Mineral Transformation and Biomass Accumulation Associated With Uranium Bioremediation at Rifle, Colorado

LI LI,* 1  CARL I. STEEFEL,†  KENNETH H. WILLIAMS,‡  MICHAEL J. WILKINS,¶  AND  SUSAN S. HUBBARD†

Earth Sciences Division, Lawrence Berkeley National Laboratory, 1 Cyclotron Road, MS 90-1116, Berkeley, California 94720, and Department of Earth and Planetary Science, University of California, Berkeley, California 94720

Received January 4, 2009. Revised manuscript received May 21, 2009. Accepted May 27, 2009.

Injection of organic carbon into the subsurface as an electron donor for bioremediation of redox-sensitive contaminants like uranium often leads to mineral transformation and biomass accumulation, both of which can alter the flow field and potentially bioremediation efficacy. This work combines reactive transport modeling with a column experiment and field measurements to understand the biogeochemical processes and to quantify the biomass and mineral transformation/accumulation during a bioremediation experiment at a uranium contaminated site near Rifle, Colorado. We use the reactive transport model CrunchFlow to explicitly simulate microbial community dynamics of iron and sulfate reducers, and their impacts on reaction rates. The column experiment shows clear evidence of mineral precipitation, primarily in the form of calcite and iron monosulfide. At the field scale, reactive transport simulations suggest that the biogeochemical reactions occur mostly close to the injection wells where acetate concentrations are highest, with mineral precipitate and biomass accumulation reaching as high as 1.5% of the pore space. This work shows that reactive transport modeling coupled with field data can be an effective tool for quantitative estimation of mineral transformation and biomass accumulation, thus improving the design of bioremediation strategies.

Introduction

Uranium bioremediation involves injection of organic carbon into the subsurface to stimulate indigenous bacteria that reduce soluble U(VI) to insoluble U(IV) (1). Depending on the presence of electron acceptors in situ, a series of biogeochemical reactions can occur, with their products further reacting to form mineral precipitates such as iron sulfide and calcite (2) and low solubility gases such as methane (3). In addition, stimulated growth of bacteria leads to biogeochemical reaction network, which was then used to match observed aqueous chemistry data. Based on a good match between modeling output and the field data, we estimate mineral transformations and biomass accumulation during the field experiments. Although direct quantitative verification of these estimates can be difficult, water chemistry data provides an indirect indicator of the system dynamics and the abundance of iron-reducing bacteria (dominated by members of Geobacteraceae) (4). The concentration of U(VI) decreases during the period of active iron reduction, but rebounds somewhat following the onset of sulfate reduction, suggesting more efficient removal of U(VI) by the iron reducing community (1).

This work focuses on a field-scale uranium bioremediation experiment conducted at the Department of Energy’s Integrated Field Research Challenge (IFRC) site near Rifle, Colorado. Specific details regarding the site and the biostimulation experiment have been documented elsewhere (1, 8). Acetate was injected for two successive years (2002 and 2003) to investigate the effects of biostimulation for the purposes of remediation of uranium. Hydrological analysis has suggested changes in flow and conductivity fields during uranium bioremediation (9). Due to the negligible concentration of dissolved oxygen and nitrate, Fe-hydroxide reduction occurs after acetate injection, followed by sulfate reduction. Previous studies have shown a rapid increase in the abundance of iron-reducing bacteria (dominated by members of Geobacteraceae) during the early stage of field experiments, followed by rapid growth of sulfate reducers (1). The concentration of U(VI) decreases during the period of active iron reduction, but rebounds somewhat following the onset of sulfate reduction, suggesting more efficient removal of U(VI) by the iron reducing community (1).

Although aqueous geochemistry data were routinely collected during field-scale remediation experiments, direct observation of the evolution of mineral composition and biomass remains challenging, partly due to the associated expense and technical difficulty. Here we combine reactive transport modeling with column experiments to establish the biogeochemical reaction network, which was then used to match field aqueous geochemistry data. Based on a good match between modeling output and the field data, we estimate mineral transformations and biomass accumulation during the field experiments. Although direct quantitative verification of these estimates can be difficult, water chemistry data provides an indirect indicator of the system dynamics and the consistency of the estimation with the observed biogeochemistry. Used in this way, reactive transport modeling provides an efficient and cost-effective tool not only to understand the dominant biogeochemical processes, but also to estimate changes in the aquifer properties. Although field bioremediation has been modeled before (e.g., 10), to our knowledge, this is the first study that integrates column and field experimental data to quantify the evolution of the mineral phases and biomass abundance associated with bioremediation at the field scale.

