


Article

Are Renewables as Friendly to Humans as to the Environment?: A Social Life Cycle Assessment of Renewable Electricity

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Abstract: The adoption of renewable energy technologies in developing nations is recognized to have positive environmental impacts; however, what are their effects on the electricity supply chain workers? This article provides a quantitative analysis on this question through a relatively new framework called social life cycle assessment, taking Malaysia as a case example. Impact assessments by the authors show that electricity from renewables has greater adverse impacts on supply chain workers than the conventional electricity mix: Electricity production with biomass requires 127% longer labor hours per unit-electricity under the risk of human rights violations, while the solar photovoltaic requires 95% longer labor hours per unit-electricity. However, our assessment also indicates that renewables have less impacts per dollar-spent. In fact, the impact of solar photovoltaic would be 60% less than the conventional mix when it attains grid parity. The answer of “*are renewables as friendly to humans as to the environment?*” is “*not-yet, but eventually.*”

Keywords: renewable energy; supply chain; social responsibility; social life cycle assessment; labor conditions; Malaysia; solar PV; Biomass; Hydro

1. Introduction

Countries around the globe are competing for the increased adoption of renewable energy technologies, and developing nations are leading this trend in the aim of meeting the growing electricity demand in a sustainable manner. Various studies have demonstrated the positive environmental externalities and macroeconomic effects of such initiatives, including their impacts on GDP, unemployment, and balance of trade [1–4]. However, there are fewer studies that examine the social impacts of renewable energy development quantitatively. This should be a point of concern for developing nations, where worker often suffer from poor labor conditions as part of global supply chains. For this reason, the adoptions of renewable energy technologies should be assessed not only from environmental and economic perspectives but also from the social responsibility perspective as well.

The question the authors intend to discuss through this analysis is, “*are renewable energy technologies as friendly to humans as to the environment?*” This analysis tries to answer this substantial question quantitatively with the help of a relatively new framework called the *social life cycle assessment*. The rest of the article is structured as follows in conformity with ISO 14040/14044. Section 2 explains the goal

and scope, methods and data employed in this analysis. Section 3 presents the results of the impact assessment. Finally, based on the results, Section 4 presents the interpretation and conclusions.

2. Methods

2.1. Goal and Scope

The goal of this analysis is to assess the adverse social impacts of renewable electricity production in a developing nation, and to compare them to that of the conventional electricity production mix. For this, the authors choose Malaysia as the case example. Estimated to increase its gross domestic product at about 5–6% annually until 2020, the electricity generation in Malaysia is also expected to grow significantly with 3.5% annual growth [5]. To meet the growing electricity demand in a sustainable manner, the Malaysian government has launched the National Renewable Energy Policy in 2010, setting a renewable energy target of 11% of the total energy mix by 2020. Local policymakers acknowledge the need to capture the environmental, economic, and social impacts of renewable energy development to accelerate this rapid expansion, which push them to publish statistical data on renewable energy projects for research purposes. This makes the country a very fitting case for this analysis. While this analysis focuses on Malaysia, the results should be representative of how renewable development in southeast Asian nations would affect the labor in the electricity supply chain, due to the similar renewable cost structures in these nations.

The products assessed in this analysis were electricity from solar PV, biomass power, hydropower and the current electricity production mix in Malaysia. The system considered in this study was *cradle-to-gate*, or the product life cycle from resource production to the electricity production but without consideration of waste disposal or decommission after the life of the plant.

2.2. Social Life Cycle Assessment

Life cycle assessment (LCA) is a commonly-used analytical framework to quantify the impact of a product or a service over its lifetime. LCA has been primarily applied to the assessment of environmental emissions, most notably to compare the carbon dioxide emissions. Social life cycle assessment, or S-LCA, is an emerging framework to assess the social impacts of products or services through LCA. S-LCA has been developed in conformity to the international standard of LCA, ISO14040/44, and its first guideline was published by United Nations Environmental Programme and Society of Environmental Toxicology and Chemistry [6]. S-LCA looks into social impact on the workers in the product supply chain, for categories including health and safety, human rights, cultural heritage, working conditions, and governance. Results from S-LCA enable the identification of areas of improvement and comparison of products from the standpoint of their social performance, which could be valuable for both policymaking and corporate decision-making is to facilitate the enhancement of social conditions.

For its helpfulness in policymaking, S-LCA has been increasingly applied to various products and services in the last several years, including tourism, farming, and recycling systems [7–10]. However, few have applied S-LCA to assess the impact of the introduction of science and technology; in fact, none of the preceding studies have applied the framework to make a comparison of different electricity generation systems.

The authors believe S-LCA can be an ideal framework to quantitatively discuss the intersections between science and society that have policy implications. S-LCA can be especially beneficial in the energy sector, where the supply chains are cross-border and the interactions between human rights, standard of living and natural resources are complex. The authors hope that this analysis fills this gap.

