Metabolic engineering design to enhance (R,R)-2,3-butanediol production from glycerol in *Bacillus subtilis* based on flux balance analysis

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**Abstract**

**Background:** Glycerol is a desirable alternative substrate for 2,3-butanediol (2,3-BD) production for sustainable development in biotechnological industries and non-food competitive feedstock. *B. subtilis*, a "generally recognized as safe" organism that is highly tolerant to fermentation products, is an ideal platform microorganism to engineer the pathways for the production of valuable bio-based chemicals, but it has never been engineered to improve 2,3-BD production from glycerol. In this study, we aimed to enhance 2,3-BD production from glycerol in *B. subtilis* through in silico analysis. Genome-scale metabolic model (GSM) simulations was used to design and develop the metabolic pathways of *B. subtilis*. Flux balance analysis (FBA) simulation was used to evaluate the effects of step-by-step gene knockouts to improve 2,3-BD production from glycerol in *B. subtilis*.

**Results:** *B. subtilis* was bioengineered to enhance 2,3-BD production from glycerol using FBA in a published GSM model of *B. subtilis*, iYO844. Four genes, *ackA*, *pta*, *lctE*, and *mmgA*, were knocked out step by step, and the effects thereof on 2,3-BD production were evaluated. While knockout of *ackA* and *pta* had no effect on 2,3-BD production, *lctE* knockout led to a substantial increase in 2,3-BD production. Moreover, 2,3-BD production was improved by *mmgA* knockout, which had never been investigated. In addition, comparisons between in silico simulations and fermentation profiles of all *B. subtilis* strains are presented in this study.

**Conclusions:** The strategy developed in this study, using in silico FBA combined with experimental validation, can be used to optimize metabolic pathways for enhanced 2,3-BD production from glycerol. It is expected to provide a novel platform for the bioengineering of strains to enhance the bioconversion of glycerol into other highly valuable chemical products.

**Keywords:** Glycerol, 2,3-Butanediol, *Bacillus subtilis*, Flux balance analysis, Genome-scale metabolic models, Gene knockout

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**Background**

Biodiesel is a liquid biofuel derived from natural fats through a chemical reaction process (esterification or transesterification). Biodiesel has advantages over fossil diesel in terms of health and environmental concerns, such as lower sulfur and carbon dioxide emissions, which may contribute to a reduction in global warming. In 2019, 24% of the global carbon dioxide emissions from fossil fuel consumption were related to the transport section, and approximately 75% of the total carbon dioxide emissions from this section was from commercial vehicles [1]. Biodiesel not only provides the same high engine performance and calories as diesel, but it also has a higher flash point and better viscosity. These advantages have led to an increase in the use of biodiesel as a renewable fuel,
offering an innovative solution to reduce the global air pollution caused by the increasing number of vehicles in Asian countries [2].

Biodiesel production generates large amounts of crude glycerol as a by-product; every ton of biodiesel produced generates approximately 100 kg of crude glycerol. Despite the increasing purification capacity and the development of new applications, biodiesel production has been associated with increased glycerol generation over the past few decades. Hejna et al. [3] reported that approximately 1 million tons of crude glycerol are produced annually from biodiesel production in Europe. Between 2015 and 2017, crude glycerol production increased from 1.16 to 2 million tons in Europe, and its production is expected to increase to more than 5 million tons by 2026 worldwide [4]. The large surplus of waste glycerol has affected the refined glycerol market and caused a significant reduction in the glycerol price. Researchers have attempted to convert glycerol into more useful products, such as 1,3-propanediol, succinic acid, and ethanol, in order to reduce the cost of biodiesel production and the toxic effects of waste glycerol on the environment [5].

Currently, sugar, starch, and molasses are the major raw materials for commercial biological products, such as proteins and biochemical monomers. Recent studies have focused on the use of renewable and non-food competitive materials as alternative resources; glycerol is a promising substrate for use as a renewable and low-cost substrate for biological production platforms.

(R,R)-2,3-Butanediol (2,3-BD) is a valuable bio-based chemical compound that has a wide range of potential applications, such as in the manufacturing of printing inks, antifreeze agents, foodstuffs, chemicals, and pharmaceuticals [6]. In addition, 2,3-BD can be used as a chemical substance for the production of more applicable chemicals, such as 1,3-butanediene, which can serve as a raw material for synthetic rubbers (methyl-ethyl-ketone), acetoain, and diacetyl, which can be used as liquid fuel additives, etc. [7]. 2,3-BD has roles in high-value manufacturing in various industries, such as the cosmetics, agricultural, and pharmaceutical industries, and the bioprocess of 2,3-BD production has advantages over chemical processes in terms of environmental friendliness and lower cost [8]. Therefore, in this study, we focused on the conversion of glycerol into 2,3-BD using biological processes.

Among industrial microbes, Bacillus subtilis, a “generally recognized as safe” organism, is an ideal platform organism for engineering bacterial pathways for the biological production of various useful metabolites, such as lactate, acetate, acetoain, and 2,3-BD [9, 10]. B. subtilis is a superior host for bioproduction as compared to Escherichia coli because it is more tolerant to fermentation products, such as butanol and ethyl acetate [11], and lacks lipopolysaccharides, generally referred to as endotoxins, which are present in the outer cell membrane of E. coli and other gram-negative bacteria [12].

