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Tonini, Davide; Hamelin, Lorie; Astrup, Thomas Fruergaard

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Environmental implications of the use of agro-industrial residues for biorefineries: application of a deterministic model for indirect land-use changes

DAVIDE TONINI¹, LORIE HAMELIN² and THOMAS F. ASTRUP¹
¹Department of Environmental Engineering, Technical University of Denmark, Miljoevej 115 2800 Kgs., Lyngby, Denmark
²University of Southern Denmark, Institute of Chemical Engineering, Biotechnology and Environmental Technology, Campusvej 55, 5230 Odense M., Denmark

Abstract
Biorefining agro-industrial biomass residues for bioenergy production represents an opportunity for both sustainable energy supply and greenhouse gas (GHG) emissions mitigation. Yet, is bioenergy the most sustainable use for these residues? To assess the importance of the alternative use of these residues, a consequential life cycle assessment (LCA) of 32 energy-focused biorefinery scenarios was performed based on eight selected agro-industrial residues and four conversion pathways (two involving bioethanol and two biogas). To specifically address indirect land-use changes (iLUC) induced by the competing feed/food sector, a deterministic iLUC model, addressing global impacts, was developed. A dedicated biochemical model was developed to establish detailed mass, energy, and substance balances for each biomass conversion pathway, as input to the LCA. The results demonstrated that, even for residual biomass, environmental savings from fossil fuel displacement can be completely outbalanced by iLUC, depending on the feed value of the biomass residue. This was the case of industrial residues (e.g. whey and beet molasses) in most of the scenarios assessed. Overall, the GHGs from iLUC impacts were quantified to 4.1 t CO₂-eq.ha⁻¹.demanded yr⁻¹ corresponding to 1.2–1.4 t CO₂-eq. t⁻¹ dry biomass diverted from feed to energy market. Only, bioenergy from straw and wild grass was shown to perform better than the alternative use, as no competition with the feed sector was involved. Biogas for heat and power production was the best performing pathway, in a short-term context. Focusing on transport fuels, bioethanol was generally preferable to biomethane considering conventional biogas upgrading technologies. Based on the results, agro-industrial residues cannot be considered burden-free simply because they are a residual biomass and careful accounting of alternative utilization is a prerequisite to assess the sustainability of a given use. In this endeavor, the iLUC factors and biochemical model proposed herein can be used as templates and directly applied to any bioenergy consequential study involving demand for arable land.

Keywords: agro-industrial biomass, biofuel, biorefineries, consequential LCA, iLUC, land-use changes

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Introduction
Agro-industrial biomass residues are receiving increased attention for their potential of providing sustainable bioproducts and bioenergy (e.g., Steubing et al., 2012; Tuck et al., 2012; Glithero et al., 2013; Mohr & Raman, 2013; Styles et al., 2014, 2015). In this context, integrated biorefinery solutions generating multiple outputs (e.g., fuels, electricity, heat, nutrients, and animal feed) have been acknowledged as promising alternatives to single-output technologies neglecting further utilization of coproducts (e.g., Zhang, 2008; Bentsen et al., 2009; Fatih Demirbas, 2009; FitzPatrick et al., 2010; Liu et al., 2011; Forster-Carneiro et al., 2013). Current use of these residues ranges from application on-field as organic fertilizer/soil improver (manure and straw, e.g., Hamelin et al., 2014 and Petersen et al., 2013) to utilization in the feed sector (food-industry residues). Life cycle assessments have been extensively applied in the literature to evaluate the environmental implications of agro-industrial residues (e.g., Hedegaard et al., 2008; Cherubini & Ulgiati, 2010; Spatari et al., 2010; De Vries et al., 2012a; Boldrin et al., 2013; Tufvesson et al., 2013; Falano et al., 2014), generally highlighting that the environmental performance is mainly related to the energy produced (overall energy balance) and the type of fuels displaced (related to LCA assumptions).
Existing studies, however, generally fall short in two aspects. First, the alternative/current management of the biomass residues (i.e., the reference system) is often not taken into account, thereby essentially assuming that the biomass is available free of burdens with no other uses for society. For example, if the alternative use for a biorefinery substrate is animal feeding, using the biomass for bioenergy production will induce new demands for feed ingredients that will ultimately be fulfilled by cultivating additional crops. This, in turn, is ultimately achieved through land expansion or production intensification, thereby inducing indirect land-use change (iLUC) effects with environmental consequences potentially canceling any initially foreseen benefits (Searchinger et al., 2008; Edwards et al., 2010; Searchinger, 2010; Tyner et al., 2010; Hamelin et al., 2014; Vazquez-Rowe et al., 2014). Such interactions with the feed market are aspects of the whole-system perspective that are often disregarded when residues are assessed (e.g., Koller et al., 2013; Leceta et al., 2014). In the few cases where these have been considered (e.g., Tufvesson et al., 2013), iLUC has not been included.

Second, the mass and energy balances describing the conversion of biomass substrates are generally incomplete or poorly defined, and often not linked to the substrate composition itself (e.g., Hedegaard et al., 2008). Although the overall mass flows of coproducts generated in biorefinery solutions are typically addressed in LCA, ignoring their composition changes leads to misleading conclusions regarding their potential for substituting ingredients in the feed market.

This study aims at evaluating the environmental implications of the use of agro-industrial residues for bioenergy. Such a broad aim involves two key milestones that are too often belittled in LCA studies: a methodological one and a validation one. In the former, a transparent modeling framework (i.e., determining what to include) has to be established along with the foundation for quantifying the involved emission flows. In the latter, the established methodology is tested with a specific case, so conclusions can be drawn.

In this study, the methodological milestone is tackled by developing: (i) a deterministic model for quantification of iLUC impacts (global warming, acidification, and N-eutrophication) in case of competition with the feed market, and (ii) a biochemical model for systematic quantification of all energy, mass, and substance flows (including biochemical composition and feed properties of all involved coproducts).

The validation is performed through a case study involving eight agro-industrial biomass residues (wheat straw, grass from natural areas, brewer’s grain, beet top, beet pulp, potato pulp, beet molasses, whey) and four conversion pathways (two involving bioethanol and two biogas).

