 $\mathbf{K}_{m(\mathbf{NO}_{2}^{-})}$  is given in the literature as 4–5  $\mu$ M [33,34]. The model, however, does not show any considerable sensitivity to this parameter even within a range as wide as 0.1–10  $\mu$ M because the simulated experiments were operating with much higher [NO<sub>2</sub><sup>-</sup>].

 $[\mathbf{O}_2]_{\text{trigger}}$  (= 9.75×10<sup>-6</sup> mol L<sup>-1</sup>) is empirically determined as the  $[\mathbf{O}_2]_{\text{LP}}$  at the outset of NO accumulation: Bergaust *et al.* [8]

**Table 3.** Initial values for the state variables.

|   | Symbol                       | Value                                       | Units                    | Reference  |
|---|------------------------------|---|--------------------------|------------|
| Sector 1: O2 Kinetics                         |                              |   |                          |            |
| Initial O <sub>2</sub> in the headspace       | $O_{2HS}(t_0) \\$            | See Table 5                                 | $mol \ vial^{-1}$        | [4,8]      |
| Initial O <sub>2</sub> in the liquid-phase    | $O_{2LP}(t_0)$               | Equilibrium with $O_{2HS}(t_{0})$           | mol vial <sup>-1</sup>   | Assumption |
| Sector II: Population dynamics of the         | cells without ( $N_{D-}$ ) a | and with ( $N_{D\pm}$ )denitrification prot | eome                     |            |
| The initial number of cells                   | $N_{D-}(t_0) \\$             | 3×10 <sup>8</sup>                           | cells vial <sup>-1</sup> | [4,8]      |
| The initial number of denitrifying cells      | $N_{\mathbf{D}+}(t_0)$       | 0   | cells vial <sup>-1</sup> | Assumption |
| Sector III: Denitrification Kinetics          |                              |   |                          |            |
| Initial $\mathrm{NO}_2^-$ in the liquid-phase | $NO_2^-(t_0) \\$             | See Table 5                                 | $molN \ vial^{-1}$       | [4,8]      |
| Initial N <sub>2</sub> in the headspace       | $N_2(t_0)$                   | 0   | molN vial <sup>-1</sup>  | [4,8]      |

doi:10.1371/journal.pcbi.1003933.t003

estimated  $[O_2]_{trigger}$  between 0.1–12  $\mu M$ , but recent batch incubation data from Pa. denitrificans have provided a more precise estimate in the range 8.8–10.7  $\mu M$  (average = 9.75  $\mu M$ ) [22]. The model, however, is not sensitive to  $[O_2]_{trigger}$  within the latter range because of a high velocity of  $O_2$  depletion.  $[O_2]_{min}$  (=1×10<sup>-9</sup> mol  $L^{-1}$ ) is assigned an arbitrary low value,

 $[\mathbf{O}_2]_{\min}$  (= 1×10<sup>-9</sup> mol L<sup>-1</sup>) is assigned an arbitrary low value, since we lack any empirical estimation/data to support it. To compensate for the uncertainty, we conducted a sensitivity analysis exploring the consequences of increasing or decreasing  $[\mathbf{O}_2]_{\min}$  by one order of magnitude (See Results/Discussion).

### Results/Discussion

The specific-probability  $(r_{den}, h^{-1})$  of recruitment of a cell to denitrification

To test the assumption of a single homogeneous population, we forced our model to achieve 100% recruitment to denitrification by setting  $r_{den} = 1 \ h^{-1}$ . In consequence, the simulated  $N_2$  accumulation (molN vial $^{-1}$ ) showed gross overestimation as compared to the measured for all the treatments (as illustrated for some randomly selected ones in Fig. 6).

To find a more adequate value,  $r_{\text{den}}$  was calibrated to produce the best possible match between the simulated and measured  $N_2$  through optimisation. (The optimisation was carried out in Vensim DSS 6.2 Double Precision, http://vensim.com/). Table 4 presents the optimal  $r_{\text{den}}$  for each treatment; no consistent effect of initial  $[O_2]$  and  $[NO_2^-]$  was found on the optimal results. The average for all the treatments = 0.0052, which appears to give reasonable fit between the simulated and measured  $N_2$  (See Figs. 7, 8, and 9). This indicates that the simulations with  $r_{\text{den}}=0.0052$  should provide a reasonable approximation of  $F_{\text{den}}$  (the fraction recruited to denitrification) during the actual experiment.

**Sensitivity analysis.**  $[\mathbf{O}_2]_{min}$  (the  $O_2$  concentration below which the recruitment is arrested) was arbitrarily chosen to be  $1 \times 10^{-9}$  mol  $L^{-1}$ . In order to evaluate the sensitivity of the model to this parameter, we tested the model performance by increasing and decreasing  $[O_2]_{min}$  by one order of magnitude. For each parameter value, we estimated  $r_{den}$  for the individual vials by optimisation (as outlined in the foregoing paragraph). A good fit was obtained for both the  $[O_2]_{min}$  values, but the optimisation resulted in slightly different  $r_{den}$  values. Increasing  $[O_2]_{min}$  by a

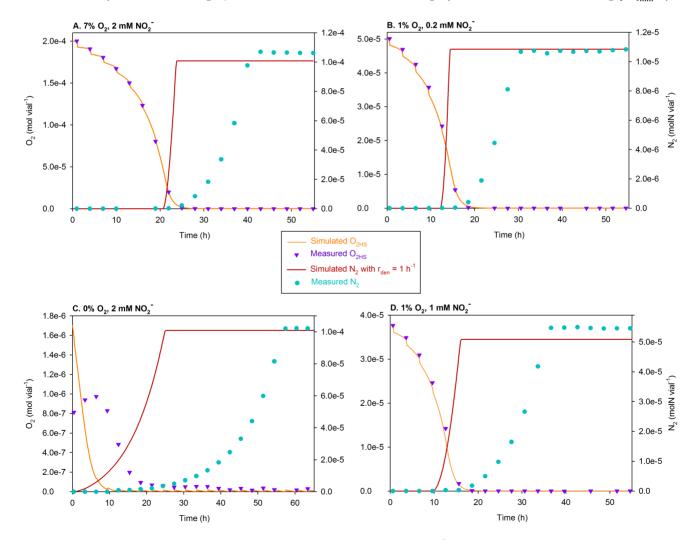


Figure 6. Comparison of the measured [4,8] and simulated data assuming  $r_{den} = 1 h^{-1}$ . Assuming a single homogeneous population, as we forced our model to achieve 100% recruitment to denitrification by setting the specific-probability of recruitment ( $r_{den}$ ) to 1  $h^{-1}$ , the simulated  $N_2$  accumulation (molN vial<sup>-1</sup>) showed considerable overestimation as compared to that measured. To illustrate this, the simulated and measured data are compared here for some randomly chosen treatments. Initial vol.%  $O_2$  in the headspace and initial  $NO_2^-$  is shown above each panel. doi:10.1371/journal.pcbi.1003933.g006