Bioinformatics tools have become increasingly attractive. Various techniques have been used to simulate and analyze biochemical systems. In silico genome-scale metabolic models (GSMs), which can be used to evaluate all metabolic reactions in a cell, are formulated on the basis of genome annotation data and experimental data. GSMs computationally describe gene-protein-reaction associations for a whole set of metabolic genes in an organism and can be used to simulate metabolic fluxes [13]. GSMs can predict metabolic flux using mathematical optimization techniques, such as flux balance analysis (FBA), through linear programming. Comparison of the predicted metabolic flux distributions under different conditions allows the evaluation of metabolic responses to identify the optimal conditions for a certain objective, such as growth rate [14]. FBA can also be applied to steady-state analysis of a given system using a stoichiometric matrix. When the objective function is resolved, the steady-state flux distribution can be obtained by solving a set of equations and is used to interpret the capacities of the metabolic system [15]. Several databases, including BsubCyc, KEGG, and NCBI, provide detailed information on the genome, genes, proteins, and metabolic and regulatory pathways of B. subtilis. Thus, in silico simulation with FBA is available for efficient metabolic pathway design of B. subtilis for glycerol assimilation and 2,3-BD production in a complex metabolic system. In preliminary experiments, we found that the B. subtilis model described by Oh et al. [16], iYO844 GSM, showed the most reliable performance for the simulation of growth on glycerol.

Previous bioengineering research on B. subtilis has focused mostly on improving the production of 2,3-BD from sugars such as glucose [17–19], not from glycerol. Only a few bioengineering studies have used glycerol for 2,3-BD production, e.g., in Klebsiella oxytoca [20], Klebsiella pneumoniae [21], and Bacillus amyoliquefaciens [22]. Bioengineering B. subtilis to improve 2,3-BD production from glycerol represents an alternative for applications in biorefineries. The iBsu1147 metabolic network model of B. subtilis has been used to generate an in silico metabolic engineering design of reactions to increase the production of 2,3-BD from glucose [23]; however, the design was not experimentally validated. To the best of our knowledge, this is the first study to use in silico FBA simulation to improve 2,3-BD production from glycerol in B. subtilis and evaluate the effects of step-by-step gene knockouts by validating the in silico design with experimental data.
The aim of this study was to enhance 2,3-BD production from glycerol in *B. subtilis* using in silico analysis to identify gene knockout targets in order to improve the metabolic pathway. A prototype of the metabolic pathways of *B. subtilis* for 2,3-BD production from glycerol was designed using the iYO844 model, considering growth in M9 medium. Changes in the bio-fermentation profiles caused by four gene knockouts are presented, and fermentation products were periodically collected and analyzed. Furthermore, the in silico analysis results were compared with experimental fermentation profiles.

**Results and discussion**

**In silico simulation of 2,3‑BD production from glycerol by *B. subtilis* strains**

We modified and optimized the *B. subtilis* model iYO844 [16], which in a preliminary study provided the most reliable results for growth on glycerol, for in silico prediction of 2,3-BD production using FBA. Briefly, the environmental conditions in the iYO844 model were modified to the use of M9 medium with glycerol and an oxygen consumption rate of 10 mmol/gCDW/h, and biomass flux was the objective function. The regulation of glycerol uptake and metabolism in the model was optimized using the glycerol kinase (glpK) pathway (see “Methods” section). The primary results of glycerol utilization and metabolic flux for 2,3-BD production are shown in Fig. 1A. The biomass flux was 0.3 h⁻¹, and acetic acid and lactic acid were the major fermentation products (3.65 and 2.0 mmol/gCDW/h). Fluxes of other volatile fatty acids (VFAs) were not observed.

To test the inhibitory effect of acetic acid production, the metabolic fluxes of the ACKr and PTAr reactions, which refer to the acetate kinase (encoded by ackA) and phosphate acetyltransferase (encoded by pta) reactions, were set to zero (Fig. 1B; Table 5). Thus, the flux of the PTAr reaction became zero, which led to zero flux in the ACKr reaction. However, a negligible amount of acetate (~0.12 mmol/gCDW/h) was produced from the essential metabolic reaction in iYO844 model, which is catalyzed by *N*-acetyldiaminopimelate deacetylase (BSU14190; abbreviation in the model: ADPDA) (see Additional file 1). The specific growth rate insignificantly decreased to 0.286 h⁻¹. Lactic acid production slightly...
increased to 2.14 mmol/gCDW/h, acetoacetate production (ACACT1r and OCOAT1 reactions) increased from 0 to 1.86 mmol/gCDW/h, and the fluxes of other fermentation products were not affected. The results from this simulation showed that inactive ACKr and PTAr reactions did not promote or contribute to 2,3-BD production.

Next, we inactivated the LDH_L reaction to evaluate the effect of L-lactate dehydrogenase gene (lctE) deletion (Fig. 1C). The results showed that there was no change in the specific growth rate (0.286 h⁻¹), whereas the LDH_L reaction flux became zero. The flux of acetoacetate production decreased to 1.5 mmol/gCDW/h. Interestingly, 2,3-BD production became a major by-product (1.43 mmol/gCDW/h), whereas the production of other VFAs remained zero. The results of this simulation clearly showed that suspension of the LDH_L reaction can enhance 2,3-BD production from glycerol in B. subtilis.