Materials and methods

Goal, scope, and functional unit

The environmental assessment was performed following the ISO standard for LCA (ISO 2006a,b). In LCA, two alternative approaches can be distinguished: attributional and consequential (Finnveden et al., 2009). While attributional (or descriptive) LCA seeks at modeling suppliers using average market data and multiple-output activities using allocation, consequential (or change-oriented) LCA seeks at modeling actual affected/unconstrained suppliers and avoids allocation using system expansion (Ekvall & Weidema, 2004). As the aim of this study was addressing the environmental consequences of changing the management of the selected substrates from the reference (current system) to a number of (future) bioenergy scenarios (i.e., it is a change-oriented study), a consequential approach was adopted as it appears as the most suitable for this purpose (Weidema 2003).

The study considers a short-term time scope (period 2015–2030). The functional unit is ‘management of 1 t of biomass residue (wet)’ (input to each conversion pathway), that is, all input and output flows are scaled to 1 t of biomass input. The geographical scope was Denmark, that is, the inventory data for biomass composition, technologies and the legislation context were specific to Danish conditions.

The case study focuses on eight agro-industrial biomass residues utilized in four biorefinery pathways (Fig. 1). These involve two main energy carriers, namely biogas and bioethanol, themselves produced along with multiples coproducts (digestate for the former; ethanol molasses and a lignin-rich solid biofuel for the latter) that can be used in several ways. The full conversion chains considered for the case study are (Fig. 1) (i) production of bioethanol for transport, bioethanol molasses for biogas (combined heat and power, CHP), and solid biofuel for combustion (CHP); (ii) production of bioethanol for transport, bioethanol molasses for animal feeding, and solid biofuel for combustion (CHP); (iii) production of biogas for transport, separation of digestate (liquid for fertilization and solid biofuel for combustion (CHP)); (iv) production of biogas for combustion (CHP), digestate fate as in iii.

As illustrated in Fig. 1, the assessment considered a baseline situation and three main sensitivity analyses performed on selected scenario uncertainties. Details on these are provided in dedicated sections later in this manuscript.

Impact assessment

The following environmental impacts were included: global warming (GW; 100 years horizon), acidification (AC), aquatic eutrophication – nitrogen (AEN), and phosphorous resource saving (Pres). The impact assessment followed current recommendations for best practices within LCA (Hauschild et al., 2012). The following assessment methods were used: IPCC 2007 for GW (Forster et al., 2007), accumulated exceedance for AC.
**Agro-industrial biomass residues**

The eight residues selected for the assessment are as follows: straw, grass from natural areas (i.e., wild), brewer’s grain, beet top, beet pulp, potato pulp, beet molasses, and whey. Straw was assumed as wheat straw. These substrates include all domestic agro-industrial residues with the exception of manure, excluded because extensively addressed in recent LCA literature (e.g., Hamelin et al., 2011, 2014; De Vries et al., 2012a,b, 2013). However, the implications of codigesting the selected agro-industrial residues with manure are included.

The domestic biomass potential was quantified based on available information (details in Table S1, Appendix S1). Straw represents the largest potential (ca. 5600 Mkg ww yr⁻¹, of which 2700 are from wheat straw); however, only 38% of the total straw produced is left on-field (unutilized). Follows grass from natural areas (ca. 2000 Mkg ww yr⁻¹ corresponding to 430 Mkg DM yr⁻¹) and industrial residues (i.e., sum of brewer’s grain, beet top, beet pulp, potato pulp, beet molasses, and whey) accounting for 3500 Mkg ww yr⁻¹ (i.e., 380 Mkg DM yr⁻¹, currently used in the feed market; see Table S1 in Appendix S1).

**Alternative uses of biomass**

The straw and grass considered for the case study are ‘wastes’, that is, the grass is straw otherwise unutilized (i.e., left on-field) that are now used for bioenergy. The lost opportunity for using these two residues for bioenergy was thus modeled as ‘avoided on-field decay’ (Fig. 2). In terms of environmental consequences, diverting straw and grass to bioenergy induces carbon and nutrient losses in the soil, and this was included in the model. Full details of the modeling are provided in Appendix S9.

For the other industrial residues (brewer’s grain, beet tops, beet pulp, potato pulp, beet molasses, whey), the lost opportunity (i.e., what would have otherwise happened with these residues) is considered to be their use for animal feed. In this context, diverting the industrial residues to bioenergy induces a corresponding extra demand for marginal energy-and protein-feed, under the assumption of full elasticity of supply, as justified in Weidema et al. (2009).

The marginal for energy-feed was assumed to be maize, while for protein-feed soy meal was considered (Fig. 2). These choices are based on detailed elaboration of recent demand trends and future projections (FAPRI, 2012; FAOSTAT, 2014), as detailed in Appendix S3. For soy meal, the choice is also supported by Dalgaard et al. (2008). The energy content of the feed was modeled in terms of Scandinavian Feed Units (SFU) (definition and calculations in Appendix S5).

**System boundaries and scenario modeling**

The full system boundary considered is illustrated in Fig. 2 for straw, when used as in scenario II (bioethanol for transport). Figures S1–S5 (Appendix S2) provide detailed flow diagrams for brewer’s grains (all scenarios) exemplifying the case of industrial residues.

As illustrated in Fig. 2, the energy outputs from biomass conversion were considered to substitute marginal fossil fuel extraction and use. Coal-fired power plants and natural gas boilers were assumed as short-term marginal technologies for electricity and heat production (Appendix S3), respectively, based on the Danish government’s energy policy milestones (Danish Ministry of Climate, Energy and Buildings, 2011), where phasing-out coal and natural gas are a target for 2030 and 2035, respectively (goal-oriented approach). A sensitivity analysis was performed using natural gas-fired power plants as alternative electricity marginal.

Gasoline was assumed as marginal fuel for transport (Appendix S3). Overall, the EU transport fuel market has been decreasing with gasoline consumption steadily being reduced since the 1990s (as opposed to diesel). Increased biofuel production is therefore most likely to further offset gasoline consumption.
Bottom ashes from biomass combustion were considered to be landfilled, while fly ashes were assumed to be utilized for backfilling of salt mines with negligible environmental impacts (Fruergaard et al., 2010). Treatment of wastewater was not included.