2.3. Social Hotspots Database

Social Hotspots Database (SHDB) was used in this analysis as the database to calculate the social impact of the supply chain of renewable electricity. SHDB is a follow-up initiative to the UNEP/SETAC

Guidelines developed by New Earth, and is the first commercially available database for S-LCA to enable the identification of the social impact along the product supply chain [11,12]. SHDB is based on the global trade analysis project (GTAP) input-output model, and the database is composed of sector and country-specific tables of indicators for 57 sectors in 133 countries to support identifying hotspots in supply chains based on potential social impacts. SHDB enables the efficient application of S-LCA by providing data for: (1) labor intensity in worker hours per unit process; (2) risk for, or opportunity to affect relevant social themes or sub-categories related to human rights, labor rights, and decent work, governance and access to community services; and (3) gravity of a social issue [13]. With SHDB, social impacts of a product system can be measured in “Risk Hours (RH).” Risk Hours represents the weighted cumulative labor hours where workers in the supply chain may be at risk for each specific social issue. The authors used the 2013 version of SHDB, which is based on the GTAP model published in 2008 and social risk data from years 2010–2012 [14]. Considering major renewable energy development projects in Malaysia took place around 2011 [15], the 2013 SHDB would provide timely valid risk data for this analysis.

2.4. Impact Assessment Method

SHDB provides an impact assessment method for S-LCA named Social Hotspot Index (SHI). SHDB and SHI have 22 midpoint impact categories (called Social Themes) and five endpoint impact categories (called Social Category) as illustrated in Figure 1. For each theme and for many indicators, impact subcategories called Characterized Social Issues are available as summarized in Table 1.

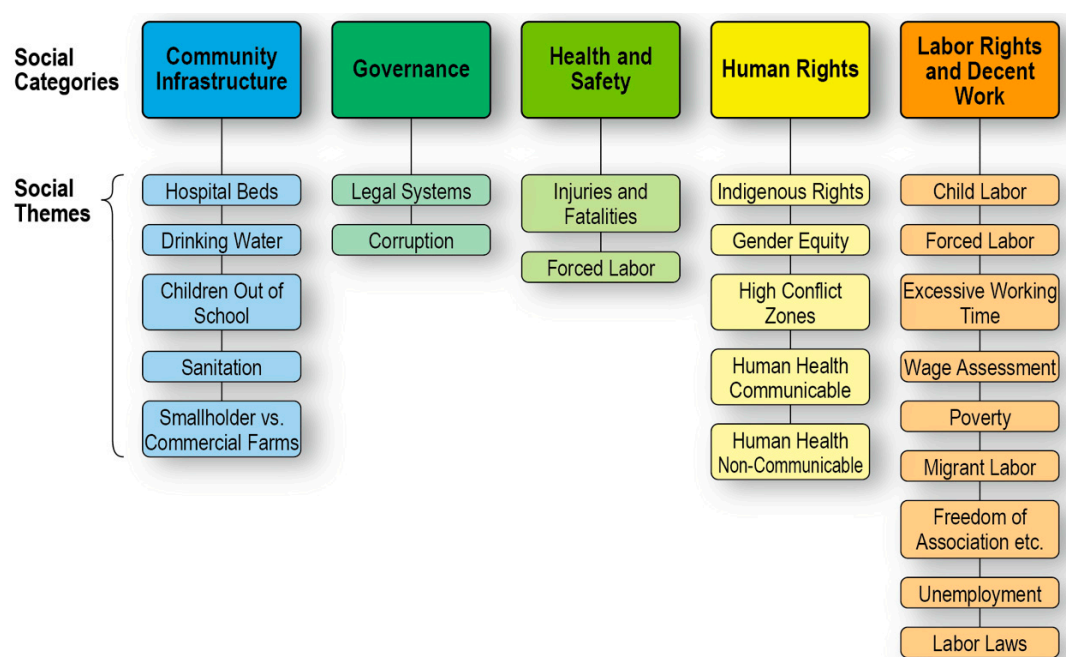


Figure 1. Social Hotspots Database midpoint and endpoint impact categories [16].

Table 1. Characterized Social Issues by Social Theme and Category [13].

Social Theme (Name of Table)	Data Indicator	Characterized Issue
Labor Laws/Conventions	Number of Labor Laws	Risk of Country not passing Labor Laws
	Number of Labor Laws by sector	Risk of Country not passing Labor Laws by Sector
	Number of Labor Conventions ratified (out of 81 possible)	Risk of Country not adopting Labor Conventions
	Number of Labor Conventions ratified by sector	Risk of Country not adopting Labor Conventions by Sector
	Year of last Minimum Wage Update	Risk of Minimum Wage not being updated

Table 1. Cont.