From the simulation results of the three inactivated reactions, we inferred that the major fermentation products were acetoacetate and 2,3-BD. Since the target of this study was 2,3-BD production alone, we had to eliminate the metabolic flux of acetoacetate production. The conversion of pyruvate into acetoacetate involves three reactions catalyzed by pyruvate dehydrogenase (PDH), acetyl-CoA C-acetyltransferase (ACACT1), and 3-oxoacid CoA-transferase (OCOAT1). The PDH reaction is an essential metabolic reaction that cannot be disabled, and the OCOAT1 reaction involves multiple genes, including BSU38990 and BSU38980. Therefore, the reaction catalyzed by ACACT1, encoded by mmaA, was a suitable target for inactivation (Fig. 1D). Growth was nearly unchanged (0.283 h⁻¹), and the metabolic fluxes of the OCOAT1 and ACACT1 reactions became zero. As expected, the PDH flux was reduced by about half, and 2,3-BD production increased to 2.54 mmol/gCDW/h. Importantly, 2,3-BD was the only major product at this stage, whereas other products, such as ethanol, diacetyl, and propionic acid, were not detected. In the next phase of our study, the order of the simulations was applied to step-by-step gene modification in B. subtilis to validate the hypothesis from these simulations.

Hao et al. [23] used a B. subtilis GSM (iBSu1147) and flux variability analysis simulations to identify target reactions of genes to overexpress or underexpress to allow 2,3-BD production from glucose. The micro-aerobic condition was applied in the simulation (0.0005 mmol/gCDW/h), referring to the constrained oxygen condition at 0.5% saturated dissolution. Candidate reactions were predicted by using the flux variability analysis. The simulation results suggested that the first four genes to underexpress were pyc (encoding pyruvate carboxylase), ytiB or yvdA (encoding putative carbonic anhydrase), ndk (encoding nucleoside diphosphate kinase), and gmk (encoding guanylate kinase). These four genes are entirely different from the target genes and not related to those of the simulation for gene deletion in this study. Also, the biomass flux was zero when FBA simulation was conducted with knockout of these four genes in the iYO844 model. The difference in the prediction results may be due to the use of different models, environmental conditions, and prediction methods.

**Effects of gene knockouts on 2,3-BD production from glycerol in B. subtilis strains**

Fermentation experiments with four B. subtilis step-by-step knockout strains (W168, W168Δ2, W168Δ3, and W168Δ4) were performed to validate the predictions based on the GSM simulation. The growth, glycerol utilization, and fermentation characteristics of each strain are shown in Fig. 2. The specific growth rate of W168 was 0.2 h⁻¹, and the glycerol consumption rate was 7.26 mmol/gCDW/h. The fermentation profile of W168 (Fig. 2A) showed that the major fermentation products were acetic acid, lactic acid, and 2,3-BD. Acetic and lactic acid production corresponded with growth, whereas acetoin and 2,3-butanediol production occurred during the stationary phase. Diacetyl, ethanol, pyruvic, and succinic acid production was not detected throughout the experiment. This result was similar to previous results of fermentation with B. subtilis W168 in M9 medium containing 10 g/L glucose under micro-aerobic conditions, in which 2,3-BD, acetoin, acetic acid, and lactic acid were the major fermentation products, succinate and ethanol accumulation were negligible, and pyruvic acid, α-ketoglutaric acid, and diacetyl were not detected throughout the fermentation process [24].

Figure 2B shows the effects of ackA and pta double knockout on growth, glycerol utilization, and VFAs production. When compared to W168, acetic acid production was delayed by 4 h in W168Δ2; W168 produced acetic acid after 16 h of incubation, whereas W168Δ2 produced acetic acid after 20 h of incubation. However, the major fermentation products were similar to those of W168, and acetic acid was still produced in significant amounts. In line herewith, Fu et al. [19] reported that pta inactivation hardly reduced acetate production in B. subtilis, which may be because of other pathways except acetate or acetyl-phosphate synthesis in B. subtilis. Besides acetic acid production by N-acetylami

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which correspond to metabolic reaction of *B. subtilis* 168 in the Kyoto Encyclopedia of Genes and Genomes (KEGG; http://www.genome.jp/kegg/). Moreover, there is an indirect pathway of *B. subtilis* 168 in which pyruvate is converted into acetic acid via pyruvate oxidase (*ydaP*, BSU04340, EC:1.2.3.3) and acylphosphatase (*yflL*, BSU07640, EC:3.6.1.7).

The specific growth rate of W168Δ2 was not different from that of W168 (0.2 h⁻¹), but the glycerol consumption rate was higher (8.7 mmol/gCDW/h). In addition, some VFAs were produced at slightly lower final concentrations than by W168, which may be explained by the decrease in glycerol consumption after the exponential phase. These results indicated that knockout of *ackA* and *pta* did not improve 2,3-BD production (Table 1). The 2,3-BD yield of W168Δ2 after 72 h of incubation was not increased compared to that of the wild-type strain.