**Table 4.** Specific-probability of recruitment of a cell to denitrification  $(r_{\rm den})$  estimated for each batch culture by optimisation (best match between the simulated and measured  $N_2$  kinetics).

|           | Treatment*: $\mathbf{O}_{\mathbf{2HS}}(\mathbf{t}_0)$ (vol.%) | Optimal r <sub>den</sub> (h <sup>-1</sup> ) |  |
|-----------|---|---|--|
| Batch No. | $\mathbf{NO_2^-}(\mathbf{t}_0)$ (mM)                          |   |  |
| 1         | ~0, 0.2   | 0.0066                                      |  |
| 2         | ~0, 1   | 0.0059                                      |  |
| 3         | ~0, 2   | 0.0029                                      |  |
| 4         | 1, 0.2  | 0.0033                                      |  |
| 5         | 1, 1  | 0.0062                                      |  |
| 6         | 1, 2  | 0.0020                                      |  |
| 7         | 7, 0.2  | 0.0018                                      |  |
| 8         | 7, 1  | 0.0117                                      |  |
| 9         | 7, 2  | 0.0066                                      |  |
|           |   | Avg. = 0.0052                               |  |

\*Treatment refers to the initial concentration of  $O_2$  in the headspace (measured as headspace vol.%) and the initial concentration of  $NO_2^-$  in the medium (mM). doi:10.1371/journal.pcbi.1003933.t004

factor of 10 (to  $1\times10^{-8}$  mol L<sup>-1</sup>) resulted in 18–38% higher  $r_{den}$  estimates (average = 28%  $\pm$ stdev 10). Decreasing  $[O_2]_{min}$  by a factor of 0.1 (to  $1\times10^{-10}$  mol L<sup>-1</sup>) resulted in 5–17% lower  $r_{den}$  estimates (average = 11%  $\pm$ stdev 6).

### The fraction recruited to denitrification $(F_{den})$

A refined estimation with the presented model. Bergaust et al. [8,16] and Nadeem et al. [9] used data from batch cultivations of Pa. denitrificans, as illustrated in Fig. 1, to assess  $F_{den}.$  Their estimation was effectively  $F_{den}\!=\!\frac{N_{D+}(t_{ex})}{N(t_{ex})},$  where  $t_{ex}$  is the time when  $\mathrm{O}_2$  is exhausted,  $N_{D+}$  (cells vial  $^{-1})$  is the number of actively denitrifying cells estimated by the measured rate of denitrification (molN h<sup>-1</sup>) divided by the cell-specific denitrification (molN cell<sup>-1</sup> h<sup>-1</sup>), and N is the total number of cells estimated on the basis of O<sub>2</sub> consumption. Although this equation indisputably estimates the fraction of the cells that was actively denitrifying at the time  $t_{ex}$ , it is a biased estimate of the 'true'  $F_{den}$ because the number of cells does not remain constant through the recruitment phase: N<sub>D</sub>- (the cells without denitrification enzymes) and  $N_{D+}$  will both grow until  $O_2$  is depleted, but  $N_{D+}$  will grow faster because their growth is supported by both O2 and NOx. As a result, the estimation of F<sub>den</sub> by this equation might be too high.

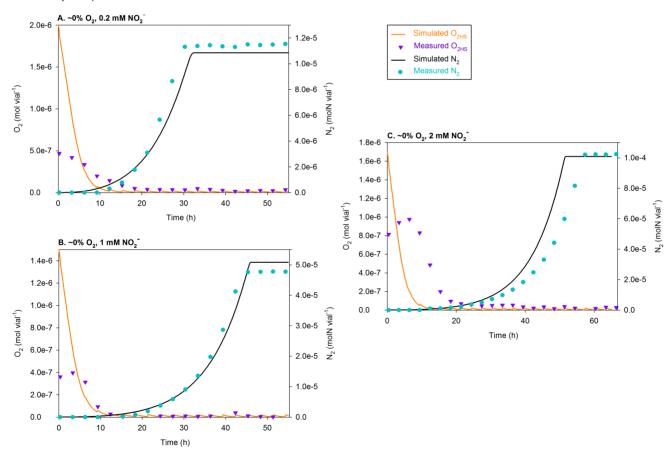


Figure 7. Simulations of the treatments with  $\sim$ 0 vol.%  $O_{2HS}$  using  $r_{den} = 0.0052~h^{-1}$ . The figure compares the measured and simulated  $O_2$  depletion (mol vial $^{-1}$ ) and  $N_2$  accumulation (molN vial $^{-1}$ ) for the  $\sim$ 0 vol.%  $O_2$  treatments of Bergaust et~al.~[4,8], i.e., the vials with near-zero  $O_2$  in the headspace ( $O_{2HS}$ ) at the time of inoculation. Separate plots are shown for each initial concentration of  $NO_2^-$  (0.2, 1, and 2 mM). The measured initial  $O_2$  was somewhat erratic due to episodes of needle clogging and/or high  $O_2$  leakage during sampling, so the initial  $O_{2HS}$  used in the simulations is chosen somewhat ad~lib so that the simulated  $O_2$  depletion coincides with that measured. The discrepancy compared to the measured  $O_2$  seems to be significant for 2 mM  $NO_2^-$  treatment. That is most likely due to the inhibitory effect of nitrite on aerobic respiration, which is not taken into account; all simulations are run with an identical  $K_{m(O_2)}$ . Near exhaustion, the simulated  $O_2$  increases slightly at each sampling time; that is due to the leakage of  $O_2$  via the injection system exceeding dilution of the headspace (with He) during each sampling. doi:10.1371/journal.pcbi.1003933.g007