Social Theme (Name of Table)	Data Indicator	Characterized Issue
Wage Assessment	Minimum Wages (USD)	Risk of Country Average Wage being < Minimum Wage
	Average Unskilled Wages (USD) in country	
	Non-Poverty Guideline (USD)	Risk of Country Average Wage being < Non-Poverty Guideline
	Average Unskilled Wages (USD) in country	
	Minimum Wages (USD)	Risk of Sector Average Wage being < Minimum Wage
	Average Unskilled Wages (USD) by sector	
	Non-Poverty Guideline (USD)	Risk of Sector Average Wage being < Non-Poverty Guideline
	Average Unskilled Wages (USD) by sector	
Population living in Poverty	Percent of Population living on <\$2/day	Risk of Population living on <\$2/day
Child Labor	Child Labor % in country	Risk of Child Labor in country
	Child Labor % by sector	Risk of Child Labor by sector
Forced Labor	Qualitative	Risk of Forced Labor in country
	Qualitative	Risk of Forced Labor by sector
Excessive Working Time	Percent working >48 h/week in country	Risk of Population working >48 h/week in country
	Qualitative	Risk of Population working >48 h/week by Sector
Freedom of Association, Collective Bargaining, Right to Strike	Qualitative	Risk of not having Freedom of Association Rights
	Qualitative	Risk of not having Collective Bargaining Rights
	Qualitative	Risk of not having the Right to Strike
Unemployment	Unemployment Average % from 2000–2009	Risk of High Unemployment in country
	Unemployment % by sector	Risk for High Unemployment by sector
Legal System	World Bank Worldwide Governance Indicator—Rule of Law	Risk of Fragility in Legal System
	Bertelsmann Transformational Index - Rule of Law, Independent Judiciary	
	CIRI Human Rights Index—Independent Judiciary	
	Global Integrity Index—Judicial Accountability	
	Global Integrity Index—Rule of Law	
	Global Integrity Index—Law Enforcement	
	World Justice Project—Average	
Indigenous Rights	Presence of indigenous population, X	Not characterized
	Indigenous Population, %	Amount of Indigenous Population
	ILO Convention adopted for Indigenous Population, Y or N	Risk of country not adopting Indigenous ILO convention and UN Declaration
	UN Declaration for Indigenous Population, endorsed (Y), abstained (A), against (N)	
	Number of Laws enacted to protect Indigenous Population	Risk of country not passing Laws to protect Indigenous Population
	Qualitative	Risk for Indigenous Rights Infringements by Sector
Gender Equity	Social Institutions and Gender Index	Risk of Gender Inequity
	Global Gender Gap	
	World Bank Gender Development Indicator	
	World Bank Gender Empowerment Index	
	CIRI Human Rights Index—Economic	
	CIRI Human Rights Index—Political	
	CIRI Human Rights Index—Social	
	Adolescent fertility rate (births per 1000 women ages 15–19)	Not characterized
	Fertility rate, total (births per woman)	Not characterized

Table 1. Cont.

Social Theme (Name of Table)	Data Indicator	Characterized Issue
	Share of women employed in the nonagricultural sector (% of total nonagricultural employment)	Not characterized
	% Unemployment, (% of female labor force unemployed/% of male labor force unemployed)	Not characterized
	% of women workers vs. men by sector	Risk of Gender Inequity by sector
High Conflict Zones	Heidelberg Conflict Barometer—# of conflicts	Risk for High Conflict
	Heidelberg Conflict Barometer—maximum intensity of conflicts (1–5)	
	Heidelberg Conflict Barometer—change in conflicts (positive = worsening)	
	Number of Refugees—UN Refugee Agency	
	Center for Systemic Peace Indicator	Risk for High Conflict specific to sectors
	Minority Rights Group Indicator	
	Top Risers from last year in Minority Rights Group Indicator, X	
Human Health—Communicable Diseases and Other Health Risks besides Disease	Qualitative	Risk for High Conflict specific to sectors
	Life expectancy at birth (years) 2008	Risk of low life expectancy
	Mortality rates for injuries (per 100,000 population) 2004	Risk of high mortality rates due to injury
	Proportion of undernourished % of total population, (–) = <5% 2005–2007	Risk of high undernourishment
	Deaths due to indoor and outdoor air and water pollution (per one million population) 2004	Risk of death due to air and water pollution
	Population affected by natural disasters, average per year per million 2000–2009	Risk of death due to natural disasters
	Cases of HIV (per 1000 adults 15–49 years) 2010	Risk of HIV 2010
	Cases of Tuberculosis (per 100,000 population) 2008	Risk of Tuberculosis 2008
	Cases of Malaria (per 100,000 population) 2008	Risk of Malaria 2008
	Cases of Dengue Fever (per 100,000 population) 2005	Risk of Dengue Fever, 2005
	Cases of Cholera 2008	Risk of Cholera 2008
	Mortality rates from communicable diseases (per 100,000 population) 2004	Risk of mortality from communicable diseases
Children Out of School	Children out of School—male	Risk of Children not attending School—male
	Children out of School—female	Risk of Children not attending School—female
	Children out of School—total	Risk of Children not attending School—total
Access to Improved Drinking Water	Access to Improved Drinking Water, %—rural	Risk of not having access to Improved Drinking Water—rural
	Access to Improved Drinking Water, %—urban	Risk of not having access to Improved Drinking Water—urban
	Access to Improved Drinking Water, %—total	Risk of not having access to Improved Drinking Water—total
Access to Improved Sanitation	Access to Improved Sanitation, %—rural	Risk of not having access to Improved Sanitation—rural
	Access to Improved Sanitation, %—urban	Risk of not having access to Improved Sanitation—urban
	Access to Improved Sanitation, %—total	Risk of not having access to Improved Sanitation—total
Access to Hospital Beds	Access to Hospital Beds—# beds/1000 pop	Risk of not having Access to Hospital Beds