The effects of *lctE* and *mmgA* deletion are shown in Fig. 2C and D. The specific growth and glycerol consumption rates of the W168Δ3 and W168Δ4 strains were similar to those of the W168Δ2 strain. Thus, knockout of *lctE* and *mmgA* has no effect on the growth and glycerol uptake rates in *B. subtilis*. VFAs indicated that no significant amounts of diacetyl, ethanol, pyruvic, and succinic acid were produced, similar to the results for the wild-type and W168Δ2 strains. *lctE* knockout caused

| Strain | VFAs* (unit: mol/mol glycerol) | Glycerol Utilization (mol) |
|--------|--------------------------------|--------------------------|
|        | Acetic acid | Lactic acid | Acetoin | 2,3-BD |                                      |
| W168   | 0.07 ± 0.014 | 0.07 ± 0.011 | 0.09 ± 0.009 | 0.16 ± 0.004 | 0.09 ± 0.002 |
| W168Δ2 | 0.09 ± 0.005 | 0.08 ± 0.011 | 0.06 ± 0.002 | 0.15 ± 0.013 | 0.08 ± 0.004 |
| W168Δ3 | 0.12 ± 0.001 | 0.00 ± 0.000 | 0.02 ± 0.002 | 0.22 ± 0.004 | 0.09 ± 0.002 |
| W168Δ4 | 0.14 ± 0.015 | 0.00 ± 0.000 | 0.03 ± 0.011 | 0.23 ± 0.019 | 0.09 ± 0.007 |

*The concentrations of diacetyl, ethanol, pyruvic, and succinic acid were < 0.01 mol/mol glycerol*
inhibition of total lactic acid production and significantly suppressed acetoin production. Acetic acid was gradually produced during the stationary phase. Interestingly, lctE deletion had a noticeable favorable effect on 2,3-BD production; the 2,3-BD yield of W168Δ3 reached 0.22 mol/mol glycerol, whereas that of W168 and W168Δ2 was 0.16 and 0.15 mol/mol glycerol, respectively. These results were in line with the results of the in silico simulations, except for acetic acid production flux, which is discussed in the next section.

When considering the metabolic pathway of glycerol dissimilation to 2,3-BD, we found that the L-lactate dehydrogenase (BSU03050, lctE) reaction might compete with the (R,R)-butanediol dehydrogenase (BSU06240; bdhA) reaction for NADH, which is mainly produced by glycerol-3-phosphate dehydrogenase (BSU22830; glpD) and glyceraldehyde-3-phosphate dehydrogenase (BSU33940; gapA). Hence, the deletion of lctE eliminates the NADH competitive reaction, which results in an increase in 2,3-BD production from glycerol. In addition, lctE deletion has been reported to increase the production of 2,3-BD in other platforms, including B. subtilis [19], K. oxytoca [25], Enterobacter cloacae [26], and K. pneumoniae [27]. Therefore, lctE deletion can be regarded as key in improving 2,3-BD production in many microorganisms.

Remarkably, mmgA knockout caused an increase in 2,3-BD production. There is no previous study on the deletion of mmgA to enhance 2,3-BD production. The 2,3-BD yield of W168Δ4 was 0.23 mol/mol glycerol. The product of mmgA, acetyl-CoA C-acetyltransferase (BSU24170, EC:2.3.1.9), is involved in the conversion of acetyl-CoA into acetoacetyl-CoA, which is a precursor of acetoacetate production for pyruvate and butanoate metabolism. Thus, mmgA deletion leads to the suppression of the metabolic flux from pyruvate to acetyl-CoA, which may contribute to an increase in the fluxes associated with 2,3-BD production. We conducted a micro-aerobic fermentation experiment to ensure the promotive effect of mmgA knockout on 2,3-BD production from glycerol. Under the same conditions, W168Δ3 produced 1.93 ± 0.10 g/L 2,3-BD, whereas W168Δ4 produced 2.13 ± 0.05 g/L after 96 h of incubation. This result confirmed that mmgA knockout contributed to the increase in 2,3-BD production.

The theoretical yield of the conversion of glycerol into 2,3-BD is 0.5 mol/mol glycerol [28]. The 2,3-BD yield of W168 strains were shown in Table 1. The maximum yield of 2,3-BD production by W168Δ4 (0.23 mol/mol glycerol) was 46% from the theoretical yield. The 2,3-BD yield of W168Δ4 was reduced by acetic acid production.

Comparison of the FBA results with experimental data

The specific growth rate (µ, h⁻¹) and glycerol consumption rate (ν, mmol/gCDW/h) derived from the experimental data of the B. subtilis strains were applied in the in silico FBA simulation to determine the oxygen consumption rate (Table 2, Additional file 1). We found that the oxygen consumption rate did not significantly differ among the strains, demonstrating the similarity of the environmental conditions in the experiment.