Scenarios I, III and IV involve anaerobic digestion producing: biogas (upgraded to biomethane in scenario III), solid biofuel (undegraded solids, e.g., fibers, separated from the liquid fraction of the digestate; modeling details in Appendix S6) and mass/energy balance results in Appendix S7. The liquid fraction of the digestate (from now on simply called ‘digestate’), while we will referred to the separated solid fraction as ‘solid biofuel’) produced from this process is considered to be used as organic fertilizer (for N, P, and K), thus substituting marginal mineral N, P, and K fertilizers. Detailed calculations for the substitution ratios used in this study are provided in Appendix S8. Marginal N, P, and K fertilizers were considered to be urea, diammonium phosphate, and potassium chloride, respectively, based on recent demand trends and expected capacity installations (IFA, 2014; details in Appendix S3).

For all scenarios involving anaerobic digestion, a wet digestion process was considered, which was here modeled as codigestion with animal manure. In the baseline scenarios, the alternative management (reference scenario) of the animal manure was considered to be digestion itself. In other words, it was assumed that this manure would have been digested anyway. This assumption is supported by several western European national renewable energy action plans (NREAP) (Beurskens et al., 2011) indicating substantial increases in biogas production from manure, thereby suggesting manure digestion as an established practice in the near future. For Denmark, the goal is that up to 50% of livestock manure can be used for bioenergy in 2020 (Danish Government, 2009).

The alternative counterfactual for manure (conventional storage and application on land) was assessed in a sensitivity analysis (Appendix S10) following the approach of Hamelin et al. (2011).

Deterministic model for indirect land-use changes (iLUC) impacts. A modeling framework was developed and applied to quantify the iLUC impacts caused by changes in the demand for land. Effects associated with demand for land were considered to be global, given the global nature of agricultural commodity

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**Fig. 2** Process flow diagram for straw for scenario II: production of bioethanol-for-transport, bioethanol molasses for feed, solid biofuel for combustion (CHP). CHP, combined heat and power; comb., combustion; exp, arable land expansion; int, intensification of agricultural production; PF, palm fruit; SB, soybean; SFU, scandinavian feed units. The “loop” symbols indicate that the “soybean meal loop” modeling framework (Dalgaard et al., 2008) was applied to quantify the avoided soy meal/maize and the induced palm oil.
The share of the response to changes in demand for land attributed to intensification and expansion was calculated by elaborating data on crop yields and production from FAOSTAT (2014) (Eqs. S10-S16, Appendix S4), following the approach suggested in Schmidt et al. (2012). As a result, it was found that for the period 2000–2010, expansion was responsible for 25% ($\bar{p}_{\text{exp}}$) of the response, while intensification for 75% ($\bar{p}_{\text{int}}$).

**Type and location of arable land expansion**

On the basis of Gibbs et al. (2010), two main aggregated land-cover types were considered as areas where expansion takes place: forest and low woody vegetation of natural origin (including savannah, cerrado, shrubland, and grassland). The latter is here referred to as 'shrubland'. The proportion of arable land deriving from ‘forest’ and ‘shrubland’ for the various regions of the world was quantified based on the findings of Gibbs et al. (2010) (Table S7, Appendix S4). According to this, overall ca. 83% of the expansion occurred on forests and ca. 8% on shrublands (period 1990–2000). Contrary to some iLUC studies (e.g., Kløverpris et al., 2010), no distinction between the conversion induced by arable versus pasture land demand was considered. This was carried out because historical data (1990–2010) show comparable expansions for both; from FAOSTAT (2014), it can be calculated that between 2000 and 2010, the average annual gross expansion of arable land and cultivated pasture land was 11.6 and 12 Mha yr⁻¹, respectively (and 51 vs. 53 Mha yr⁻¹ between 1990 and 2000). On this basis, it is assumed that both arable and pasture land demand will trigger expansion in similar proportions also in the mid-term future.

On the basis of IPCC (2006b), ‘forest’ losses were further divided into types (tropical, subtropical, temperate, boreal and polar; each, except the latter, further subdivided in relevant vegetation categories, for example, ‘coniferous’, ‘tundra’ and ‘mountain’ for boreal forest). This subdivision for forest losses was necessary in order to assign specific carbon emission factors to each type of biome loss (Appendix S4, section 4.2.2). Additional subdivision was not necessary for ‘shrubland’ losses as this type of land conversion occurs in the tropical biome only (Gibbs et al., 2010). This led to establish 20 different categories of area loss (19 ‘forest’ and 1 ‘shrubland’ categories). Breaking down the world in 10 aggregated regions and using the deforestation data 2000–2010 from FAO (2010), the proportion of area loss was determined for all 20 area categories in each of the 10 world regions, resulting in a total of 83 region- and biome-specific areas of land cover loss (CEF, Eq. S17, Appendix S4). Gross (rather than net) deforestation losses were used for the calculation, as it is assumed that afforestation programs do not affect agricultural land expansion.

**Carbon and nitrogen emissions from expansion**

For quantifying the carbon emissions, it is necessary to first quantify an average annual growth (as biomass C, at natural state) in each individual biome during the whole time horizon considered for the global warming potential (100 years; GWP100). To this end, biomass growth data from IPCC (2006b) were used to calculate CO₂-C emission factors (CEF) in each of the 83 region- and biome-specific areas of land cover loss (CEF, Eq. S17, Appendix S4). This, combined with the earlier
step of establishing the proportion of land cover loss in each of the 83 region- and biome-specific areas, allowed deriving a global CEF associated with arable land expansion (Eq. S20, Appendix S4): 2.4 t CO$_2$-C ha$^{-1}$ yr$^{-1}$. Similarly, N-emission factors (NO and N$_2$O) were calculated based on the CEFs (Eqs. S21–S25; Appendix S4) as IPCC provides emission factor per unit dry matter loss (CEF and dry matter are related by carbon content). These were as follows: 7.1 kg NO ha$^{-1}$exp yr$^{-1}$ and 0.88 kg N$_2$O ha$^{-1}$exp yr$^{-1}$.