Based on the inventory data for 705 indicators, SHDB weights and calculates the SHI as Equation (1).

$$SHI = \frac{\sum_{T=1}^n R_{avg} W}{\sum_{T=1}^n R_{max} W} \quad (1)$$

where n represents the number of Social Themes, R_{avg} represents the average risk of the Social Theme, R_{max} represents the maximum risk for a theme and W represents the weighting factor [17]. The weighting factor is assigned based on the risk levels: 10 for very high, 5 for high, 1 for medium and 0.1 for low. For normalization of the results, the SHDB simply weights all Social Categories equally.

This social life cycle impact assessment methodology is illustrated by Shemfe et al. as Figure 2 [17]. The 22 midpoint impact categories of SHDB are part of the international auditable certification for the promotion of labor rights, SA8000 [18]. Although currently there are no ISO norms specifically for S-LCA, SHDB is based on the principles of LCA ISO norms (ISO 14040/14044) and is in conformity to the UNEP/SETAC Guidelines [19].

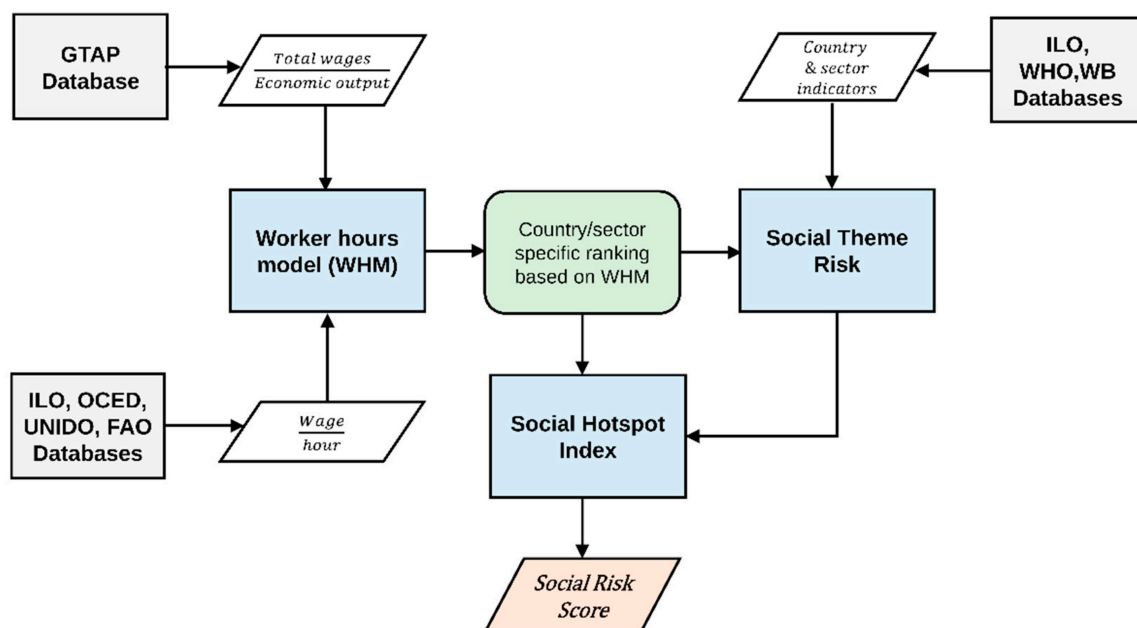


Figure 2. Methodology of the Social Life Cycle Impact Assessment with SHDB [17].