In the FBA, the highest acetic production rate was obtained for the W168 strain, which corresponded to the experimental result of acetic production of W168 during the exponential phase. However, in the FBA, after inactivation of the ACKr and PTAr reactions, the acetic

| Parameter (unit: mmol/gCWD/h) | B. subtilis strains |
|-------------------------------|------------------|
|                               | W168      | W168Δ2    | W168Δ3    | W168Δ4    |
| Glycerol consumption (ν)*      | –7.26     | –8.70     | –8.70     | –8.70     |
| B. subtilis biomass (µ, h⁻¹)*  | 0.20      | 0.20      | 0.20      | 0.20      |
| Oxygen consumption             | –7.27     | –7.35     | –7.35     | –7.40     |
| Acetate exchange               | 2.83      | 0.09      | 0.09      | 0.09      |
| L-Lactate exchange             | 1.62      | 3.70      | 0.00      | 0.00      |
| (R,R)-2,3-butanediol exchange  | 0.00      | 0.00      | 2.47      | 2.77      |

* Experimental data
production rate became 0.09 mmol/gCDW/h, which was not in line with the experimental findings after ackA and pta knockout. Although acetic production during the exponential phase of these strains was low, acetic acid production did occur during the stationary phase, likely because of the existence of alternative pathways of synthetic acetic acid in B. subtilis. The iYO844 model contains many reactions related to acetic production, but the FBA results of these reaction fluxes, including AGDC (N-acetylglucosamine-6-phosphate deacetylase), AHSERL3 (cysteine synthase), and ALDD2x (aldehyde dehydrogenase), were zero. Moreover, other reactions involved in acetic production, such as those catalyzed by pyruvate oxidase (BSU04340, EC:1.2.3.3) and acylphosphatase (BSU07640, EC:3.6.1.7), are not represented in this model. While we found that many reactions are missing the iYO844, this model provided the most reliable result of biomass production from glycerol, as mentioned above. Therefore, further development and improvement of the B. subtilis model are necessary to allow for more accurate in silico simulations in future. To date, substantial research has focused on reconstructing GSMs of B. subtilis as a representative model for other gram-positive bacteria [13].

The specific growth rates of the mutant strains were not different from that of W168 (0.2 h⁻¹), whereas the glycerol consumption rates were higher (8.7 mmol/gCDW/h). However, the final concentrations of some of the major VFAs were slightly lower for W168Δ2 than for W168, which may have been due to the decrease in glycerol consumption after the exponential phase. For ackA and pta knockout in W168Δ2, the FBA results predicted that the fluxes of 2,3-BD production were zero. Indeed, knockout of these two genes did not affect 2,3-BD production (Table 2); the 2,3-BD production rates for W168 and W168Δ2 were 0.61 and 0.60 mmol/gCDW/h, respectively.

FBA predicted that in W168Δ3, the flux of lactic acid production would be zero as the LDH_L reaction was inactivated. Indeed, we found that lactic acid production was inhibited by lctE knockout. Moreover, the flux of 2,3-BD production increased to 2.47 mmol/gCDW/h in W168Δ3, which was substantially higher than that of W168Δ2 (Table 1). To assess the effect of mmgA knockout, FBA results of W168Δ3 and W168Δ4 on 2,3-BD production were compared. The FBA results predicted that the 2,3-BD production rate of W168Δ4 would be slightly greater than that of W168Δ3, which was in line with the experimental result; the 2,3-BD production rate of W168Δ4 was 1.06 mmol/gCDW/h, whereas that of W168Δ3 was 0.96 mmol/gCDW/h. We found that acoA (acetoin dehydrogenase, BSU08060) was knocked out in previous studies in order to enhance 2,3-BD production in B. subtilis [19, 24]. However, the FBA simulation in this study did not suggest an ethanol production flux from this reaction. Moreover, the fermentation results of B. subtilis W168 in M9 medium containing glucose or glycerol did not indicate ethanol to be a significant product. Therefore, knockout of acoA was not required. This demonstrates that the use of in silico simulation can save time, energy, labor, and costs when compared to trial-and-error experiments.

Although the FBA results were not exactly the same as the experimental results, the guidance of gene knockout based on in silico simulations was very accurate, and the trends in the simulation results were consistent with the experimental results. 2,3-BD production from glycerol by the modified B. subtilis strains was substantially higher than that of the wild-type strain. The promotive effect of mmgA knockout on 2,3-BD production from glycerol, which was predicted based on the FBA simulation results, was demonstrated for the first time in this study.

Conclusions

Glycerol is a renewable and low-cost substrate that has potential as a raw material for industrial biological production processes. 2,3-BD is a valuable bio-based chemical compound for a wide range of potential applications and was selected as the target product in this study. The iYO844 genome-scale metabolic model of B. subtilis was applied to design and develop the metabolic pathways. FBA results provided precise target reactions to guide gene knockouts to improve the metabolic pathway for 2,3-BD production from glycerol. Furthermore, the use of in silico simulation of FBA could provide a novel platform solution for the bioengineering of B. subtilis to enhance 2,3-BD production from glycerol, which was the guidance of mmgA knockout. Additionally, the prediction results can save time and costs from unnecessary gene knockouts. The methodology of this study can be applied to increase other target fermentation products through the use of FBA in GSMs of other microorganisms. We are currently focusing on the development of GSMs and prediction strategies to provide better simulation results in future studies. The experimental knowledge and updated information derived from these experiments can be used to develop accurate and efficient models and predictions.