Experimental Section

Column Experiment. Column experiments were carried out using groundwater and sediments obtained from the Rifle IFRC site. The groundwater was pumped from a location upgradient of the experimental area and was unimpacted by previous experiments at the site. The sediments were obtained through trenching and were sourced from a location beneath the water table at a depth of approximately 4 m below ground surface. The sediments and groundwater were kept in no-headspace vessels and stored at 4 °C in the dark until use. The sediments were sieved (<4 mm) prior to being wet packed into 4.8 cm diameter × 30.5 cm long glass
chromatography columns. The porosity of the sediments was determined to be 0.35 ± 0.02, yielding a pore volume of approximately 216 cm$^3$. Prior to the addition of acetate, ten pore volumes of N$_2$-sparged groundwater were passed through the column. Influent groundwater was then amended with 3 mM sodium acetate, degassed with ultra high purity N$_2$, and sealed under the same gas headspace for the duration of the experiment (103 days). Groundwater was passed through the columns using a peristaltic pump with a flow rate of 125 µL/min (0.83 pore volumes per day), yielding a pore water velocity of 25.4 cm/day. Effluent samples were analyzed by ICP-MS and ICP-OES.

Field Site and Experiments. The Rifle IFRC site is part of the Uranium Mill Tailings Remedial Action (UMTRA) program operated by the U.S. Department of Energy. It is located on a flood plain of the Colorado River, with groundwater flow mainly within a gravelly sandy aquifer with an average hydraulic conductivity of approximately 35 m/d and mean porosity of 0.27. The groundwater contains dissolved uranium at the micromolar concentration level along with species such as Ca$^{2+}$, Mg$^{2+}$, SO$_4^{2-}$, and HCO$_3^-$, as listed in Table S1 in the Supporting Information. The minerals in the sediments include quartz, calcite, a variety of clays, and various iron-containing minerals, including iron (hydro)oxide (goethite, hematite, magnetite, and ferrihydrite). The amount of Fe(III) in the sediment samples significantly varies over space depending on the lithology of local sediments, with Fe(III) content in the fine-grained sediment (<4 mm fraction) used for the column experiment being approximately 24.5 µmol/g, measured by hydroxylamine reduction for 16 h. A dithionite extraction (CBD) showed additionally 23.6 µmol/g of Fe in more crystalline Fe-oxide phases. The sediments include both coarse-grained material (gravels and cobbles), as well as beds of fine-grained silt and clay. The percentage of fine-grained material in the sediments, the adjusted extractable iron content (including both the 0.5 N hydroxylamine and CBD extractable) would be about 26 µmol/g, and the adjusted hydroxylamine extractable iron is about 13 µmol/g. The amount of “bioavailable” Fe(III) was obtained by fitting both the column and field experimental data.

During the field experiments, acetate and bromide were mixed with background groundwater and injected into the subsurface for 119 days in 2002 and for 111 days in 2003. There were three rows of monitoring wells located at approximately 3.7, 7.3, and 14.6 m downgradient from the injection gallery. Influent groundwater was then amended with 3 mM sodium acetate, degassed with ultra high purity N$_2$, and sealed under the same gas headspace for the duration of the experiment (103 days). Groundwater was passed through the columns using a peristaltic pump with a flow rate of 125 µL/min (0.83 pore volumes per day), yielding a pore water velocity of 25.4 cm/day. Effluent samples were analyzed by ICP-MS and ICP-OES.