2.5. Data Collection

The authors collected the life-cycle cost structure data for renewable energy sources in Malaysia from a report published by ASEAN Centre for Energy in 2016 [20], which reports the levelized costs of electricity (LCOE) of renewable electricity based on surveys of 21 solar photovoltaic (PV), five biomass, and five hydro power plants in Malaysia. In this report, LCOE is defined as the net present value of the unit cost of electricity over a lifetime, calculated by dividing the net present value of all costs over the lifetime of the project by the total electricity output of the project [20]. Due to the difficulties in assuming different lifetimes for every project, a lifetime of 20 years was assumed for all plants. A discount rate of 10% is used for all projects, which applies to social impacts as well. Each cost breakdown is then assigned a GTAP model section code as well as a harmonized commodity description and coding systems 6-digit trade category code (HS-6) by the authors.

The countries-of-origins data for each capital expenditure (CAPEX) sector were collected from the economic atlas, which are derived from the countries reporting to the United Nations Statistical Division (COMTRADE), and raw trade data on services are from the International Monetary Fund (IMF) Direction of Trade Statistics database [21]. The authors referenced the import trade statistics of Malaysia in 2016 for each respective HS-6 code, with the cut-off importing a share value of 5%.

2.6. Product Modeling

Based on the data collection method detailed in the previous section, the levelized cost structure of each renewable electricity in Malaysia in 2016 is modeled as Figure 3. For solar PV, plants are classified based on their capacity as small (below 100 kW output), medium (above 100 kW but below 1000 kW), and large (above 1000 kW). For biomass and hydro power electricity, the cost modeling was based on the country average data. The costs reported by ASEAN Centre for Energy were converted into USD2002 by multiplying by 0.74957 to adjust for inflation. Because of the lack of necessary trade data, the mineral products from Brunei were calculated as the ones from Malaysia. Since the mineral products from Brunei only amounts to 0.2% of the LCOE of hydroelectricity, this change does not affect the results significantly. For the conventional electricity model in Malaysia, the default product model provided by SHDB for electricity in Malaysia was used without modifications. (The cost structures and the country-of-origins models are attached as Supplementary Materials as Tables S1 and S2.)

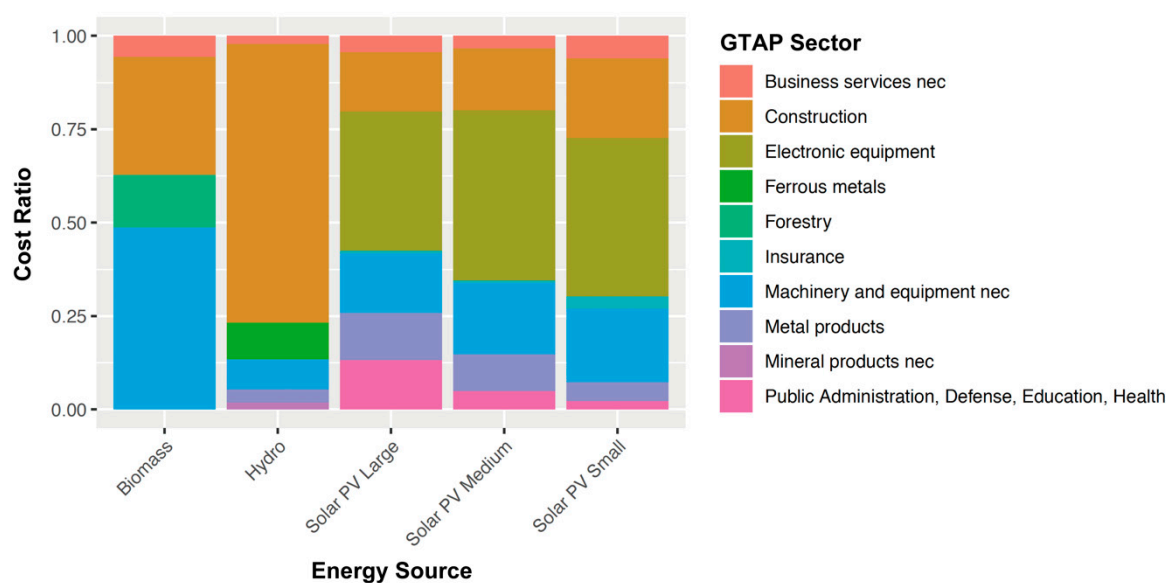


Figure 3. GTAP sector cost ratios of renewable electricity in Malaysia in 2016.