Increased N, P, K fertilizer use for intensification, and related emissions

Based on the best available data for global fertilizer consumption (IFAS, 2014), the annual change in global N, P, and K fertilizer use was quantified for all years in 2000–2010: on average this was 2.24 Mt N yr$^{-1}$ for N, 0.77 Mt P$_2$O$_5$ yr$^{-1}$ for P, and 0.54 Mt K$_2$O yr$^{-1}$ for K (Table S12; Appendix S4). These values were further split into shares for expanded land (Eq. S30; Appendix S4) and intensified land (calculated by difference, Eq. S29). The global fertilizer use for intensification was then calculated as the mean value of the individual years (Eq. S31; Appendix S4): 166 kg N ha$^{-1}$int yr$^{-1}$, 68 kg P$_2$O$_5$ ha$^{-1}$int yr$^{-1}$, and 47 kg K$_2$O ha$^{-1}$int yr$^{-1}$. The N emissions following fertilizer application on land were directly quantified based on N-emission factors (Eq. S32–S36) for the marginal N fertilizer considered in this study (urea); 1.5% for N$_2$O-N (IPCC, 2006a), 2% for NH$_3$ (Hamelinck et al., 2004). The final aggregated inventory for land use is summarized in Table 1.

Biochemical energy model. For proper accounting of biochemical conversion of the substrates, a biochemical energy conversion model was developed in Matlab (c, The MathWorks Inc., version R2012) and integrated with the LCA model (Fig. S8; Appendix S6). With the (bio)chemical composition of each biomass substrate as the starting point of the modeling, fuel energy production (bioethanol or biogas) and the composition of the associated coproducts (bioethanol molasses, solid biofuels, digestates, etc.) were determined based on detailed energy, mass, and substance balance calculations. Inputs such as specific hydrolysis efficiencies, fermentation yields, separation efficiencies, and parasitic energy consumptions for each individual unit process were included. Detailed modeling equations are provided in Appendix S6. The application of the model to the investigated energy conversion pathways is described below.

Biomass composition

The biochemical composition of the eight selected biomasses was derived from a review of 40 studies reporting composition data (details in Appendix S5, Tables S16–S23), including sugar contents (sucrose, starch, cellulose, hemicellulose, and pectin), lipids, proteins, lignin, ash, water, carbon, higher and lower heating value (HHV and LHV), nitrogen, phosphorous, potas-

| Table 1 Final aggregated inventory for arable land use (one hectare of arable land demanded)** |
|---------------------------------|--------------|---|
| Expansion                        | Emissions to air                      |                           |                           |
|                                  | CO$_2$     | 2.2 t CO$_2$ ha$^{-1}$ dem yr$^{-1}$ |
|                                  | N$_2$O     | 0.22 kg N$_2$O ha$^{-1}$ dem yr$^{-1}$ |
|                                  | NO$_x$     | 1.8 kg NO ha$^{-1}$ dem yr$^{-1}$     |
| Intensification: NPK fertilizer production |
| Materials                        | N-fertilizer | 125 kg N ha$^{-1}$ dem yr$^{-1}$ |
|                                  | P-fertilizer | 52 kg P$_2$O$_5$ ha$^{-1}$ dem yr$^{-1}$ |
|                                  | K-fertilizer | 35 kg K$_2$O ha$^{-1}$ dem yr$^{-1}$   |
| Intensification: N emissions Emissions to air |
|                                  | N$_2$O (dir + ind)* | 3.4 kg N$_2$O ha$^{-1}$ dem yr$^{-1}$ |
|                                  | NH$_3$     | 3 kg NH$_3$ ha$^{-1}$ dem yr$^{-1}$    |
|                                  | NO$_x$     | 4.5 kg NO$_x$ ha$^{-1}$ dem yr$^{-1}$  |
| Emissions to water NO$_3$-N | 25 kg NO$_3$-N ha$^{-1}$ dem yr$^{-1}$ |

**Sum of direct and indirect N$_2$O.

Bioethanol production modeling

Bioethanol production (scenarios I–II) was assumed to occur from C$_6$-sugars hydrolyzed from cellulose. Cellulose and hemi-cellulose hydrolysis efficiencies were set to 95% and 75%, respectively, based on Hamelinck et al. (2005). Nonhydrolyzed and unconverted sugars (both C$_5$ and C$_6$), along with unconverted lipids, proteins, and lignin were routed to a mixed residual stream, later separated into a liquid fraction (bioethanol molasses), and a solid fraction by centrifuging (see Appendix S6 for more details; Figs. 2 and S1–S5 exemplify mass balances for selected substrates; all mass/energy balances are detailed in Appendix S7). The solid biofuel (dry matter, DM, ca. 40% at this stage) was further heat-dried to 90% DM for cofiring in power plants. Bioethanol molasses were also considered to be heat-dried with waste heat from the distillation unit to about 65% DM to facilitate transportation (for use as feed or substrate for biogas) conformingly with current practice (Larsen et al., 2008, 2012).

Biogas production modeling

Anaerobic digestion was modeled as mesophilic codigestion of the individual biomasses (scenarios III–IV) or bioethanol molasses (scenario I) with pig manure. Manure addition is needed to ensure a sufficient level of micronutrients for the digestion process, as well as to adjust the high C:N ratio (above 25) for most biomasses; Table S15 in Appendix S5) to a more optimal level (Mshandete et al., 2004; Alvarez et al., 2010; Ngene et al., 2012).
The modeled theoretical methane potential for the biomasses ranged from 430 (potato pulp) to 540 (brewer grains) Nm³ t⁻¹ VS (Table S15; modeled based on the substrate composition, as described in Symons & Buswell, 1933). For manure, it corresponded to 450 Nm³ t⁻¹ VS (Hamelin et al., 2011). The mixture of biomass and pig manure can be calculated by different approaches: by concurrently optimizing the energy output, C : N ratio and dry matter in the digestate, or according to a legislative context. Here, a manure: biomass ratio of 3 : 1 (dry basis) was assumed, reflecting the Danish legislation (Danish Ministry of Food, Agriculture and Fisheries, 2006, 2012).