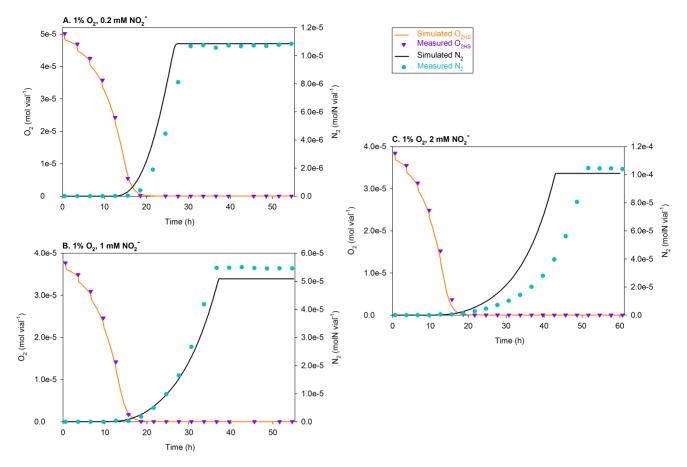


Figure 8. Simulations of the treatments with 1 vol.%  $O_{2HS}$  using  $r_{den}$  = 0.0052  $h^{-1}$ . The figure compares the measured and simulated  $O_2$  depletion (mol vial $^{-1}$ ) and  $N_2$  accumulation (molN vial $^{-1}$ ) for the treatments with 1 vol.%  $O_2$  in the headspace ( $O_{2HS}$ ) at the time of inoculation; separate plots are shown for each initial concentration of  $NO_2^-$  (0.2, 1, and 2 mM). At each sampling time, the simulated  $O_2$  is visibly reduced; that is because sampling implies 3.4% dilution of the headspace (with He). This contrasts with the simulations of the treatments with low  $O_2$  (Fig. 7), where the leakage of  $O_2$  into the system is more dominant. doi:10.1371/journal.pcbi.1003933.g008

Besides, the experimental estimation is prone to error because of infrequent sampling, since the sampling time does not necessarily coincide with  $t_{\rm ex}$ .

In contrast, our model directly and more precisely calculates  $F_{den}$  (Eq. 9) by a) explicitly simulating the actual kinetics of the recruitment of the cells to denitrification (in contrast to estimating total and denitrifying cell numbers from gas kinetics) and b) avoiding aerobic and anaerobic growth of the cells. Table 5 shows the model's estimations of  $F_{den}$  and the time-span of the recruitment ( $t_m - t_t$ ) along with the  $F_{den}$  estimations of Bergaust  $et\ al\ [8,16]$ .

In the ~0%  $O_2$  treatments,  $F_{den}$  is supported by the sampling leaks of  $O_2$ . Due to low cell density in the ~0%  $O_2$  treatments (initial  $O_2 = 1.5$ –2  $\mu$ mol), the  $O_2$  leakage into the vial during sampling (every 3 hours) caused oxygen concentrations to exceed  $[O_2]_{min}$  for 0.1–2.4 hours. This resulted in various spikes of recruitment after the initial  $O_2$  was depleted. The recruitment through these spikes amounted to, on average, ~19% of  $F_{den}$  in the ~0%  $O_2$  treatments.

 $F_{den}$  << 100%. The model's estimations of  $F_{den}$  (Table 5) corroborate the suggestion of Bergaust *et al.* [8,16] and Nadeem *et al.* [9] that in batch cultures of *Pa. denitrificans*  $F_{den}$  remains far below 100%. According to Bergaust *et al.* [8,16],  $F_{den}$  was 2–21% (average = 10%), whereas the model estimated it between 3.8–16.1% (average = 8.2%).

 $F_{den}$  is inversely related to cell density. Bergaust *et al.* [16] argued that as the velocity of  $O_2$  depletion is proportional to cell density, the time-frame available for the cells to produce (necessary initial) denitrification proteome would be inversely related to the cell density at the time of  $O_2$  depletion. Simulation results (Table 5) support this: high initial  $O_2$  concentrations resulted in high cell densities at the time of  $O_2$  depletion, shortening the time-span for the recruitment to denitrification, hence resulting in the low  $F_{den}$ .

**Underlying cause of the low F\_{den}.**  $F_{den}$  remains low because of a) the limited time-window available to the cells for the recruitment and b) the low  $r_{den}$  (specific-probability of the recruitment), presumably due to a low probability of initiating nirS transcription (subsequently reinforced through positive feedback by NO).

### Simulation of the 'diauxic lag'

To investigate whether the recruitment of a small fraction of the cells to denitrification could explain the 'diauxic lag' observed by Liu *et al.* [24], we used our model to simulate the conditions they reported for their experiment. In short, Liu *et al.* [24] incubated *Ps. denitrificans* (ATCC 13867) in oxic batch cultures, which were sparged with  $N_2$  as the cultures had reached different cell densities (OD<sub>550</sub> = 0.05–0.17). The sparging resulted in apparent diauxic lags, i.e., periods with little or no detectable growth. The length of

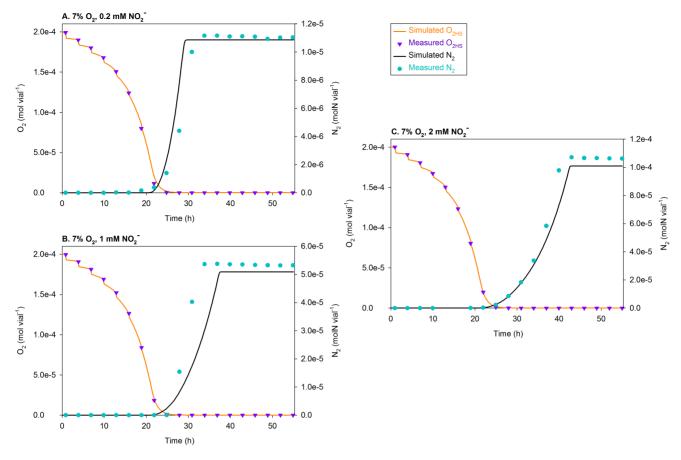


Figure 9. Simulations of the treatments with 7 vol.%  $O_{2HS}$  using  $r_{den}$  = 0.0052  $h^{-1}$ . The figure compares the measured and simulated  $O_2$  depletion (mol vial $^{-1}$ ) and  $N_2$  production (molN vial $^{-1}$ ) for the treatments with 7 vol.%  $O_2$  in the headspace ( $O_{2HS}$ ) at the time of inoculation; separate plots are shown for each initial concentration of nitrite (0.2, 1, and 2 mM). At each sampling time, the simulated  $O_2$  is visibly reduced because of sampling, which results in 3.4% dilution of the headspace (with He). doi:10.1371/journal.pcbi.1003933.g009