Methods

Strains and plasmid construction

The strains and plasmids used in this study are listed in Table 3. The parent strain, B. subtilis wild-type 168 (W168), was obtained from the Microbe Division at the RIKEN BioResource Research Center, Japan. We used the CRISPR-Cas9 system for gene knockout in B. subtilis [30]. Plasmid pJOE8999 was obtained from the Bacillus
Table 3 Strains and plasmids used in this study

| Strain or plasmid | Genotype or relevant characteristic | Reference |
|-------------------|-------------------------------------|-----------|
| **Strains**       |                                     |           |
| E. coli Nova blue | endA1 hsdR17(rK2− mK2+ +) supE44 thi-1 gyrA96 relA1 lac recA1/F[proAB + lacIq ZDM15::Tn10(TetR)] | Novagen   |
| B. subtilis W168  | trpC2                               | JMC       |
| B. subtilis W168Δ2 | trpP2ackAΔpta                       | This study |
| B. subtilis W168Δ3 | trpP2ackAΔpta lctE                   | This study |
| B. subtilis W168Δ4 | trpP2ackAΔpta lctE mmmA             | This study |
| **Plasmids**      |                                     |           |
| pJOE8999          | KanR, PmamP.ackA, lacPOZ− gRNA, oop ter, T7P, repP194, repC, P Cori | Altenbuchner [30] |
| pJOE_ackA         | pJOE8999 derivative, containing 20-nt spacer targeting ackA and fused homologous arms of ackA | This study |
| pJOE_pta          | pJOE8999 derivative, containing 20-nt spacer targeting pta and fused homologous arms of pta | This study |
| pJOE_lctE         | pJOE8999 derivative, containing 20-nt spacer targeting lctE and fused homologous arms of lctE | This study |
| pJOE_mmmA         | pJOE8999 derivative, containing 20-nt spacer targeting mmmA and fused homologous arms of mmmA | This study |

*Microbe Division/Japan Collection of Microorganisms (JMC), RIKEN Bioresource Research Center

Genetic Stock Center (Ohio State University, Columbus, OH, USA). For plasmid construction, we used Escherichia coli NovaBlue (Novagen, Cambridge, MA, USA).

For plasmid construction for ackA gene knockout (pJOE_ackA), plasmid pJOE8999 was cut at the BsaI restriction site and purified by gel purification (Wizard SV Gel and PCR Clean-up System Kit, Promega, WI, USA). The primers used in this study are listed in Table 4. An oligonucleotide encoding a 20-nucleotide sgRNA with flanking BsaI sites (P_ackA01 and P_ackA02 primers) was introduced into pJOE8999 by polymerase chain reaction (PCR) using KOD FX Neo DNA polymerase (Toyobo, Osaka, Japan). The PCR product was digested with DpnI restriction enzyme to remove the methylated DNA templates and gel-purified. Plasmid pJOE8999 containing a 20-nucleotide sgRNA for ackA deletion was cut using the SfiI restriction enzyme and purified. Meanwhile, two homology arms for ackA deletion were PCR-amplified from B. subtilis W168 chromosomal DNA using the P_ackA03 and P_ackA04, and P_ackA05 and P_ackA06 primers. After PCR product purification, these two PCR fragments were combined by overlap extension PCR using the P_ackA03 and P_ackA06 primers. The overlap extension PCR fragment was integrated into the SfiI site of the vector using NEBuilder HiFi DNA Assembly Master Mix (New England Biolabs, Ipswich, MA, USA), and transformed into E. coli NovaBlue for plasmid construction and replication. Transformants were selected on 25 μg/mL kanamycin in Luria–Bertani (LB) medium. pJOE_pga, pJOE_lctE, and pJOE_mmmA were constructed using the same protocol as that used for pJOE_ackA plasmid construction.

Plasmids pJOE_ackA, pJOE_pga, pJOE_lctE, and pJOE_mmmA were sequentially introduced into B. subtilis W168 by natural transformation [31] for multiple gene deletions in the order of the in silico simulation. Cells were plated on LB agar containing 5 μg/mL kanamycin and 0.2% mannose to introduce cas9 under the control of P_mamP. The plate was incubated at 30 °C for 2 days, and colonies were selected and transferred to LB agar without antibiotics and incubated at 50 °C overnight. The replication of pJOE8999 derivatives was temperature-sensitive, so the plasmid was removed during this process. The colonies were streaked on LB agar and incubated at 42 °C to obtain single colonies. The colonies were checked for target gene deletion by PCR using the above primers; for example, P_ackA07 and P_ackA08 primers were used to check for ackA deletion. Loss of the plasmid was evaluated by plating single colonies onto LB agar with and without kanamycin. Finally, target gene deletion was confirmed using DNA sequencing.

**Culture media and fermentation conditions**

LB liquid medium contains (per L) 10.0 g tryptone, 5.0 g yeast extract, and 10.0 g sodium chloride, and for plates, 1.5% bacto-agar is added for solidification. LB medium was used for the maintenance and preservation of all strains, and for plasmid construction. All strains were preserved at −80 °C and revived by inoculation on LB agar plates.