Electricity consumption (8% of the electricity produced) was assumed according with average literature values (Boerjesson & Berglund, 2006; Hamelin et al., 2011; Bacenetti et al., 2013), and heat consumption was calculated as the energy required to heat the substrates from 8 to 37 °C.

Methane losses from the digesters were assumed to 1% of the produced CH₄, assuming the implementation of best available technologies and management practices in accordance with recent LCA studies (Hamelin et al., 2011; De Vries et al., 2012b; Tonini et al., 2012; Hamelin et al., 2014). Further details on anaerobic digestion modeling are provided in Appendix S6, section 6.2.

Energy use of biorefinery products

Biogas combustion for CHP production (scenarios I, IV) was assumed to occur in a gas engine with net electricity and heat efficiencies of, respectively, 45% and 55%, relative to the LHV of the biogas (Danish Energy Agency, 2012), and the related air emissions were based on Nielsen et al. (2010) (Table S25).

Biogas upgrading to methane and tailpipe emissions associated with combustion in vehicles (scenario III) were based on data provided in Jungbluth et al. (2007) (Appendix S6). Upgrading was based on CO₂ removal technologies (pressure swing adsorption, PSA), considering an electricity consumption of 0.014 kWh MJ⁻¹ CH₄.

Combustion of the solid biofuels (coproduced with bioethanol in scenarios I-II; solid fractions of the digestate in scenarios III and IV) was modeled as cofiring in large-scale coal-fired CHP plants with net electricity and heat efficiencies of 38% and 62%, respectively (after Tonini et al., 2012). Air emissions (Table S25) were based on Nielsen et al. (2010).

Use-on-land of biorefinery products

The amount and composition of the liquid fraction of the digestate derived from anaerobic digestion of the biomasses was calculated based on a mass balance approach, that is, as the difference between the initial nutrients and dry matter fed to the digestion process and the amount transferred to the biogas and to the recovered solid fraction (Table S28, Appendix S7). For whey and beet molasses, no solid biofuel could be recovered as lignin is not found in their composition (mostly composed of easily degradable sugars).

The emission and leaching of nutrients were quantified as follows: direct N₂O-N emissions were calculated equal to 1.5% of the N applied with the digestate (IPCC, 2006a). The emission of NH₃-N was modeled as 11% of the N in the digestate, which is the average figure calculated from a number of studies (Amon et al., 2006; Bruun et al., 2006; Clemens et al., 2006; Matsunaka et al., 2006). The emission of NOₓ-N was assumed 1.1% of the N in the digestate based on Hamelin et al. (2012). The leaching of N (as nitrates) was calculated equal to 51% of the digestate N content based on Hamelin et al. (2011) (or, alternatively, equal to 45% of the N in the digestate after subtracting the emission of NH₃-N). The indirect N₂O emissions (i.e., N₂O produced subsequent to the re-deposition of the emitted NH₃, NO, and leached N) were quantified based on IPCC (2006a).

Losses of P to soil and water were considered to correspond to 5% of the P applied in excess, based on Hamelin et al. (2012). The K losses to soil and water were not further considered, as not affecting the environmental categories considered.

The share of the applied C that entered the soil C pool and that was emitted as CO₂ was determined based on the findings of Bruun et al. (2006). Based on the average values reported in Bruun et al. (2006) for arable loamy soil in West Denmark, it was considered that 20% of the initial C applied is emitted as CO₂ after 10 years, 67% after 50 years, and 90% after 100 years (the degradation kinetic is described in Appendix S9, section 9.3).

Transport and other processes

Overseas transport distances for soy meal and maize were included based on expected sea-transport distances. These were 11400 km for maize and 12000 km for soy meal. Palm meal was assumed to be used locally for the south-east Asian market (no overseas transport). Crops inland transportation of 400 km (with trucks) was considered. Transport of the selected biomass residues to the energy conversion facility was assumed to occur locally (30 km). Transport distances for energy conversion residues were as follows: 25 km for digestate, 80 km for bottom ash, and 500 km for fly ash based on typical EU figures (Tonini et al., 2013).

The following impacts were disregarded: (i) impacts associated with capital goods (i.e., materials and construction/demotion of facilities); (ii) impacts associated with preparation of the feed meals from the crops (e.g., soy meal, palm meal, maize feed). In both cases, this is due to the poor information available, and it was judged that the importance of these impacts in the whole system was largely outweighed by the uncertainty it comes with.

Sensitivity and uncertainty analysis. Sensitivity and uncertainty analyses were addressed at two levels: (i) scenario uncertainties and (ii) parameter uncertainties, following the tiered approach suggested in Clavreul et al. (2012). Uncertainties related to the LCA methodology itself were not addressed.

Scenario uncertainties (related to assumptions) were addressed by three main sensitivity analyses: (S1) No treatment and use-on-land for the reference manure management (instead of digestion and use-on-land; applies to scenario I, III, and IV); (S2) natural gas power plant as marginal technology for electricity generation (instead of coal power plant; applies to all scenarios); (S3) 100% (maximum) recovery of proteins in the...
bioethanol–molasses flow (instead of the baseline value of 33%; Appendix S6 (Table S24); applies to scenario I and II). These sensitivity analyzes are summarized in Fig. 1.

Parameter uncertainties were addressed by assigning mean value and standard deviation to the parameters used in the modeling (Table S37, Appendix S10), assuming normal distributions. While for some parameters (crop yields, biomass composition), a normal distribution behavior around the mean was observed based on statistical data, for others (technology efficiencies, N emissions), only a minimum–maximum range around the mean was available. For simplicity, this was still approximated with normal distribution to give ‘higher probability’ to the average value, even though other approaches may be applied (e.g., Clavreul et al., 2013). All scenarios were modeled using Monte Carlo analysis (1000 simulations): for each individual biomass, scenarios I–IV were compared against each other to evaluate the number of occurrences when one scenario was performing better than another.