**Table 5.** The model's and Bergaust et al.'s [16] estimations of the fraction recruited to denitrification ( $F_{den}$ ).

| Batch No. | $O_{2HS}(t_{0})$ (vol.%) $NO_{2}^{-}\left(t_{0}\right)$ (mM) | $O_{2HS}(t_0)$ ( $\mu$ mol)* | Model-based Estimations |                  | Estimations of [16] |  |
|-----------|--|------------------------------|-------------------------|------------------|---------------------|--|
|           |  |                              | $t_m - t_t^{**}$        | F <sub>den</sub> | F <sub>den</sub>    |  |
| 1         | 0, 0.2   | 2                            | 25.8                    | 0.141            | 0.19                |  |
| 2         | 0, 1   | 1.5                          | 29.2                    | 0.161            | 0.21                |  |
| 3         | 0, 2   | 1.7                          | 27.2                    | 0.156            | 0.19                |  |
| 4         | 1, 0.2   | 50.1                         | 10.1                    | 0.052            | 0.03                |  |
| 5         | 1, 1   | 37.8                         | 11.1                    | 0.056            | 0.07                |  |
| 6         | 1, 2   | 38.4                         | 11.3                    | 0.057            | 0.04                |  |
| 7         | 7, 0.2   | 199                          | 7.4                     | 0.038            | 0.02                |  |
| 8         | 7, 1   | 200                          | 7.4                     | 0.038            | 0.07                |  |
| 9         | 7, 2   | 200                          | 7.4                     | 0.038            | 0.08                |  |
|           |  |                              |                         | Avg. = 0.082     | Avg. = 0.1          |  |

\*Refers to the initial values of  $O_2$  in the headspace ( $O_{2HS}$ ) used in the simulations. The values show some inconsistency for the treatments corresponding to the same vol.% because of traces of  $O_2$  left behind after He-washing.

\*\* $t_t$  is the time when  $[O_2]$  in the liquid falls below  $[O_2]_{trigger}$  (=9.75  $\mu$ M [22], the concentration below which recruitment of the cells to denitrification is assumed to trigger), and  $t_m$  is the time when  $[O_2]$  in the liquid falls below  $[O_2]_{min}$  (=1 nM, a practically zero concentration below which the recruitment is assumed to terminate). Due to low cell density in the  $\sim$ 0%  $O_2$  treatments, the  $O_2$  leakage into the vial during sampling (every 3 hours) caused oxygen concentration to exceed  $[O_2]_{min}$  for 0.1–2.4 hours. This resulted in various recruitment spikes after the initial  $O_2$  was depleted. If such recruitment is omitted,  $F_{den}$  = 0.126, 0.142, and 0.133 for the treatments 1, 2, and 3, respectively.

doi:10.1371/journal.pcbi.1003933.t005

such lags increased with the cell density present at the time of sparging.

**Structural amendments and parameterisation of the model.** To tentatively simulate their experiment, two changes were made in the  $O_2$  kinetics sector (Fig. 4A). Firstly, the net sampling loss of  $O_{2HS}$  ( $\Delta O_{2(S)}$ ) was omitted, since it was specifically set up for the robotised incubation system [28] used by Bergaust *et al* [4,8]. Secondly, a sparging event was introduced, which immediately takes  $O_{2HS}$  down to very low levels (=  $1 \times 10^{-9}$  mol vial<sup>-1</sup>). Since we lack information about the exact concentration of  $O_2$  immediately after the sparging, the present exercise is only qualitative.

Liu *et al.* [24] inoculated the culture to have an initial  $OD_{550} = 0.07$ , which would correspond to  $\sim 6.5 \times 10^9$  cells vial<sup>-1</sup> [4,8]. We used this number to initialise the  $N_{D-}$  pool (shown in Fig. 4B). They used  $NO_3^-$  (=157 µmolN vial<sup>-1</sup>) instead of  $NO_2^-$ , so we replaced the  $NO_2^-$  pool (Fig. 4C) by the  $NO_3^-$  pool, initialised it accordingly, and adjusted Eqs. 11 and 15: In Eq. 11,  $Y_{NO_2^-}$  was replaced with the cell yield per molN of  $NO_3^-$  as the e<sup>-</sup>-acceptor ( $Y_{NO_3^-} = 9.65 \times 10^{13}$  cells molN<sup>-1</sup> [4,8]). In Eq. 15,  $v_{max(NO_2^-)}$  was replaced with the maximum cell-specific velocity of  $NO_3^-$  consumption ( $v_{max(NO_3^-)} = 2 \times 10^{-15}$  molN cell<sup>-1</sup> h<sup>-1</sup>), calculated using the maximum specific  $NO_3^-$ -based growth rate (=0.322 h<sup>-1</sup>) reported for their experiment. Finally, in Eq. 4,  $v_{max(O_2)}$  was calibrated (=2.28×10<sup>-15</sup> mol cell<sup>-1</sup> h<sup>-1</sup>) with the reported maximum specific aerobic growth rate (=0.342 h<sup>-1</sup>).