Reactive Transport Modeling. This section provides a brief introduction on the key elements involved in modeling. More details are provided in Supporting Information. The code CrunchFlow (11), a software for modeling multicomponent reactive flow and transport (12), was used to simulate the reactive transport processes for both column and field experiments. The reaction network includes both kinetic and instantaneous reactions. The kinetic reactions include microbially mediated reduction of uranium, bioavailable iron hydroxide, and sulfate, as well as mineral dissolution and precipitation. These reactions, together with instantaneous reactions such as sorption/desorption and ion exchange, are listed in Table 1. Aqueous speciation reactions are listed in Table S2 in the Supporting Information.

The microbially mediated reactions were formulated so as to partition electron donors between respiration and cell growth according to the microbial energetics (13). Iron-reducing and sulfate-reducing bacteria (FeRB and SRB) are represented by the formulas C$_6$H$_{12}$O$_7$N$_{probi}$ and C$_3$H$_8$O$_3$N$_{srb}$, respectively, and are considered as reaction products. The rates of the reactions were modeled using a dual Monod rate expression with an inhibition term, which simulates the successive redox sequence of electron acceptors. The rate law explicitly incorporates the dependence of the rate on biomass, as shown in the following equation.

$$r_i = -k_{max}X_iK_M + C_S K_{MEA} + C_{TEA} K_T + C_I$$

Here the reaction rate $r_i$ of substrate $S$ depends on the maximum utilization rate $k_{max}$ (mol/L/s-cells), the concentration of the bacteria that degrade the substrate $X_i$ (cells), the concentrations of the substrate $C_S$ and that of the terminal electron donor $C_{TEA}$ through the dual Monod terms with their respective half-saturation constants $K_M$ and $K_{TEA}$ (mol/L), as well as the inhibition term, with $K_T$ being the inhibition constant (mol/L) and $C_I$ being the concentration of the inhibiting substance $I$ (mol/L). Here sulfate reduction is assumed to be inhibited by iron hydroxide present locally while iron and uranium reduction are not inhibited due to the negligible concentrations of dissolved oxygen and nitrate.

The biomass is conceptually modeled as an immobile phase based on literature observations that bacterial cells usually associate with the solid phase as a biofilm or as discrete biomass colonies (13). The bacteria grow as the products of the reduction reactions at the rate specified by eq 1, which is in turn affected by the amount of bacteria produced. As such, the model explicitly simulates the microbial community dynamics and includes the effects of the dynamic evolution on microbially mediated reaction rates (14).

The rates of mineral dissolution and precipitation, including the abiotic reduction of Fe(OH)$_2$ and H$_2$S (23), follow a rate law derived from Transition State Theory (TST) (26). All reaction rate constants, including the $k_{max}$ values for the microbially mediated reactions and the log $k$ values for the mineral dissolution and precipitation, and the amount of biomass available Fe(III), are obtained through calibration for the field data and are listed in Table 1. All other kinetic and thermodynamic parameters were taken from literature. Data from the column experiments were used to determine the kinetic reactions involved during uranium bioremediation. The same reaction network was then applied to simulate the field experiments using 1D reactive transport modeling. Based on the constraints provided by the field aqueous data, the temporal evolution of the solid phase composition and the amount of biomass were determined.

The domain for the field experiment is considered homogeneous, with a constant porosity of 0.27 and permeability of 7.28 × 10$^{-12}$ m$^2$. The simulations were carried out within a 24 m domain, with 480 nodes and a resolution of 5.0 cm. The dispersion coefficient was chosen to be 0.04 m$^2$/s to best fit the tracer field data, also a typical value at a spatial scale of tens of centimeters (27). The initial and boundary conditions for the aqueous species were the measured background composition at Rifle, as listed in Table S1. Initially both FeRB and SRB were considered to exist ubiquitously, occupying 10$^{-4}$ volume percentage of the solid phase. This volume percentage corresponds to a cell density of 10$^9$ and 10$^7$ cells/mL of the sediment, which is typical for natural sediments (19). The simulations were carried out with the initial conditions of the 2003 experiment set to be the final state of the 2002 experiment.