Results

GW impact from the biorefinery scenarios

The GW environmental impacts related to the 32 bioenergy scenarios are shown in Fig. 4a. Figure 4b, c illustrate the GW breakdown for energy recovery and iLUC, respectively. Net impacts/savings for the individual bioenergy scenarios were obtained by subtracting the avoided impacts (negative values in the figures) from the induced impacts (positive values). The zero axis represents the reference: net values below the zero axis indicate environmental improvements compared with the reference scenario (in which: electricity, heat, and transport are provided by coal, natural gas, and gasoline, respectively, and the biomass residues are used for feeding or left decaying on-field in the case of grass and straw). Although the functional unit is the management of 1 tonne of residue (wet weight), results are shown per tonne dry weight for illustrative purposes, as the breakdown of the impacts for wet substrates would otherwise not be visible to the reader (scaling problem as substrates have very different water content). Results per tonne of wet weight are nevertheless shown in Appendix S10 (Table S55).

Two major results can be observed: i) from a substrate perspective, the residues not competing with the feed sector (straw and grass) always provided GW benefits, across all scenarios; ii) from an energy pathway perspective, scenario IV (biogas and solid fuel for CHP) outperformed the remaining scenarios. The first observation reflects the importance of the iLUC impacts, this applying whenever a changed demand for land is induced, here for all substrates but straw and grass for which no competition with the feed market was involved. The reason for the second observation is the high net electricity recovery (substituting for coal-based electricity) involved in that scenario. This was expected, as this scenario, by its very nature, involves that a maximum of biomass C is converted to electricity. Scenario IV still outperformed the other scenarios when substituting natural gas-based electricity instead of coal-based electricity (Fig. 7; sensitivity S2), although the difference with the remaining scenarios significantly decreased.

The magnitude of the iLUC impacts was directly related to the nutritional value of the biomass: whey, brewer’s grain, and beet top showed the highest GWP100 for iLUC impacts in scenarios I, III, and IV where 100% of the input biomass was used for bioenergy production (GWP100 varying between 1.2 and 1.4 t CO2-eq. t−1 DM, Fig. 4c). This was a direct consequence of the high proteins and energy content (high SFU) of these residues relatively to the other biomass (Table S15; Appendix S5). However, when part of the input biomass was recovered as animal feed (scenario II: thereby partially avoiding iLUC), the iLUC GHG impacts were significantly decreased (by about 44% on average, across all biomasses; Fig. 4c). These results emphasize the importance of a rigorous methodology to quantify the quality of the coproduced feed: in existing LCA literature of biorefineries, very little attention is paid to the potential avoidance of iLUC through feed recovery.

The impact of biogas production on the total induced GW impact was significant (ca. 50% of total) for scenario III due to the PSA biogas upgrading process where, on top of the high energy consumption for compression, the CO2 in the biogas is removed and lost to the atmosphere without providing any services. Emerging upgrading technologies where the CO2 in the biogas is instead converted to more methane could thus improve this process significantly. For the remaining scenarios, the impacts from the production of the energy carrier itself accounted for one-third of the induced impacts. For energy recovery (Fig. 4b), the major GHG savings were associated with biogas and solid biofuel combustion in CHP, due to the substitution of coal-based electricity and to the overall high energy recovery in CHP units, these reflecting the highly efficient technologies in place in northern Europe. Gross production of transport biofuel was significantly higher in scenario III (as biogas for transport) compared with I–II (as bioethanol). However, due to the energy input required for upgrading in scenario III and the C losses involved in that process (CO2 and fugitive CH4), the overall GHG balance was in favor of scenarios I–II. This highlights the important role of future innovations on biogas upgrading technologies to decrease such impact and render biogas a more attractive fuel for transport.

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Other impacts from the biorefinery scenarios

As shown in Fig. 5a, b, for each individual biomass residue, AC and AEN impacts were comparable across scenarios I, III, and IV, while they were significantly lower in scenario II. This was due to the fact that scenario II involves (i) no use-on-land of digestate and associated N emissions; (ii) lower iLUC impacts; and (iii) lower net demand for crop cultivation. Use-on-land, iLUC, and related crop cultivation were, in fact, the most important contributors to the N-related impact categories.

In the category AC, a significant impact was associated with oversea transport of the “compensatory” feed crops (soy meal and maize) shown under the process ‘Other’. This did not apply to straw and grass as for these substrate no competition with the feed market applies.
Results for AEN are in accordance with previous studies involving use-on-land (e.g., Hamelin et al., 2011, 2014; De Vries et al., 2012b; Tonini & Astrup, 2012; Tonini et al., 2012). When combusted (as a transport fuel or for CHP), the N as a macronutrient is completely oxidized, while when recovered and applied on-land, only a fraction is lost (volatilization, other gaseous loss, leaching, etc.) a consistent portion is subject to leaching,
depending upon plants N-uptaking efficiency. As shown in Fig. 5, the impact due to N-leaching (i.e., UOL (digestate) is not completely counterbalanced by the avoided application of mineral fertilizers (these also involve losses), reflecting the high efficiency of ammonium fertilizers under European conditions (Hansen et al., 2006). For some of the biomasses (e.g., brewer’s grain, whey, beet molasses, beet top), contributions from use-on-land to the total impacts were very important as a result of their relatively high N (protein) content.

P-resource depletion was dominated by consumption of P mineral fertilizer for cultivation of the compensatory crops providing compensation for the energy- and protein-feed diverted from feeding to energy market, that is, those reacting to the demand change. For some biomasses with significant P content (whey and brewer’s grain), this was partly counterbalanced by substitution of mineral P fertilizer following digestate use-on-land. Scenario II displays a better performance due to the lower need for compensatory crops, compared with the other scenarios.

Indirect land-use change impacts

Based on the inventory results obtained with the iLUC model (Table 1), the characterized impacts of demanding 1 hectare of arable land (1 ha\textsubscript{dem}) were quantified and illustrated in Fig. 6. The net GWP\textsubscript{100} was 4.1 t CO\textsubscript{2}-eq. ha\textsuperscript{-1} dem yr\textsuperscript{-1}. The highest contribution came from expansion (ca. 2.3 t CO\textsubscript{2}-eq. ha\textsuperscript{-1} dem yr\textsuperscript{-1}). Intensification contributed with 1.8 t CO\textsubscript{2}-eq. ha\textsuperscript{-1} dem yr\textsuperscript{-1} as the sum of NPK fertilizers production (ca. 0.85 t CO\textsubscript{2}-eq. ha\textsuperscript{-1} dem yr\textsuperscript{-1}) and N-fertilizer emissions in the form of N\textsubscript{2}O (ca. 1 t CO\textsubscript{2}-eq. ha\textsuperscript{-1} dem yr\textsuperscript{-1}).