The 'diauxic lag' is plausibly the initial growth phase of a minute  $F_{den}$  (fraction recruited to denitrification). As the experiment of Liu *et al.* [24] was simulated with the model's estimated  $r_{den} = 0.0052 \text{ h}^{-1}$  (specific-probability of recruitment),  $F_{den}$  turned out to be 1.1% for the treatment sparged at h = 1.1 and 0.2% for the one sparged at h = 2.55. Simulations of the total cell density ( $N_{D-} + N_{D+}$ ) for these cases(Fig. 10A) showed long apparent lags comparable to 10–30 h lag phases observed in their later experiments [25]. However, lags in the range that Liu *et al.* [24] observed (= 3 and 6 h for sparging at h = 1.1 and 2.55,

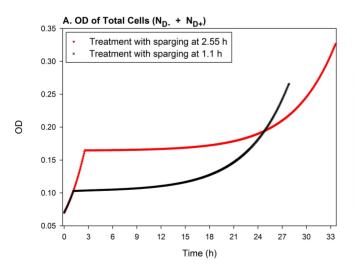
respectively) could be achieved by our model by assuming higher residual  $\rm O_2$  concentrations after sparging (resulting in a higher  $\rm F_{den}$ ). Fig. 10B isolates the OD of  $\rm N_{D+}$  for the simulated treatments and shows them on a logarithmic scale so that their exponential growth, right from the onset of anoxic conditions, becomes apparent. The figure initially shows a quick recruitment of the cells from the  $\rm N_{D-}$  to the  $\rm N_{D+}$  pool, followed by the exponential growth-phase of  $\rm N_{D+}$ .

This exercise serves to illustrate that the 'diauxic lags' observed [24-26] may simply be a result of low recruitment to denitrification in response to sudden removal of O<sub>2</sub>. This is possibly a more plausible explanation than suggested by the authors and further elaborated by Hamilton et al. [35], claiming that there is a true lag caused by extremely slow production of denitrification enzymes due to energy limitation. Our explanation of the apparent diauxic lag is corroborated by a chemostat culturing experiment conducted by Bauman et al [36]: A steady state carbon (acetate) limited continuous culture with Pa. denitrificans was made anoxic and monitored for denitrification gene transcription, N-gas production, and acetate concentrations. A transient (8–10 h) peak of acetate accumulation after O<sub>2</sub> depletion suggested an apparent diauxic lag in the metabolic activity, but denitrification started immediately and increased gradually throughout the entire 'lag' period. They further observed that the number of denitrification gene transcripts peaked sharply during the first 1-2 hours. These observations are in good agreement with our model.

The aforestated observation of Liu *et al.* [24] that the length of the apparent lags increased with the aeration period (or the cell density at the time of sparging) is also in agreement with our model demonstrating that the time available for the cells to switch to denitrification is inversely related to the cell density at the time of  $O_2$  depletion.

# Model-based hypothesis: Initial $O_2$ determines the timespan to denitrify all $NO_2^-$ to $N_2$ in a batch

Two sensitivity analyses were run to investigate the system's response to initial  $O_2$  in the headspace,  $O_{2HS}(t_0)$ : one corresponding



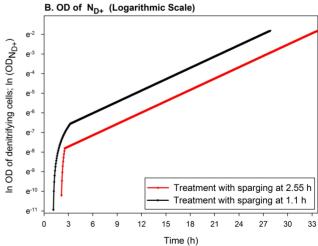


Figure 10. Simulation of the 'diauxic lags' observed by Liu et al [24]. A. The panel shows cumulated OD (optical density) of the cells without  $(N_{D-})$  and with  $(N_{D+})$  denitrification enzymes for the simulated experiment of Liu et al. [24], where one treatment was sparged at time = 2.55 h and the other at 1.1 h. The simulations show, qualitatively, similar 'lags' in the two ODs as observed by the experimenters. These apparent lags are due to exponential growth of a minute fraction of the cells that successfully switched to denitrification. The growth of this fraction remains practically undetectable (the "lag" phase) until it reaches a level comparable to the large population trapped in anoxia. B. This panel isolates the ODs of  $N_{D+}$  and show them on a logarithmic scale so that the exponential growth of  $N_{D+}$ , right from the onset of anoxic conditions, becomes visible. The graph initially shows a quick recruitment of the cells from the  $N_{D-}$  to the  $N_{D+}$  pool, followed by the exponential growth-phase.

to a range of initial  $[O_2]$  in the liquid-phase  $([O_2]_{LP}(t_0))$  below  $[O_2]_{trigger}$  (see Eqs. 7–8) and the other for a range much higher than  $[O_2]_{trigger}$ . All other model parameters and initial values remained as listed in Tables 2 and 3, respectively. The exercise helps illustrate the relative importance of aerobic growth versus the recruitment  $(F_{den})$  in determining the time taken to deplete the  $NO_2^-$  pool.

**Sensitivity analysis (1).** Sensitivity analysis (1) was run with three  $\left[O_2\right]_{LP}(t_0)$  within a very low range, starting from a concentration marginally below  $\left[O_2\right]_{trigger}$ :

- 1)  $O_{2HS}(t_0) = 2.02 \times 10^{-5} \text{ mol vial}^{-1} ([O_2]_{LP} = 9.75 \,\mu\text{M}),$
- 2)  $O_{2HS}(t_0) = 1.01 \times 10^{-5} \text{ mol vial}^{-1} ([O_2]_{LP} = 4.88 \,\mu\text{M}),$
- 3)  $O_{2HS}(t_0) = 5.04 \times 10^{-6} \text{ mol vial}^{-1} ([O_2]_{LP} = 2.44 \,\mu\text{M})$

This is rather a simple case demonstrating that increasing  $[O_2]_{LP}(t_0)$  within this low range (Fig. 11A) will result in increasing rates of denitrification (Fig. 11D) by increasing the number of aerobically grown cells  $(N_{D-}, \, \text{Fig. 11B})$  and, thus, the rate of recruitment  $(R_{rec}, \, \text{Fig. 11C})$ .