This work extends previous modeling efforts (8) for the Rifle site in several ways: (1) a complete reaction network was developed from the column experiments; (2) biomass growth was modeled to keep track of the evolution of the
### TABLE 1. Important Reactions Resulting from Biostimulation and Their Kinetic and Thermodynamic Parameters

<table>
<thead>
<tr>
<th>no.</th>
<th>reaction stoichiometry</th>
<th>kinetic parameters (as shown in equation 1)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>microbially mediated redox reactions</td>
<td>$k_{max}$ (mol/L/s/cell) $K_M$ (mol/L) $K_{MTEA}$ (mol/L) $K_I$ (µmol/g)</td>
</tr>
<tr>
<td>1a</td>
<td>FeOOH(s) + 1.925H$^+$ + 0.033NH$_4^+$ + 0.208CH$_3$COO$^-$ → Fe$^{2+}$ + 0.033C$_2$H$_5$O$<em>2$N$</em>{FeRB}$ + 0.25HCO$_3^-$ + 1.6H$_2$O</td>
<td>9.48 × 10$^{-17}$ 1.0 × 10$^{-3}$ 5.0 (column)/2.5 (field) 5.0 (µmol/g)</td>
</tr>
<tr>
<td></td>
<td>amount of “bioavailable” iron</td>
<td></td>
</tr>
<tr>
<td>2a</td>
<td>SO$_4^{2-}$ + 1.082CH$_3$COO$^-$ + 0.052H$^+$ + 0.035NH$_4^+$ → 0.035C$_5$H$_7$O$_2$N(SRB) + 0.104H$_2$O + 2HCO$_3^-$ + HS$^-$</td>
<td>5.55 × 10$^{-16}$ 1.0 × 10$^{-3}$ 1.0 × 10$^{-3}$ 1.8</td>
</tr>
<tr>
<td>3a</td>
<td>UO$_2^{2+}$ + 0.067NH$_4^+$ + 0.417CH$_3$COO$^-$ + 0.8H$_2$O → UO$_2$(s) + 0.0667C$_5$H$_7$O$_2$N(FerRB) + 0.5HCO$_3^-$ + 2.15H$^+$</td>
<td>1.11 × 10$^{-17}$ 1.0 × 10$^{-4}$ 1.0 × 10$^{-6}$ 5.0 (µmol/g)</td>
</tr>
<tr>
<td></td>
<td>mineral dissolution and precipitation</td>
<td>log $k$ (mol/m$^3$/s) $K_{eq}$</td>
</tr>
<tr>
<td>4</td>
<td>CaCO$_3$(s) + H$^+$ ↔ Ca$^{2+}$ + HCO$_3^-$</td>
<td>−5.2 1.85</td>
</tr>
<tr>
<td>5</td>
<td>MgCO$_3$(s) + H$^+$ ↔ Mg$^{2+}$ + HCO$_3^-$</td>
<td>−6.3 (column)/−7.3 (field) 2.29</td>
</tr>
<tr>
<td>6</td>
<td>Fe$^{2+}$ + H$_2$S(aq) ↔ FeS(amt) + 2H$^+$</td>
<td>−9.0 3.50</td>
</tr>
<tr>
<td>7</td>
<td>Fe(OH)$_3$(s) + 0.5H$_2$S(aq) + 2.5H$^+$ ↔ Fe$^{2+}$ + 0.5S(s) + H$_2$O + 2HCO$_3^-$</td>
<td>−5.0 (column)/−3.0 (field) −19.60</td>
</tr>
<tr>
<td>8</td>
<td>FeCO$_3$(s) + H$^+$ ↔ Fe$^{2+}$ + HCO$_3^-$</td>
<td>−9.0 −0.19</td>
</tr>
<tr>
<td></td>
<td>surface complexation</td>
<td>log $K'$</td>
</tr>
<tr>
<td>9</td>
<td>&gt;FeOOH + Fe$^{2+}$ ↔ &gt;FeOOFe$^+$ + H$^+$</td>
<td>3.70 × 10$^{-6}$ 79.8 −0.54</td>
</tr>
<tr>
<td>10</td>
<td>&gt;FeOOH + Fe$^{2+}$ + H$_2$O ↔ &gt;FeOOFeOH + 2H$^+$</td>
<td>−10.89</td>
</tr>
<tr>
<td>11</td>
<td>&gt;FeOOH ↔ &gt;FeOO$^-$ + H$^+$</td>
<td>−9.51</td>
</tr>
<tr>
<td>12</td>
<td>&gt;FeOOH + H$^+$ ↔ &gt;FeOOH$^2_-$</td>
<td>7.47</td>
</tr>
<tr>
<td></td>
<td>ion exchange (Vanselow convention)</td>
<td>cation exchange capacity (eq/g) log $K'$</td>
</tr>
<tr>
<td>13</td>
<td>NaX ↔ Na$^+$ + X$^-$</td>
<td>8.0 × 10$^{-5}$ 0.00</td>
</tr>
<tr>
<td>14</td>
<td>KX ↔ K$^+$ + X$^-$</td>
<td>−0.60</td>
</tr>
<tr>
<td>15</td>
<td>CaX$_2$ ↔ Ca$^{2+}$ + 2X$^-$</td>
<td>−0.50</td>
</tr>
<tr>
<td>16</td>
<td>MgX$_2$ ↔ Mg$^{2+}$ + 2X$^-$</td>
<td>−0.40</td>
</tr>
</tbody>
</table>

