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## Development and validation of a technoeconomic analysis tool for early-stage evaluation of bio-based chemical production processes

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### HIGHLIGHTS

- Multi-stage bioprocesses that convert sugar to chemicals are of commercial interest.
- We model these processes by making simplifying assumptions.
- Our model estimates the cost of producing a chemical through a specific route.
- Early-stage cost estimates are crucial for evaluating competing technologies.
- The model also can illustrate process bottlenecks to help guide research.

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### ABSTRACT

By using cost correlations and standard scale-factors, a spreadsheet-based early-stage cost estimation tool was developed. Named BioPET (Biorenewables Process Evaluation Tool), this tool allows users to specify up to seven primary unit operations – fermentation, separation, three catalytic stages, and purification – along with key parameters for each. BioPET then computes an estimated minimum selling price for the pathway. Model validation was conducted by selecting three molecules (ethanol, succinic acid, and adipic acid), and comparing BioPET's results to literature values and to results from a commercial process design tool. BioPET produced virtually identical prices to the process design tool, although the costs were not identically distributed amongst the categories. BioPET produced estimates that were within 40% of other literature values at low feedstock costs, and within 5% at high feedstock costs.

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### 1. Introduction

Bio-based chemicals represent an opportunity to produce value-added products from sugars. These chemicals are an attractive alternative to biofuels because of their higher market prices compared to biofuels. In a 2004 study, Werpy and Petersen identified ten chemicals that presented the greatest short-term opportunity for bio-based chemical production in the U.S., spurring tremendous efforts to increase the economic viability of these bio-based chemicals (Sánchez et al., 2005; Song and Lee, 2006; Werpy and Petersen, 2004). However, little is known about the economics of producing these value-added bio-based chemicals at commercial scale.

One chemical that does possess a good deal of process information is ethanol due to its large-scale deployment as a 1st-generation biofuel. The broad ethanol literature encompasses process improvements, technoeconomic analyses (TEAs), and life-cycle assessments, and can provide fundamental knowledge to inform

studies about other bio-based chemicals (Kwiatkowski et al., 2006). Robust TEA's, in particular, have the ability to illuminate process bottlenecks and to clarify how process alternatives will impact production costs (Oleskiewicz-Popiel et al., 2012). Typically, these TEA's require extensive knowledge of process parameters and design details only available during the latter stages of a project. However, early-stage cost estimation is critical to helping companies and applied academic research centers chart a course through translational research and towards economic viability.

As novel metabolic pathways are explored or novel hybrid fermentative-catalytic processes are proposed (Nikolau et al., 2008); comprehensive and accurate process data necessary for detailed TEAs of these operations at full-scale will be years away. And yet, strong evidence regarding the economic viability of a particular chemical is needed early in the process to warrant continued investment of resources. By making simplifying assumptions and estimates for key variables, it is possible to develop an early-stage TEA of novel processes. Strong TEA capabilities exist commercially in tools such as Aspen Process Economic Analyzer and Intelligen SuperPro Designer®, both of which provide estimations of capital and operating costs. But these tools also require a level of detail