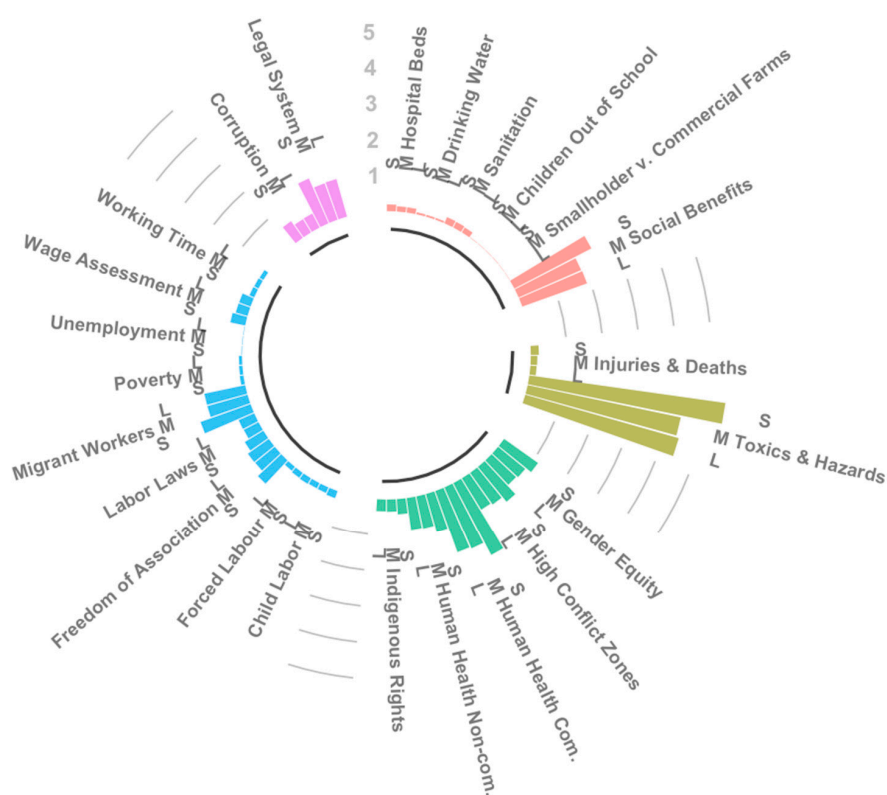
3. Results

3.1. Social Life Cycle Impact Assessment

The social impact of 1 kWh of electricity from five renewable sources (biomass, solar PV small, solar PV medium, solar PV large, and hydro) as well as the conventional production mix were assessed with SHDB on openLCA 1.7.2. The calculated Risk Hours inventories were then weighted and converged into 24 Social Themes and 5 Social Categories with the Social Hotspot Index as illustrated in Figures 4 and 5.

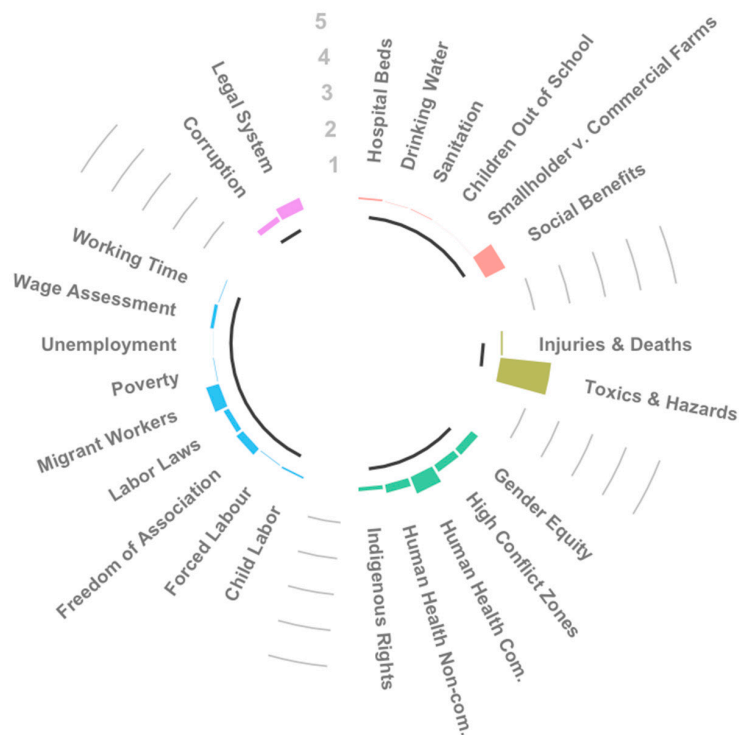


(a) Biomass



(b) Solar PV (S: below 100 kW, M: between 100-1,000 kW, L: above 1 MW)

Figure 4. Cont.