* See details in Supporting Information for the derivation of the reactions. ^ Ref 15. Range between 10.5 and 235.0 µmol/g (18, 17). $^d$ Range between 0.0043 and 0.1 mM (18). $^f$ Range between 0.1 and 7.5 mM (19, 20). $^g$ Range between 0.8 and 5 µmol/g (8, 21). $^h$ No literature data available. $^i$ Range between 0.0001 and 24 mM (21, 22). $^j$ All surface complexation parameters are from ref 23. $^k$ All log $K$ values from ref 24.
iron- and sulfate-reducing populations; (3) rate laws for microbially mediated reactions explicitly incorporated the impact of biomass growth; (4) the focus of the modeling was on a quantitative estimation of mineral transformation and biomass accumulation; and (5) all field aqueous data measured were used as constraints (not just bromide, acetate, U(VI), and sulfate).

Results and Discussion

Reaction Network Based on Column Experiments. As indicated by the increase in the effluent Fe(II) in Figure 1, the injection of acetate led to the reductive dissolution of iron hydroxide. After about 25 days, presumably accompanying the near depletion of bioavailable iron hydroxide as indicated by modeling results, the increasing activity of sulfate-reducing bacteria began, which led to the decrease in concentrations of sulfate and Fe(II). The sulfide species reacted with ferrous ion to form iron sulfide, indicated by the accumulation of visibly dark regions along the column (Figure S1). The effluent pH increased significantly over time in the column, reaching as high as 8.5, in contrast to the fairly constant pH of 7.0 in the field experiments. The increase in the column pH appears to be due to the use of a 100% N₂ headspace, which resulted in degassing of initially 1% CO₂ from the groundwater in the column. In the model we simulate this as a process of CO₂ depletion, driving the pH from 7.0 initially to 8.5 at later times. Alkalinity increased as a result of the acetate oxidation, causing calcite precipitation as confirmed by the decrease in Ca²⁺ concentration at later time points. In addition, modeling results suggested that magnesite and siderite could have precipitated as the alkalinity increased, although this has not been confirmed by destructive characterization. Elemental sulfur precipitated due to the reductive dissolution of iron oxide by H₂S is also predicted by the modeling.

Modeling of Field Experiments. The reaction network inferred from the column experiment was used in the simulation of the field experiments. The modeling results were compared with the measured aqueous concentrations averaged over the 5 monitoring wells within each row for the field experiments in 2002 and 2003, as shown in Figure 2 and Figure S2, respectively. Compared to the column experiments, the field data are much more variable as indicated by the large error bars corresponding to the double standard deviation range of the averaged concentrations; this variability is likely due to the spatial variation in aqueous geochemistry and aquifer properties. Although the same reaction network was applied, some of the parameters had to be adjusted for the field data to obtain a good fit. For example, the initial bioavailable Fe(III) content was changed from 5.0 µmol/g in the column experiments to 2.5 µmol/g in the field. This is likely due to the fact that the column experiment used the <4 mm fraction of the sediment, which contains a higher Fe(III) content than does the coarse fraction present as well.
in the field experiment. The amount of bioavailable Fe(III) predicted from modeling comprises about 10% of the hydroxylamine extractable Fe(III) for fine-grained sediments used in the column (about 24 µmol/g) and 20% of the hydroxylamine extractable Fe(III) for field sediments (about 13.0 µmol/g). This may seem a small percentage; however, these numbers fall into the reported percentages of low or medium surface area crystalline Fe(III) that can be used by bacteria (8–18%) (17).