**Sensitivity analysis (2).** Sensitivity analysis (2) was run with three initial  $O_2$  concentrations much higher than  $[O_2]_{trigger}$ :

- 1)  $O_{2HS}(t_0) = 2 \times 10^{-4} \text{ mol vial}^{-1} ([O_2]_{LP} = 93 \,\mu\text{M}),$
- 2)  $O_{2HS}(t_0) = 1.19 \times 10^{-4} \text{ mol vial}^{-1} ([O_2]_{LP} = 55 \,\mu\text{M}),$
- 3)  $O_{2HS}(t_0) = 3.84 \times 10^{-5} \text{ mol vial}^{-1} ([O_2]_{LP} = 18 \,\mu\text{M})$

In this case, the cumulated  $N_2$  reached stable plateaus at nearly the same time for all the runs (Fig. 12E), despite that the time taken to deplete  $O_2$  below  $\left[O_2\right]_{trigger}$  decreased with increasing

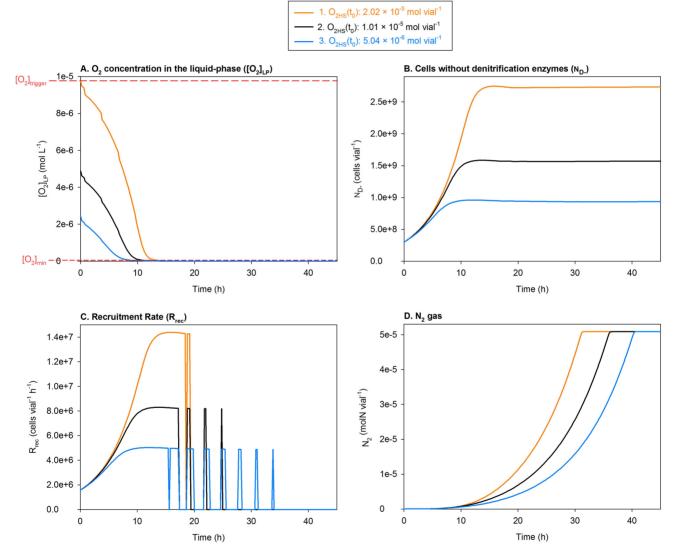


Figure 11. Sensitivity analysis (1): Varying initial  $O_2$  in the headspace  $(O_{2HS}(t_0))$  within a low range. The figure shows the impact of varying  $O_{2HS}(t_0)$  within a low range on: **A.**  $O_2$  concentration in the liquid-phase  $([O_2]_{LP})$ , **B.** The number of aerobically growing cells  $(N_{D-})$ , which do not possess denitrification enzymes, **C.** The rate of recruitment of  $N_{D-}$  to denitrification  $(R_{rec})$ , and **D.**  $N_2$  accumulation. Marked in Panel A,  $[O_2]_{trigger}$  is the  $[O_2]_{LP}$  below which  $R_{rec}$  triggers, and  $[O_2]_{min}$  is the  $[O_2]_{LP}$  below which  $R_{rec}$  terminates. In Panel C, the spikes of recruitment (following the initial recruitment) are due to spikes of  $O_2$  by sampling, causing  $[O_2]_{LP}$  to transiently exceed  $[O_2]_{min}$ . The model predicts that reducing  $[O_2]_{LP}(t_0)$  within a low range (Panel A) will lower the number of aerobically grown cells (Panel B) and, thereby, the recruitment rate (Panel C), thus increasing the time taken to deplete  $NO_2^-$  (slower  $N_2$  accumulation, Panel D). doi:10.1371/journal.pcbi.1003933.g011

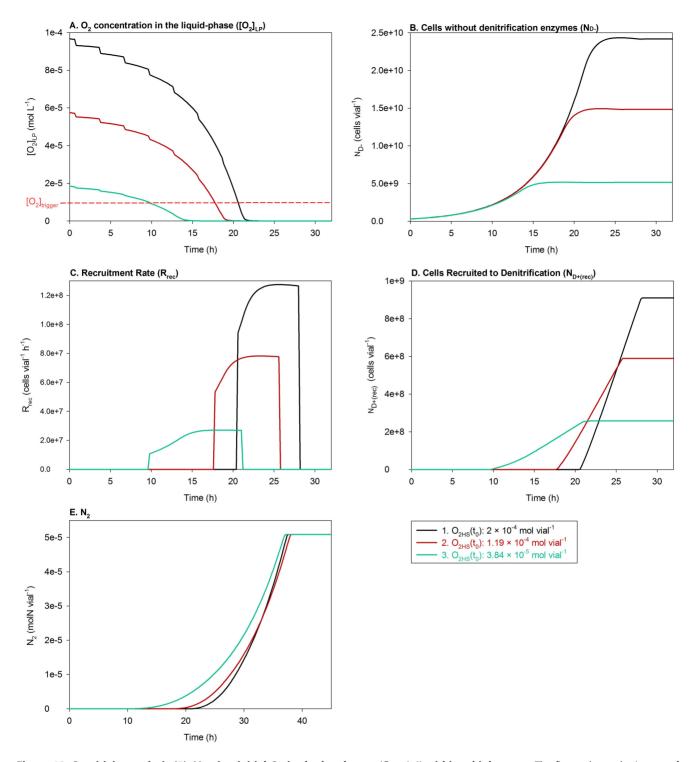


Figure 12. Sensitivity analysis (2): Varying initial  $O_2$  in the headspace  $(O_{2HS}(t_0))$  within a high range. The figure shows the impact of varying  $O_{2HS}(t_0)$  within a range much higher than  $[O_2]_{trigger}$  (the  $[O_2]$  below which recruitment of the cells to denitrification is assumed to trigger) on: **A.**  $O_2$  concentration in the liquid-phase  $([O_2]_{LP})$ , **B.** The number of aerobically growing cells  $(N_{D_-})$ , which do not possess denitrification enzymes, **C.** The rate of recruitment of  $N_{D_-}$  to denitrification  $(R_{rec})$ , **D.** The number of cells as a result of the recruitment alone  $(N_{D_+(rec)})$ , i.e., the denitrifying cells  $(N_{D_+})$  but without aerobic and  $NO_x$ -based growth, and **E.** Cumulated  $N_2$ . The cumulated  $N_2$  reached stable plateaus at nearly the same time for all the runs (Panel E), despite the fact that the time taken to deplete  $O_2$  below  $[O_2]_{trigger}$  decreased with increasing  $[O_2]_{LP}(t_0)$  (Panel A). Thus, once denitrification was initiated, the rates increased with increasing initial  $[O_2]_{LP}$  due to an increasing population of oxygen-grown cells (Panels B–D). The fraction of the cells recruited to denitrification  $(F_{den})$  declined with increasing initial  $O_2$  concentration (not shown), but this was not sufficient to compensate for the increasing number of oxygen-raised cells. doi:10.1371/journal.pcbi.1003933.g012