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**Nomenclature**

TEA	technoeconomic analysis	$V_m$	Maximum attainable size of a centrifuge
PFD	process flow diagram	AR	adsorption ratio (kg product adsorbed/Mg adsorbent)
$C_n$	new cost for newly sized piece of equipment	$K_f$	Freundlich coefficient
$S_n$	new size of equipment	[A]	concentration of product in solution (kg/m <sup>3</sup> )
$S_o$	size of equipment where previous cost data exists	n	Freundlich exponent
$C_o$	cost of equipment where previous data exists	NA	adsorbent needed (Mg)
n	empirically-derived cost exponent	$\tau_A$	adsorption time constant (hrs <sup>-1</sup> )
$V_{fm}$	annual volume of fermentation media (m <sup>3</sup> )	$Y_A$	adsorption yield
$m_T$	annual production of fermentation product (kg)	$m_p$	mass of product per batch (kg)
$C_{fm}$	final titer of fermentation product (kg/m <sup>3</sup> )	$S_c$	cost of crystallizer
$FV_{fm}$	useable fraction of fermenter volume (% purchased volume/usable volume)	X	flow rate of crystals (klb hr <sup>-1</sup> )
$\eta_T$	mass conversion efficiency (dimensionless, kg final prod./kg fermentation prod.)	$N_{min}$	minimum number of stages
MR	mass ratio (dimensionless, molecular weight of final product/molecular weight of molecule produced in fermentation)	$X_b$	fraction of product in the bottoms
$N_b$	number of annual batches produced	$X_d$	fraction of product in distillate
$N_d$	number of days of plant operation (days)	$Y_b$	fraction of solute in bottoms
$t_{fm}$	total time to complete a fermentation batch (days)	$Y_d$	fraction of solute in distillate
$N_{fm}$	number of fermenters required	$\alpha$	relative volatility of product and distillate
$V_{max}$	maximum attainable volume in a purchased fermenter (m <sup>3</sup> )	$N_{actual}$	actual number of sieve trays required
$N_c$	number of centrifuges	$\epsilon_{tray}$	Murphee tray efficiency
$t_d$	fermentation downtime (hrs, time used for cleaning and refilling)	$E_e$	extraction factor
$VE_c$	Centrifuge volumetric energy requirement (kWh/m <sup>3</sup> fermentation liquid)	$K_{Di}$	distribution coefficient of the product in its respective solvent
		E	extract flow rate
		R	raffinate flow rate
		$\phi_c$	fraction of product not extracted

that is typically unavailable at early stages in process evaluation. While preliminary cost evaluation methods have been outlined by several authors (Peters et al., 2003; Turton et al., 2010), the authors are unaware of any widely-available early-stage TEA model or tool for bio-based chemicals. To address this gap, a spreadsheet-based tool was developed to provide early-stage TEAs of bio-based chemicals named BioPET (Biorenewables Process Evaluation Tool). Key criteria used in the development of BioPET were as follow: (1) ease of use, (2) minimal data inputs, (3) results comparable to simplified models implemented in existing cost-modeling software, and (4) simple graphical reporting of estimated minimum selling prices and cost breakdowns. To operate the tool, users need a basic knowledge (or educated guesses) for each unit operation comprising their overall process design of interest. Once developed, BioPET was compared against SuperPro Designer<sup>®</sup> and results from Patel (2006) for a suite of three chemicals: ethanol, succinic acid, and adipic acid. The objectives of this research were as follow: (1) To develop a tool capable of informing economic decisions regarding new pathways developed for bio-based chemicals, and (2) To evaluate the tool by comparing its output to literature values and to more sophisticated process modeling/design tools.

## 2. Methods

### 2.1. Methods for BioPET development

BioPET was designed with the objectives of evaluating multiple processes with (re)construction of new process flow diagrams (PFD) with each new evaluation. With organisms capable of consuming many types of feedstocks, the model remained agnostic to where the feedstock was derived. In doing so, the tool does

not consider upstream processes such as starch hydrolysis or pre-treatment and hydrolysis of lignocellulosic biomass – these were considered outside the scope of the model and the feedstock price was considered as a lumped parameter to include the costs of the initial source and conversion technology if required. The feedstock can then be directly fed to fermentation or catalysis, and any other subsequent unit operations.

BioPET assumes the following carbon flow: fermentation, followed by a separation stage, followed by up to three catalytic processes, finishing with up to two purification stages with different unit operations allowed within each stage. All stages in the tool can be toggled on and off to allow for process flexibility. The following approach was taken to accommodate inherent complexity of the separation, catalysis, and purification processes while allowing for a relatively simple user interface: The types separation and catalytic methods along with assumptions are listed in the following sections. This shields the user from having to provide full process details that are often not available at early stages of a project. Finally, within the hypothetical plant, BioPET only examines a stream of material consisting of a primary product and solvent. This binary system uses mass balance equations and relationships to characterize all steps post-fermentation. Using a series of inputs, respective assumptions, and equations described in the following sections, process cost estimations can be made.

Unless otherwise specified, a standard scale-factor approach, as embodied in Eq. (1) (Peters et al., 2003), was used to adjust capital costs based on unit operation size.

$$C_n = \frac{S_n}{S_o} \times C_o^n \quad (1)$$

Where,  $C_n$  is the new cost for newly sized piece of equipment,  $S_n$  is the new size of equipment,  $S_o$  is the size of equipment where previous cost data exists,  $C_o$  is the cost of equipment where previous data

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