(c) Hydro

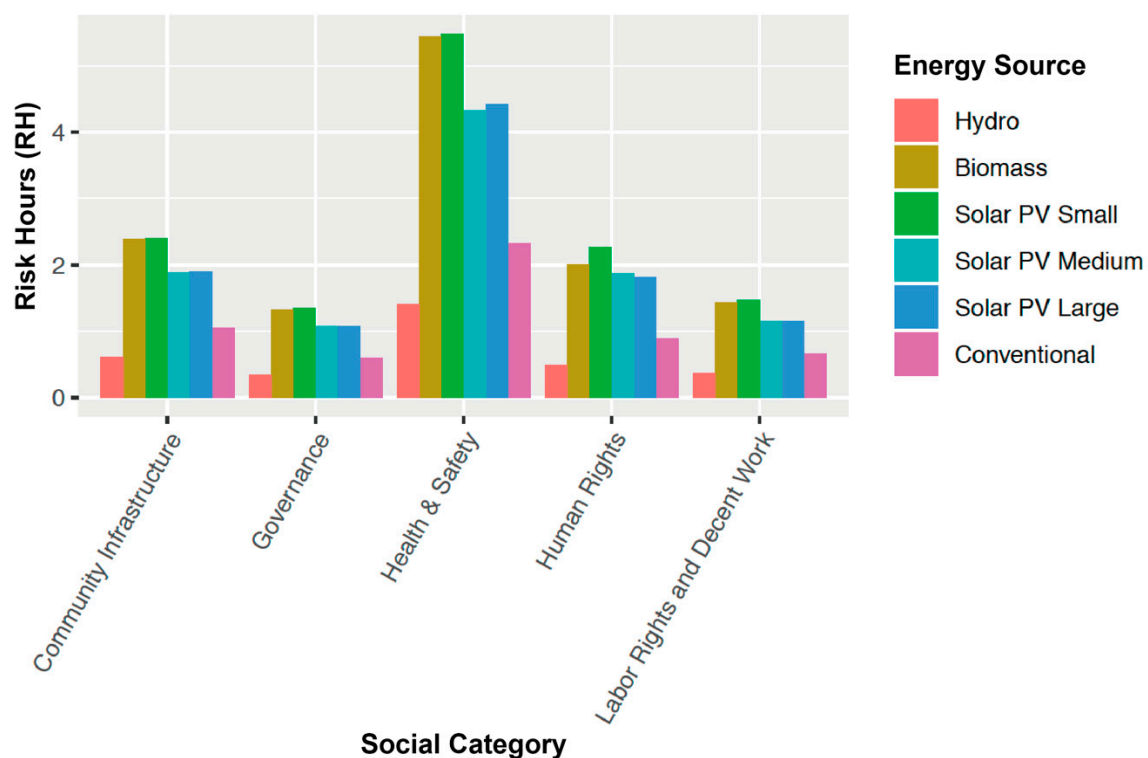
Figure 4. Midpoint life cycle impact assessment results: (a) Biomass; (b) Solar PV; (c) Hydro.**Figure 5.** Comparison of social impacts of various electricity production method by Social Categories.

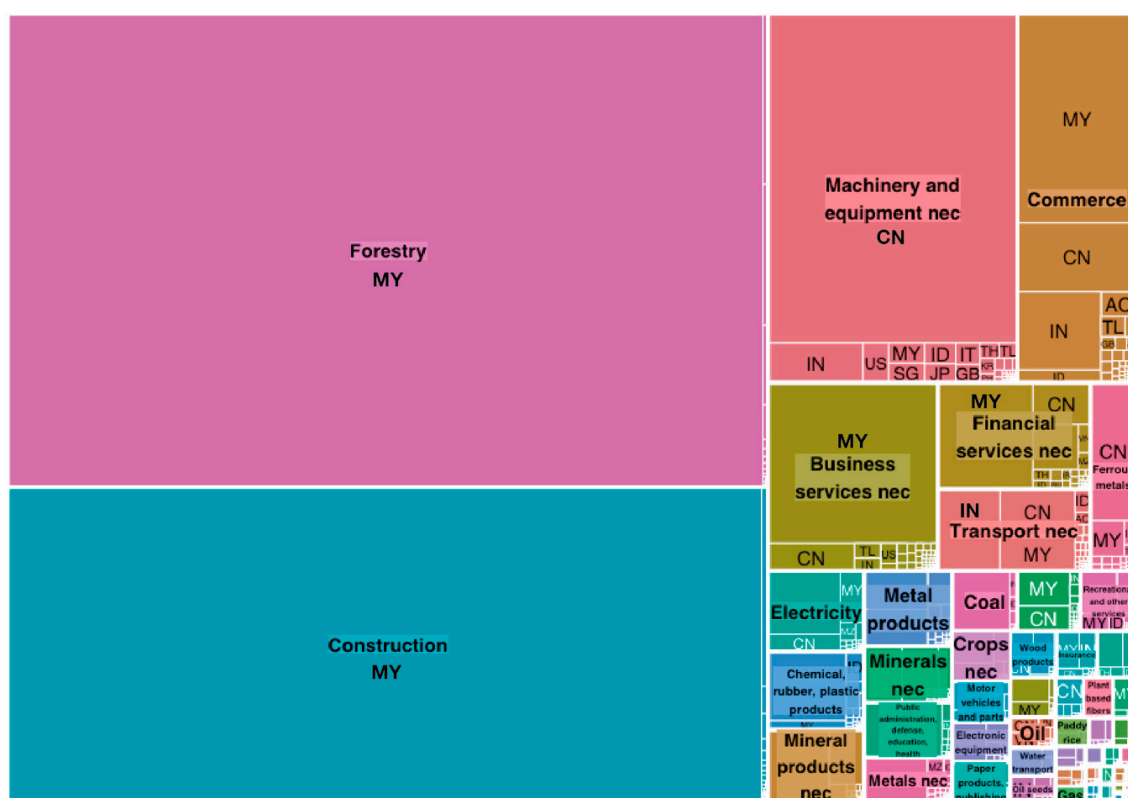
Figure 5 indicate that regardless of the energy sources, renewable electricity has similar social impact patterns: the health and safety category has the greatest adverse social impact in the supply