M9 medium [32] containing 10 g/L of glycerol was used as the minimal medium to investigate the bio-fermentation profile of all B. subtilis strains. Colonies of B. subtilis strains were cultured in 5 mL of M9 medium at 37 °C under rotation at 180 rpm overnight. Then, 1% (v/v) of the culture medium was activated by culture in 5 mL of M9 medium at 37 °C and 180 rpm for 18 h. Subsequently, 0.5% (v/v) liquid culture was inoculated into a 300-mL flask with 150 mL of M9 medium at 37 °C and 180 rpm for 72 h. Bacterial growth was periodically observed by
measuring the optical density of the liquid culture at 600 nm (OD$_{600}$). At the same time, liquid samples were collected and analyzed for residual glycerol and fermentation products, including acetic acid, lactic acid, pyruvic acid, succinic acid, diacetyl, acetoin, 2,3-BD, and ethanol. To confirm the effect of $mmgA$ knockout on 2,3-BD production, micro-aerobic fermentation was performed using 120 mL of M9 medium at 37 °C and 100 rpm for 96 h.

**Analytical methods**
The growth of *B. subtilis* strains was periodically observed by measuring the OD$_{600}$ using a UVmini-1240 spectrophotometer (Shimadzu, Kyoto, Japan). The cell dry weight (CDW) was converted from the OD$_{600}$ value; 1 OD$_{600}$ is equal to 0.325 g CDW/L [19]. The concentrations of glycerol and VFAs (acetic acid, lactic acid, pyruvic acid, succinic acid, diacetyl, acetoin, 2,3-BD, and ethanol) in the culture supernatant were monitored using a high-performance liquid chromatography instrument (Shimadzu, Japan) equipped with a UV/RI detector and an Aminex HPX-87H column (300 × 7.8 mm. (Bio-Rad, USA)). A 10-μL sample was injected into the chromatography system operated at 55 °C and was eluted with 5 mM H$_2$SO$_4$ at a flow rate of 0.5 mL/min.

**In silico simulations**
The GSM of *B. subtilis* strain 168 in this study referred to the iYO844 model constructed by Oh et al. [16]. This model was updated on October 31, 2019, and is available.

| Table 4 Primers used in this study |
|------------------------------------|
| **Name**  | **Sequence (5′ → 3′)** | **Usage** |
| P_ackA01  | CGGTATCGGTGAAAAACAGTGTTTATAGCTAGAAATAGCAAGTT | Construction of pJOE_ackA sgRNA |
| P_ackA02  | GAGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_ackA spacer |
| P_ackA03  | GGCTGACGGGCAACAGGAGGCCTCCTACGCGGATATGGAATAGCAAGTT | Construction of pJOE_ackA spacer |
| P_ackA04  | TCAAGAAGATGTCGCCCTTATGCAGATGGTGCGTCCCTTATATCTCTGA | Construction of pJOE_ackA deletion |
| P_ackA05  | TCAAGAAGATGTCGCCCTTATGCAGATGGTGCGTCCCTTATATCTCTGA | Construction of pJOE_ackA deletion |
| P_ackA06  | ATGACAGAGGATGTCGCCCTTATGCAGATGGTGCGTCCCTTATATCTCTGA | Construction of pJOE_ackA deletion |
| P_ackA07  | CGACGGAAGATCACAAGCTCCCTAG | PCR of check ackA deletion |
| P_ackA08  | ACTGGCAGATCACAAATAGCAGTG | PCR of check ackA deletion |
| P_pta01   | GAATGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_pta spacer |
| P_pta02   | GAGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_pta spacer |
| P_pta03   | GGCTGACGGGCAACAGGAGGCCTCCTACGCGGATATGGAATAGCAAGTT | Construction of pJOE_pta spacer |
| P_pta04   | GAGCTGACCATGTCGCCCTTATATCTCTCA | Construction of pJOE_pta spacer |
| P_pta05   | TTGACCATATTATTTAAAACCTCCCTCAAAATAGTACAA | Construction of pJOE_pta spacer |
| P_pta06   | GGCACAAAAGCTGATTTATTGGGCAATAGCAGGTTTTCCTAG | Construction of pJOE_pta spacer |
| P_pta07   | GTCCGGGAGCCCTTCTTCTCT | PCR of check pta deletion |
| P_pta08   | GAGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | PCR of check pta deletion |
| P_lctE01  | GATGATGATTAAAACAGGTTTATAGCTAGAAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE02  | GAGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE03  | GGCTGACGGGCAACAGGAGGCCTCCTACGCGGATATGGAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE04  | GAAACATACCTGGAAGAGGTGACGCGCAATTATAGCTAGAAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE05  | GAAACATACCTGGAAGAGGTGACGCGCAATTATAGCTAGAAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE06  | GAAACATACCTGGAAGAGGTGACGCGCAATTATAGCTAGAAATAGCAAGTT | Construction of pJOE_lctE spacer |
| P_lctE07  | TGAGCTGACCTTTATCTCTAGATGATAGTCG | PCR of check lctE deletion |
| P_lctE08  | TGCTGGAGCCCGGAATCAGGAACTTACG | PCR of check lctE deletion |
| P_mmgA01  | CGGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_mmgA spacer |
| P_mmgA02  | GAGAATGTGAAATATTGACCTACGCGGATGGGATATGGAATAGCAAGTT | Construction of pJOE_mmgA spacer |
| P_mmgA03  | GGCTGACGGGCAACAGGAGGCCTCCTACGCGGATATGGAATAGCAAGTT | Construction of pJOE_mmgA spacer |
| P_mmgA04  | TTTCCACTCCCTTATATCTAGTGGTTTATACCTCTACTCAGAAGATAAG | Construction of pJOE_mmgA spacer |
| P_mmgA05  | TTTCCACTCCCTTATATCTAGTGGTTTATACCTCTACTCAGAAGATAAG | Construction of pJOE_mmgA spacer |
| P_mmgA06  | TTTCCACTCCCTTATATCTAGTGGTTTATACCTCTACTCAGAAGATAAG | Construction of pJOE_mmgA spacer |
| P_mmgA07  | GGCTGACCATGTCGCCCTTATATCTCTGA | PCR of check mmgA deletion |
| P_mmgA08  | ATCAGGCTCTTACCATCCGCTCCG | PCR of check mmgA deletion |
on the BiGG Model website (http://bigg.ucsd.edu/). The last version of the iYO844 model contained 990 metabolites, 1250 metabolic reactions, and 844 genes. For simulations to determine the metabolic flux distributions, FBA was performed with the commercial GLPK (GNU Linear Programming Kit) package, which supplies a solver for large-scale linear programming (LP) and mixed integer programming (MIP), and the COBRA toolkit [33] using MATLAB software (MathWorks, Inc., USA). M9 minimal medium for *B. subtilis* was applied by setting a lower bound to allow unlimited uptake on the exchange reactions of the necessary minerals. The ALCD19_D and ALCD19_L reactions, which involve alcohol dehydrogenases, were deleted from the model because of the determined roles of YhdN in methylglyoxal resistance [34].