 $[O_2]_{LP}(t_0)$  (Fig. 12A), reducing the time available to the cells for switching to denitrification (See Fig. 5). Thus, once denitrification was initiated, the rates increased with increasing  $[O_2]_{LP}(t_0)$  due to an increasing population of oxygen-grown cells (Fig. 12B–D). F<sub>den</sub> (Eq. 9) declined with increasing  $[O_2]_{LP}(t_0)$  (F<sub>den</sub> = 0.058, 0.041 and 0.028 for runs 3, 2 and 1, respectively), but this was not sufficient to compensate for the increasing number of oxygen-raised cells.

If the model is run without any initial  $O_2$ , there would be no recruitment and, hence, no denitrification. Verification of this in batch cultures is difficult because traces of  $O_2$  remain after Hewashing of the batches. However, we (Bergaust *et al.*, unpublished data) have been able to demonstrate that the aerobically grown *Pa. denitrificans* cells are indeed entrapped in anoxia if transferred to anoxic conditions as instantaneously as in the experiments conducted by Højberg *et al.* [15].

#### Conclusion

The prevailing wisdom in denitrification research is that, under impending anoxic conditions, all cells in a batch culture of denitrifying bacteria will switch to denitrification. However, recent experiments with batch cultures of Pa. denitrificans have provided evidence that, in response to  $O_2$  depletion, only a small fraction (F<sub>den</sub>) of the entire population is able to switch to denitrification [4,8,9]. The evidence is based on indirect analyses of e<sup>-</sup>-flow rates to O<sub>2</sub> and NO<sub>3</sub> during the transition of the cells from aerobic to anaerobic respiration. To provide a direct and refined estimation of F<sub>den</sub>, we constructed a dynamic model and directly simulated kinetics of recruitment of the cells to denitrification. We first formulated a hypothesis as to the underlying regulatory mechanism of cell differentiation under approaching anoxia. Briefly, it is that the low F<sub>den</sub> is due to a low probability of initiating transcription of the nirS genes, but once initiated, the transcription is greatly enhanced through autocatalytic positive feedback by NO, resulting in the recruitment of the transcribing cell to

#### References

- 1. Zumft WG (1997) Cell biology and molecular basis of denitrification. Microbiol Mol Biol Rev 61: 533–616.
- Schlesinger WH (2009) On the fate of anthropogenic nitrogen. Proc Natl Acad Sci U S A 106: 203–208.
- Bakken LR, Dörsch P (2007) Nitrous oxide emission and global changes: modelling approaches. In: Bothe H, Ferguson SJ, Newton WE, editors. Biology of the Nitrogen Cycle Amsterdam: Elsevier B. V. pp. 382–395.
- Bergaust L, Bakken LR, Frostegård Å (2011) Denitrification regulatory phenotype, a new term for the characterization of denitrifying bacteria. Biochem Soc Trans 39: 207–212.
- Vasiliadou IA, Siozios S, Papadas IT, Bourtzis K, Pavlou S, et al. (2006) Kinetics
  of pure cultures of hydrogen-oxidizing denitrifying bacteria and modeling of the
  interactions among them in mixed cultures. Biotechnol Bioeng 95: 513–525.
- Woolfenden HC, Gates AJ, Bocking C, Blyth MG, Richardson DJ, et al. (2013) Modeling the effect of copper availability on bacterial denitrification. Microbiologyopen 2: 756–765.
- 7. Bakken LR, Bergaust L, Liu B, Frostegård Å (2012) Regulation of denitrification at the cellular level a clue to understanding of  $N_2O$  emissions from soils. Philos Trans R Soc Lond B Biol Sci 367: 1226–1234.
- Bergaust L, Mao Y, Bakken LR, Frostegård Å (2010) Denitrification response patterns during the transition to anoxic respiration and posttranscriptional effects of suboptimal pH on nitrogen oxide reductase in *Paracoccus denitrificans*. Appl Environ Microbiol 76: 6387–6396.
- Nadeem S, Dörsch P, Bakken LR (2013) The significance of early accumulation of nanomolar concentrations of NO as an inducer of denitrification. FEMS Microbiol Ecol 83: 672–684.
- van Spanning RJM, Richardson DJ, Ferguson SJ (2007) Introduction to the biochemistry and molecular biology of denitrification. In: Bothe H, Ferguson SJ, Newton WE, editors. Biology of the Nitrogen Cycle Amsterdam: Elsevier B. V. pp. 3–20.
- Proshkin S, Rahmouni AR, Mironov A, Nudler E (2010) Cooperation between translating ribosomes and RNA polymerase in transcription elongation. Science 328: 504–508.

denitrification. Then, as we implemented this hypothesis in the model, the simulation results showed that the specific-probability  $(F_{\rm den})$  of 0.0052  $(h^{-1})$  for a cell to switch to denitrification is sufficient to robustly simulate the measured denitrification gas kinetics. The model estimated the resultant  $F_{\rm den}$  between 3.8–16.1% only (average = 8.2%). The phenomenon may be considered as a 'bet-hedging' regulation 'strategy' [12]: the fraction switching to denitrification benefits if the anoxic spell is long and  $NO_x$  remains available, whereas the non-switching fraction benefits, by saving energy required for the protein synthesis, if the anoxic spell is short. The strategy has important implications for the interpretation of numerous experiments on  $Pa.\ denitrificans$  and other denitrifying organisms, as this study has illustrated by presenting a more plausible explanation of the apparent diauxic lags [24] on the basis of the low  $F_{\rm den}$ .

### **Supporting Information**

**Dataset S1** contains a Vensim simulation model (Hassan\_et\_al\_ 2014.mdl) used in this study along with two files (7%\_Oxygen\_ 2mM\_Nitrite.vdf and Measured\_Data) containing simulated and measured data, respectively. (ZIP)

### **Acknowledgments**

We wish to thank Lars Molstad, Senior Engineer, Norwegian University of Life Sciences, for his useful suggestions and assistance with mathematical formulations of the model.