The field data suggested a sequence of redox reactions similar to what is observed in the column experiments, as indicated by an early increase in Fe(II) concentration, a later decrease in SO$_4^{2-}$ concentration and increase in S(-II) concentration. In general, as shown in Figure 2, the model has captured the biogeochemical dynamics in the field. Except for the pH values at early times, the modeling results match the average field data very well. The good fit of the solute breakthrough data provides a solid foundation for the estimation of solid phase composition and biomass accumulation. Significantly, such good fit was possible only after explicitly modeling the microbial community dynamics and its impact on the rates of microbially mediated reactions.

**Estimation of Mineral Transformation and Biomass Accumulation.** Based on the good fit to field aqueous geochemistry data, the change in the solid phase composition and the amount of biomass accumulation were estimated. Figure 3 shows the temporal evolution of important minerals. After acetate injection, the bioavailable iron hydroxide was consumed initially within regions close to the injection wells and this zone of Fe-hydroxide depletion expanded over time in the downgradient direction. Most microbially mediated Fe reduction occurred within the first 50 days in 2002. In addition to enzymatic reduction by bacteria, iron oxide was also reduced abiotically by H$_2$S generated from sulfate reduction. As shown in Figure 3B, this pathway actually consumed much more Fe(OH)$_3$ (with a maximum value of 0.13% of the pore space, equivalent to 6.9 µmol/g sediments at the end of 2003) than the microbially mediated Fe reduction did. Calcite and FeS(am) are predicted to be the primary precipitates, with maximum accumulation close to the injection wells. Although not shown in Figure 3, elemental sulfur also precipitated according to the modeling. Its time evolution mirrors that of the H$_2$S-reacted iron hydroxide (Figure 3B), being stoichiometrically equal to half of the Fe(OH)$_3$, reacted in this way on a molar basis. In addition, modeling results suggest that siderite, magnesite, and UO$_2$(s) precipitated, although in much smaller quantities.

For the biomass, the time evolution of the iron and sulfate reducers differs significantly, as shown in Figure 4. For the first 50 days in 2002, the FeRB grew rapidly, while the sulfate reducers remained negligible due to the inhibition of the sulfate reduction pathway until exhaustion of the bioavailable iron hydroxide. In addition, the spatial distribution of the two microbial populations differs. Initially the FeRB grew in the vicinity of the injection wells, and then expanded downgradient to areas with more bioavailable iron hydroxide. In contrast, SRB accumulated mostly at the location of injection wells, while at other locations the accumulation is orders of magnitude smaller. This is due to two reasons: (1) the concentration of sulfate is so high that the amount of acetate injected is never large enough to consume all of it, with the result that acetate is depleted close to the injection well, and (2) sulfate is present in the groundwater and therefore is continuously replenished from upgradient, thus fixing the sulfate reduction zone in space. In addition, the accumulation of SRB was much higher in 2003 than in 2002, largely because the amount of acetate injected was approximately 3 times higher in 2003 than it was in 2002. The localization of sulfate reduction in the vicinity of the injection wells also explains why most of mineral dissolution and precipitation occurs there as well, since these are driven by their reaction products (alkalinity, sulfide, Fe(II)). In total, the amount of precipitates and