To evaluate the growth and metabolic pathways of glycerol utilization as the sole substrate in the simulation of *B. subtilis* strains under limited oxygen conditions, the glycerol and oxygen consumption rates were set at 10 mmol/gCDW/h (the maximum uptake rate), and the biomass objective function was set to maximize the target product for the growth simulation. Gene knockout simulations were performed by setting the minimum and maximum fluxes of the related reactions of the target genes to zero. The available reactions for gene knockout did not include exchange reactions, transporter reactions, or essential gene reactions. Details of the gene targets for knockout are listed in Table 5. W168 corresponds with the *B. subtilis* wild-type strain 168, which was originally used in the iYO844 model. The modified strains correspond with the gene knockout strains of W168: W168Δ2 (deletion of the ACKr and PTAr reactions), W168Δ3 (deletion of the ACKr, PTAr, and LDH_L reactions), and W168Δ4 (deletion of the ACKr, PTAr, LDH_L, and ACACT1r reactions).

The specific cell growth rate (\( \mu, \text{h}^{-1} \)) and glycerol consumption rate (\( \nu, \text{mmol glycerol/gCDW/h} \)) were evaluated from experimental data during the logarithmic phase using Eqs. 1 and 2, respectively. The necessary data for these calculations are cell density (\( X, \text{g CDW/L} \)), residual glycerol concentration (\( S, \text{mmol/L} \)), and time (\( t, \text{h} \)). The calculation results were applied to the FBA simulation for comparison.

\[
\mu = \frac{\ln X_1 - \ln X_0}{t_1 - t_0} \tag{1}
\]

\[
\nu = \frac{S_1 - S_2}{\frac{1}{2}(X_2 + X_1) \times (t_2 - t_1)} \tag{2}
\]

After deriving the experimental results of the specific cell growth and glycerol consumption rates, FBA simulations of each strain were performed again to compare the in silico results with the experimental results. The biomass flux was used as an objective function. Glycerol consumption rates were fixed according to the experimental results of each strain, and oxygen consumption rates were varied to obtain the specific cell growth rates based on the experimental results. The production of the target VFAs was compared between the experimental results and FBA simulation results.

**Abbreviations**

2,3-BD: (R,R)-2,3-butanediol; CDW: Cell dry weight; FBA: Flux balance analysis; GSMs: Genome-scale metabolic models; LB: Luria–Bertani; PCR: Polymerase chain reaction; sgRNA: Single-guide RNA; VFAs: Volatile fatty acids.

**Supplementary Information**

The online version contains supplementary material available at https://doi.org/10.1186/s12934-021-01688-y.

**Additional file 1.** FBA analysis results from the experimental data.

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**Authors’ contributions**

NV performed the experiments; NV, TS, and AK designed the study; NV wrote the paper. All authors read and approved the final manuscript.

| Table 5 | Information on target genes for knockout in the iYO844 model |
| --- | --- |
| Gene-protein-reaction | Gene | Enzyme | Abbreviation in model | EC number |
| BSU29470 | ackA | Acetate kinase | ACKr | 2.7.2.1 |
| BSU37660 | pta | Phosphate acetyltransferase | PTAr | 2.3.1.8 |
| BSU03050 | lctE | L-lactate dehydrogenase | LDH_L | 1.1.1.27 |
| BSU24170 | mmgA | Acetyl-CoA C-acetyltransferase | ACACT1r | 2.3.1.9 |

\[ \text{a Data from the National Center for Biotechnology Information (NCBI), USA refers to the sequence: NC_000964.3} \]

\[ \text{b iYO844 model referred to Oh et al. [16]} \]
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