#### **Author Contributions**

Conceived and designed the experiments: LLB LRB. Performed the experiments: LLB. Analyzed the data: JH LLB LRB. Contributed reagents/materials/analysis tools: IDW. Wrote the paper: JH LLB LRB. Constructed the model: JH IDW LRB. Analysed the simulation results: JH LRB. Edited and revised the text: JH LLB IDW LRB.

- Veening J-W, Smits WK, Kuipers OP (2008) Bistability, epigenetics, and bethedging in bacteria. Annu Rev Microbiol 62: 193–210.
- Shieh W, Lin Y, Jean W (2004) Pseudovibrio denitrificans gen. nov., sp. nov., a marine, facultatively anaerobic, fermentative bacterium capable of denitrification. Int J Syst Evol Microbiol 54: 2307–2312.
- Zumft WG, Kroneck PMH (1997) Respiratory transformation of nitrous oxide (N<sub>2</sub>O) to dinitrogen by bacteria and archaea. Adv Microb Physiol 52: 107–227.
- Højberg O, Binnerup SJ, Sørensen J (1997) Growth of silicone-immobilized bacteria cells on polycarbonate membrane filters, a technique to study microcolony formation under anaerobic conditions. Appl Environ Microbiol 63: 2920–2924.
- Bergaust L (2009) Regulatory biology of denitrification in Agrobacterium tumefaciens and Paracoccus denitrificans; responses to environmental controllers. Ås: Norwegian University of Life Sciences.
- Bouchal P, Struhárová I, Budinská E, Šedo O, Vyhlídalová T, et al. (2010) Unraveling an FNR based regulatory circuit in *Paracoccus denitrificans* using a proteomics-based approach. Biochimica et Biophysica Acta (BBA) - Proteins and Proteomics 1804: 1350–1358.
- Wood NJ, Alizadeh T, Bennett S, Pearce J, Ferguson SJ, et al. (2001) Maximal expression of membrane-bound nitrate reductase in *Paracoccus* is induced by nitrate via a third FNR-like regulator named NarR. J Bacteriol 183: 3606–3613.
- Bergaust L, van Spanning RJM, Frostegård Å, Bakken LR (2012) Expression of nitrous oxide reductase in *Paracoccus denitrificans* is regulated by oxygen and nitric oxide through FnrP and NNR. Microbiology 158: 826–834.
- Spiro S (2007) Regulators of bacterial responses to nitricoxide. FEMS Microbiol Rev 31: 193–211.
- Raj A, van Oudenaarden A (2008) Nature, nurture, or chance: stochastic gene expression and its consequences. Cell 135: 216–226.
- Qu Z (2014) Respiratory regulation in Paracoccus denitrificans and in soil, implications for N<sub>2</sub>O emissions. Ås, Norway: Norwegian University of Life Sciences.
- 23. Beckman JS, Koppenol WH (1996) Nitric oxide, superoxide, and peroxynitrite: the good, the bad, and ugly. Am J Physiol 271: C1424–C1437.

- Liu P-H, Svoronos SA, Koopman B (1998) Experimental and modeling study of diauxic lag of *Pseudomonas denitrificans* switching from oxic to anoxic conditions. Biotechnol Bioeng 60: 649–655.
- Lee D-U, Woo S-H, Svoronos S, Koopman B (2009) Influence of alternating oxic/anoxic conditions on growth of denitrifying bacteria. Water Res 44: 1819– 1824.
- Durvasula K, Jantama K, Fischer K, Vega A, Koopman B, Svoronos SA (2009)
   Effect of periplasmic nitrate reductase on diauxic lag of *Paracoccus pantotrophus*.
   Biotechnol Prog 25: 973–979.
- Lucking DR, Fraley RT, Kaplan S (1978) Intracytoplasmic membrane synthesis in synchronous cell populations of *Rhodopseudomonas sphaeroides*. Fate of "old" and "new" membrane. The Journal of Biological Chemistry 253: 451–457.
- Molstad L, Dörsch P, Bakken LR (2007) Robotized incubation system for monitoring gases (O<sub>2</sub>, NO, N<sub>2</sub>O, N<sub>2</sub>) in denitrifying cultures. J Microbiol Methods 71: 202–211.
- 29. Hannon B, Ruth M (2014) Modeling Dynamic Biological Systems; Ruth M, Hannon B, editors. New York: Springer
- de Gier J-WL, Lübben M, Reijnders WNM, Tipker CA, Slotboom D-J, et al. (1994) The terminal oxidases of *Paracoccus denitrificans*. Molecular Microbiology 13: 183–196.

- Pitcher RS, Watmough NJ (2004) The bacterial cytochrome cbb3 oxidases. Biochim Biophys Acta 1655: 388–399.
- Fukumori Y, Yamanaka T (1984) Two K<sub>m</sub> values for cytochrome c of aa<sub>3</sub>-type two-subunit cytochrome c oxidase from Nitrobacter agilis. FEBS Lett 170: 301– 304.
- Gates AJ, Luque-Almagro VM, Goddard AD, Ferguson SJ, Roldán MD, et al. (2011) A composite biochemical system for bacterial nitrate and nitrite assimilation as exemplified by Paracoccus denitrificans. Biochem J 435: 743– 753.
- Pan Y, Ni B-J, Yuan Z (2013) Modeling electron competition among nitrogen oxides reduction and N2O accumulation in denitrification. Environ Sci Technol Lett 47: 11083–11091.
- Hamilton R, Casasu A, Rasche M, Narang A, Svoronos SA, et al. (2005)
   Structured model for denitrifier diauxic growth. Biotechnol Bioeng 90: 501–508.
- Baumann B, Snozzi M, Zehnder AJB, van der Meer JR (1996) Dynamics of denitrification activity of Paracoccus denitrificans in continuous culture during aerobic-anaerobic changes. J Bacteriol 178: 4367–4374.
- 37. Wilhelm E, Battino R, Wilcock RJ (1977) Low-pressure solubility of gases in liquid water. Chem Rev 77: 219–262.