

Flow cytometry as tool to monitor chain elongation performance

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HIGHLIGHTS:

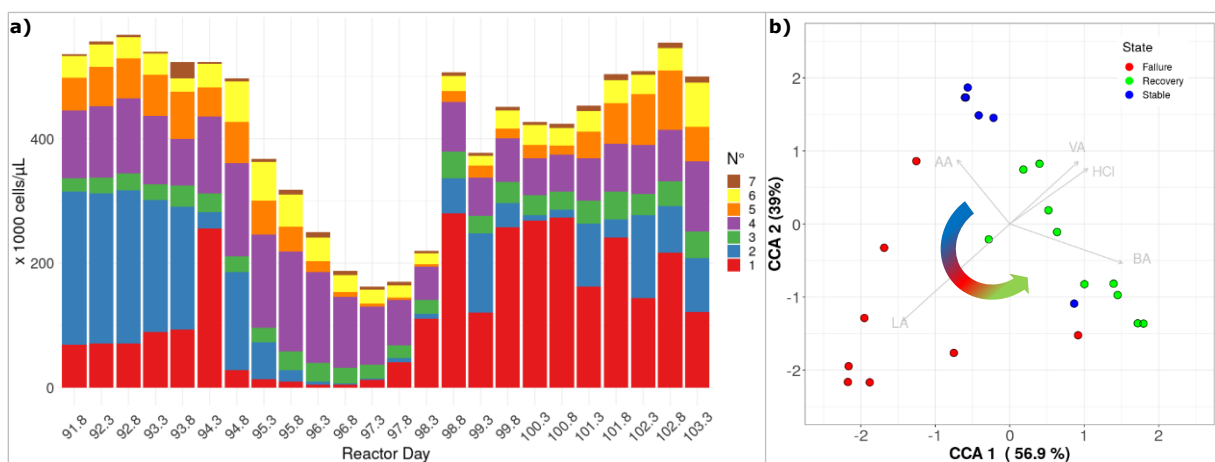
- Flow cytometry is a rapid and reliable method to monitor the microbial community in a chain elongating reactor
- Phenotypic fingerprints of different reactor states can be differentiated via flow cytometry.
- A reactor microbiome for lactic acid chain elongation, disturbed due to caproic acid toxicity, has the potential to return to its stable state during recovery.

BACKGROUND: During the past years there has been a growing awareness that we need to strive to a circular bio economy, which has stimulated the development of novel biological processes for the production of added-value platform chemicals from organic waste streams. These new developments go hand in hand with an increasing need for process stability and performance insurance. Therefore, a good monitoring strategy is essential. The state-of-the-art mainly relies on monitoring physicochemical input and output parameters. Here we report the development of a novel monitoring strategy in which microbial community (MC)-parameters play a key role. The MC-parameters are obtained via flow cytometry (FCM), which is a MC analysis technique that allows a fast assessment of its phenotypical diversity [1,2]. FCM as a process monitoring and control tool has mostly been applied to the MC in environments containing little nutrients and lower microbial abundance, such as drinking water [3]. More recently, the use of FCM is emerging for the analysis of the MC and the detection of disturbances through flow cytometric fingerprinting [3,4].

In this study, FCM is studied as a key tool for the fast detection of community changes and applied to caproic acid (CA) production via lactic acid (LA) chain elongation, with the potential of applying this on other fermentation processes in the future. We operated mixed-culture CSTR-bioreactors that were fed with a synthetic medium that contains LA (20.8 g/L) as main carbon source. During reactor operation, several physicochemical parameters (pH, electroconductivity, biogas yield and composition, carboxylate yields and spectrum, etc.) were monitored along with the microbiology. From the FCM data, the cell counts and diversity parameters were derived. The community structure was determined as a phenotypic fingerprint (PFP) based on the identification of phenotypes with a model-based approach based on Gaussian Mixture Models (GMM) [5].

45 **RESULTS & DISCUSSION:** During reactor operation, periods of decreased
 46 performance or process failure were observed. The results of one period of
 47 process failure are presented. Physiochemically, process failure was initially
 48 observed as a sudden, complete stop of the gas production, while prior to
 49 this event, around $1.2 \text{ L}_{\text{biogas}} \cdot \text{L}_{\text{reactor}}^{-1} \cdot \text{d}^{-1}$ containing 23.25% H_2 and 73.75%
 50 CO_2 was produced. H_2 and CO_2 are by-products of LA chain elongation [6-
 51 9]. Analysis of the carboxylate spectrum showed the highest CA
 52 concentration (8.0 g/L) observed in the reactors at the moment of process
 53 failure. LA accumulated up to 8.4 g/L and the CA concentration decreased.
 54 Activity resumed once the CA-concentration had decreased to 3.2 g/L. It is
 55 therefore hypothesised that CA-toxicity was a possible cause of process
 56 failure in the process under study.

57 A GMM-model was generated to determine the PFP of the reactor samples.
 58 Clear shifts in the community structure based on the PFP could be observed
 59 during this period (figure 1, a). Ordination of the samples based on the
 60 Bray-Curtis dissimilarities via Canonical Correspondence Analysis (CCA)
 61 shows the evolution of the reactor samples during normal operation,
 62 process failure and recovery (figure 1, b). Samples taken prior to process
 63 failure cluster together in the CCA plot, whereas the samples taken during
 64 process failure and recovery are ordinated in a counterclockwise pattern.
 65 The further in the recovery phase, the closer the samples are ordinated
 66 to the plotted to the cluster of samples from before process failure. This
 67 indicates that the microbiome performing LA chain elongation can return to
 68 its stable state prior to the crash.



69 Figure 1. a) absolute phenotypic distribution over time, based on a GMM-
 70 model with seven identified phenotypes. b) CCA analysis of the samples
 71 based on the Bray-Curtis dissimilarities, constrained by the LA, acetic acid
 72 (AA), butyric acid (BA) and valeric acid (VA) concentrations, and the HCl
 73 requirements for pH control.
 74

75 **CONCLUSION:** FCM is a suitable technique for fast determination of the
 76 PFP of a mixed-culture microbiome. Changes in reactor performance can be
 77 monitored as well on microbial level. FCM is therefore a promising tool to
 78 be included into the monitoring strategy of LA chain elongation, with the
 79 potential of extending its application on other bioproduction processes.

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