Concerning the impacts on the remaining categories, intensification-related impacts were by far the largest contributors. While for the AC category fertilizers production (due to energy consumption) and N emissions (following use-on-land) were equally important, the latter were by far the most important contributor for AEN. As expected, for the category P-resource depletion (Pres), the impact was entirely associated with P-fertilizer production.

Sensitivity and uncertainty analysis

Key sensitivity analyses (S1–S3) results are presented in Fig. 7. The most important scenario assumption was related to the reference scenario for pig manure management (Fig. 7; Fig. S10 illustrates the breakdown of the impacts): when including savings associated with avoided conventional manure management (i.e., storage and application on-land without any treatment), all net results turned into savings for scenarios III and IV in all impact categories with the exception of Pres (when manure was digested, a fraction of the manure P was associated with the solid fraction used for combustion, thereby decreasing the on-land P recovery). For scenario I, however, avoiding conventional manure management allowed for net savings only for GW.

While it can be debated whether the default reference scenario for animal manure should be storage and direct application on-land without any treatment (sensitivity S1) or digestion (baseline), an a-priori inclusion of the large savings associated with avoiding manure storage and direct application on-land would cloud the impacts/savings associated specifically with the studied biomass residues. Essentially, the results from sensitivity S1 simply illustrate that raw manure digestion is environmentally beneficial compared with conventional storage and direct use-on-land. As such, manure digestion should be promoted regardless of the availability of cosubstrates, and this is the situation reflected in the baseline scenario.

Sensitivity S2 (natural gas for marginal electricity instead of coal) essentially showed that the GW savings from scenario IV would be much reduced by replacing a ‘cleaner’ electricity fuel, rendering biogas for transport more attractive (Fig. 7). Regardless of the choice for marginal electricity, the ranking between scenarios did not change. As such, the choice of marginal electricity

![Fig. 6](https://example.com/fig6.png)  
**Fig. 6** Breakdown of the environmental impacts (for GW, AC, AEN, and Pres) associated with demanding one hectare of arable land (1 ha\textsubscript{dem}) for crop cultivation.

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was not critical for the overall conclusions regarding preferred scenarios.

For Sensitivity S3 (100% protein recovery, Fig. 7), it is worth emphasizing that for high protein-containing biomasses (e.g., brewer’s grain and beet top), the GHG impacts would significantly decrease (about half of the baseline) if all proteins were recovered as part of the bioethanol conversion process (also assuming that the bioethanol molasses are used for animal feeding, as in scenario II).

The Monte Carlo analysis of parameter uncertainties confirmed that scenario IV was the most favorable conversion pathway in a GW perspective, under baseline assumptions (Table S36, Appendix S10).

Discussion

Methodological milestone: iLUC model

This study provides a deterministic (in other studies also referred to as causal descriptive, or biophysical, or agro-physical) approach for quantification of indirect land-use changes induced by demands for crops and applies it to 32 biorefinery pathways.

The GW impact calculated with this model (4.1 t CO₂-eq. ha⁻¹ dem yr⁻¹) appears lower than previous estimates elaborated with CGE (compute general equilibrium) models: in Tonini et al. (2012) and Hamelin et al. (2014), the iLUC GHG impact was quantified to 16–18 t CO₂ ha⁻¹ dem yr⁻¹ based on GTAP results (Global Trade Analysis Project). In that case, the higher value was due to assuming a 20-year amortization time (i.e., the initial figure was ca. 310–350 t CO₂ ha⁻¹ dem) and to the fact that intensification accounted for only 30% of the total response compared with 75% in this study. For maize ethanol, Searchinger et al. (2008) estimated an iLUC of about 12 t CO₂ ha⁻¹ dem yr⁻¹ (30 year amortization) using data from FAPRI (Food and Agricultural Policy Research Institute). A comprehensive review from Broth et al. (2013) found a range of ca. 130–350 t CO₂ ha⁻¹ (ca. 4–12 t CO₂ ha⁻¹ dem yr⁻¹ assuming 30 years amortization) based on a range of different CGE studies.

Regardless the temporal issue related to the C emission (addressed in the next paragraph), general (e.g., GTAP) or partial (e.g., FAPRI) economic equilibrium models have a different scope (compared with deterministic models specifically developed for LCA) as they aim at modeling how global/partial market segments reach an equilibrium (a balance) in a relatively short-term after inducing a particular shock in the market (e.g., a specific biofuel policy/strategy), in comparison with a business-as-usual scenario (e.g., Igos et al., 2015). They thus include short-term effects on prices and related price elasticities (which is not consistent with the principle of ‘full elasticity of supply’ typically applied in consequential LCA, see Weidema et al., 2009). Bearing in mind this difference, these two types of models (economic and deterministic) should be seen as complementary rather than competing as they serve different purposes as highlighted by Schmidt et al. (2015).

In this study, the temporal issue related to the C emission (due to nonproportionality between annual occupation for cropping and initial transformation of land) is overcome by quantifying the C losses related to the lost opportunity of growing biomass during a year (as in this generic year, the land is demanded/occupied for crop cultivation, being released the following year for uses/crop other than that for which the land was initially demanded). The idea is to find a function describing the annual growth of biomass (t C ha⁻¹ yr⁻¹). However, this ideally requires having specific information on the age of the vegetation in each individual biomes affected by deforestation. To overcome this, an average annual growth (as biomass C, at natural state) was calculated for a time horizon of 100 years.