**FIGURE 3.** Time evolution of volume percentage of the pore space occupied by various minerals. The average groundwater direction is toward the right. The injection wells are at a distance of 3.7 m.
also been observations that different redox reactions can and has been observed in many field systems, there have been a consequence of assuming that redox reactions only a very small percentage of the total iron oxide present. The modeling results also suggest that the "bioavailable iron" is represent the heterogeneous nature of the field system. The captured. Systems such as flow cell experiments may better capture the amount of "bioavailable" iron. This indicates that differences between the column and field systems, including the location of the injection wells has been observed by previous studies (1). As such, the modeling can at least direct data collection activities in future bioremediation experiments. The good match obtained between modeling output and field data suggests that the column experiments have captured the major biogeochemical dynamics in the field system. However, some of the parameters had to be changed due to the differences between the column and field systems, including the amount of "bioavailable" iron. This indicates that although the column experiments can be an analogue of the field system, not all features of the field systems can be captured. Systems such as flow cell experiments may better represent the heterogeneous nature of the field system. The modeling results also suggest that the "bioavailable iron" is only a very small percentage of the total iron oxide present. This may be a consequence of assuming that redox reactions occur in sequence. Although this is true thermodynamically and has been observed in many field systems, there have also been observations that different redox reactions can occur in parallel under certain conditions (28).

Importance of Explicitly Modeling Microbial Community Structure. The maximum rate of Fe(III) reduction occurred at the point where the reaction front was formed. The highest reaction rate occurred close to the injection wells initially and propagated downgradient over time. For sulfate reduction, no such downgradient propagation was predicted; however, the several orders of magnitude increase in SRB close to injection wells caused a commensurate increase in the reduction rates following eq 1. Although not shown here, a reaction rate without a dependence on the amount of biomass leads to an underestimation of rates close to the injection wells and an overestimation of rates downgradient, indicating the importance of explicitly modeling the spatial and temporal evolution of microbial communities.

Possibility of Pore Clogging. The total amount of accumulated biomass and precipitates may seem small (up to 1.5% of the pore space), but the effects on the physical and chemical properties of the aquifer can be significant. How the precipitates and biomass clog the pore space depends on the details of where the reactions occur and the pore size distribution of the sediments. Multiple studies have shown that it is easier to clog a porous medium with fine-grained materials (29–31). The current estimate is based on the assumption that the system is a homogeneous system with uniform distributions of aquifer properties. As such, the estimate indicates the "average" behavior of the system. Because the Rifle aquifer is actually heterogeneous and contains both coarse-grained and fine-grained sediments (32), it is possible that such heterogeneity can have a significant local impact, resulting in preferential clogging of pores within the fine-grained sediments. The effect of heterogeneities on mineral transformation and biomass accumulation is the topic of another study (33).

Although the biomass accumulated in amounts similar to the mineral precipitates, it is likely to have a larger impact on pore clogging. Biomass not only occupies the pore space by volume, but also produces materials like extracellular polymeric substance (EPS), which is sheet-like and can decrease the flow-conducting volume (34). Using a relationship between conductivity reduction and the amount of biomass accumulated (6), the volume reduction at the location of the injection wells reached as high as a factor of 0.037 (or more than 1 order of magnitude) by the end of 2003.

Environmental Implications of Bioremediation. To the best of our knowledge, this is the first field-scale study on quantification of mineral transformation and biomass accumulation associated with biostimulation. Although a few field bioremediation studies have indicated or demonstrated alteration in flow fields (7, 9), this study suggests that in situ bioremediation activities involving organic carbon amendment can generate clogging processes in the vicinity of injection wells where the highest concentrations of organic carbon exist. The results suggest that to maximize the bioremediation efficacy, it is important to carefully design the injection wells and their spatial location to minimize the impact of clogging processes. In addition, specific injection conditions, such as the concentration of organic carbon and its rate of injection, will also be important in minimizing pore clogging. In this context, reactive transport modeling, coupled with the requisite field data, can serve as an efficient and cost-effective tool to integrate the effects of complex biogeochemical and transport processes that occur in situ, to quantitatively estimate the temporal and spatial evolution of the precipitates and biomass in different injection scenarios, and to help improve bioremediation strategies.

Acknowledgments

Funding for this study was provided by the U.S. Department of Energy, Biological and Environmental Research Program Contract DE-AC0205CH11231 to the LBNL Sustainable Systems Scientific Focus Area. We also acknowledge Phil Long (PNNL) and the Rifle IFRC research team for facilitating
collaboration and access to Rifle data sets. We thank the reviewers for constructive comments that have improved the paper.

Supporting Information Available

Details of reactive transport modeling and simulation conditions. This information is available free of charge via the Internet at http://pubs.acs.org.

Literature Cited