Other methods depart from the total amount of biomass stock cleared (as t C ha⁻¹) and apply amortization periods of 20 years (e.g., PAS 2050, 2011 and WRI/WSB, 2013) to attribute the C emission to one generic year (t C ha⁻¹ yr⁻¹); this 20-year time horizon is also recommended by IPCC (2006c) and the Renewable Energy Directive (European Union, 2009) when addressing the time span during which a land remains in a conversion category after a change in land-use. An alternative approach, avoiding amortization, is proposed by Schmidt et al. (2015) where it is assumed that the effect of 1 ha y occupation is simply that deforestation is brought 1 year forward in time (anticipated emissions). By doing so, the net C emissions from deforestation become null, but effects on GW are considered by ‘discounting’ the C emissions applying a time-
weighted version of the GWP method. However, it can be argued that this approach, while avoiding amortization in the inventory phase of the LCA (LUC modeling), introduces/moves it later on in the characterization phase whose results will be totally affected by the choice of 20, 50, 100 years ‘discounting’ of the GWP.

As there are no agreements on which temporal approach should be applied, the analytical framework used in this model allows for accommodating it to any of the temporal approaches described above. Additionally, the model has been developed with a consequential approach; however, it can be adapted to an attributional approach by considering: (i) the total annual arable land in use (instead of the additional annual arable land in use, in Eq. S29-S31) and (ii) the total annual consumption of NPK fertilizers (instead of the additional annual consumption, in Eq. S28-S36). By doing so, the related iLUC figures are likely to decrease as they would be distributed according to all the land in use.

All in all, the model presented in this study offers the advantage of being transparent and replicable, as all input data can be found in available and up-to-date statistics (e.g., FAOSTAT, 2014) and all equations are thoroughly detailed in the appendices. The iLUC model developed herein and its resulting emission factors per hectare demanded can be applied in any consequential LCA study involving demand for arable land and could be further detailed using up-to-date statistics or be tailored to specific regions (rather than to the whole globe), following the equations detailed in Appendix S4. Similarly, the model can be applied for other time series, for example, using predictions on crop yields and fertilizers use from FAPRI (2012).

Methodological milestone: the biochemical model

The biochemical model developed in this study allows linking the quality and composition of the biorefinery coproducts to the substrate input composition. This is important for quantifying the potential for substituting ingredients in the feed market, as well as for estimating energy potentials. This both regards bioethanol and biogas refinery processes.

For anaerobic digestion, while only few data are available on separation, recovery, and composition of solid fractions from digestates, this approach allows quantifying their energy and mass flows based on rigorous balances. To provide an example, in the case of straw anaerobic digestion, not accounting for the energy recoverable by separating the solid fraction in the digestate would lead to underestimate the total energy recovery by about 30% (energy balances detailed in Table S28, Appendix S7). The same applies to all substrates having significant lignin content. It is worth noticing that such state-of-the-art technologies are becoming more and more established (an example is Maabjerg Bioenergy Drift A/S, 2014).

The same approach based on mass/energy balances allows estimating the quality of the bioethanol molasses flow given a specific input composition. While this may depend upon hydrolysis and separation efficiencies, the model can easily be tailored to specific processes through changes in parameter values. The approach finally allows estimating feed and energy potential of the bioethanol molasses flows (i.e., the liquid fraction after fermentation) which are then directly related in the LCA to the substitution of ingredients/energy in the feed/energy market.

Key learnings from the biorefinery case study

One of the key messages of this study is that residual biomass, albeit residual, can involve substantial environmental impacts exceeding those of the reference system (where residual biomass is used for feeding and fossil fuels are used for energy). In this study, it happened in particular for residues having considerable feed value (SFU ≥ 1; Table S15, Appendix S5). With the selected industrial residues as illustrative examples, we showed how bioenergy use incurred higher impacts than the alternative use (feeding). Similar trends were found in Tufvesson et al. (2013), although the magnitude of the impacts was lower than in this study. This is due to the fact that this study also included iLUC impacts. This highlights the importance of accounting for the upstream impacts associated with the alternative management of the biomass, including potential land related effects.

As opposite to the industrial residues, straw and wild grass (from natural areas) were identified as promising substrates for bioenergy, as their use for the energy appeared preferable to their current alternatives in all the environmental categories investigated. The results for straw are in agreement with the findings of recent LCA studies (Cherubini & Ulgiati, 2010; Boldrin et al., 2013; Hamelin et al., 2014; Turconi et al., 2014). This also applies to grass (Roesch et al., 2009; Recchia et al., 2010; De Vries et al., 2012b). These results are of course conditional to the counterfactual being that these residues are left on-field. In a future where it is economically profitable to use these for feed, their environmental performance when used for bioenergy would then be tightly connected to their feed value, as for the industrial residues investigated in this study.

Focusing on the fate of bioethanol molasses, it is worth noticing that using these for feed (scenario II) was, for most of the substrates, environmentally prefer-
able (or at least comparable) to using it as a feedstock to produce biogas for combustion in CHP (scenario I). This would not be the case if iLUC was disregarded, highlighting their importance when comparing biorefinery solutions involving feed coproducts.

Among the conversion pathways considered, the results of this study essentially confirmed the findings of previous studies highlighting conversion pathways involving electricity and heat provision (CHP) as environmentally advantageous over those involving liquid fuels, under the current energy system (among the others: Steubing et al., 2012 and Hedegaard et al., 2008).

Implication and perspectives

For biomass residues that may otherwise be used for feed/food production, this study clearly indicates that contributions from iLUC to the total environmental impacts, although uncertain, may be substantial and potentially cancel any benefits related to fossil fuel displacement. As demonstrated in this study, this is not only true for GHG emissions but also for other impact categories affected by intensification and crop cultivation.

On the basis of these results, future research and renewable energy policies should carefully consider the possible alternative utilization of residual biomasses, agree on a methodology to quantify the feed value of these products and prioritize bioenergy solutions coproducing high-quality feed/food. Currently, biomass residues are often considered burden-free by researchers and/or stakeholders. This is unlikely the case.

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Supporting Information

Additional Supporting Information may be found in the online version of this article:

Appendix S1. Agro-industrial biomass residues in Denmark: potential.
Appendix S2. Scenarios modeling and system boundary.
Appendix S3. Identification of marginal technologies/products.
Appendix S4. Model for indirect land-use changes (iLUC).
Appendix S7. Mass, energy, and nutrients balances.
Appendix S8. Substitution of mineral fertilizer.
Appendix S9. Decay processes on-field.
Appendix S10. Additional LCA and sensitivity/uncertainty results.