chain due to the great impact from toxics and hazards. A similar pattern is also observed for the conventional electricity mix.

3.2. Process and Country Contributions

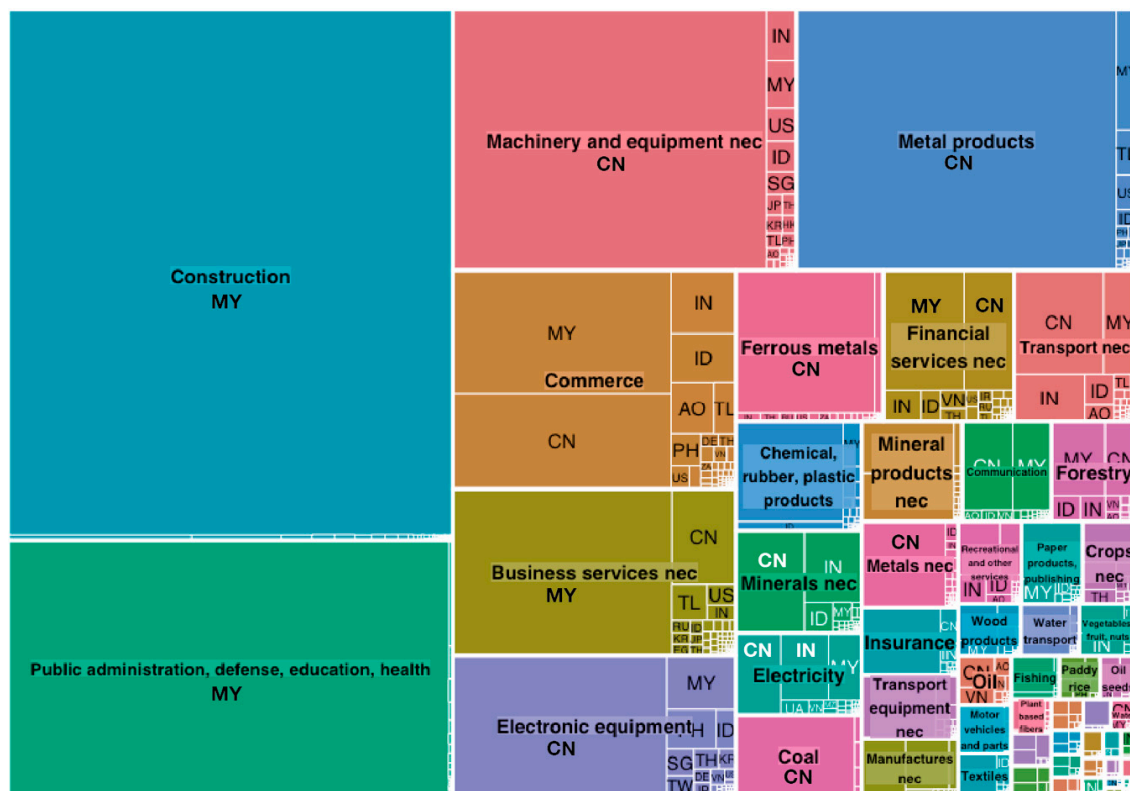
The calculated process and country contributions of biomass, solar PV and Hydro are illustrated in Figure 6a–c, respectively. Here, the areas for each process are drawn in proportion to the total Risk Hours of each sector/country.

Figure 6 shows that the construction in Malaysia is the largest social footprint contributor for both hydro and solar PV electricity. In particular, construction is responsible for more than 60% of the social footprint of hydroelectricity. For solar PV, electronic equipment and metals from China are the second largest contributor when combined, which reflects the fact that a large share of the solar panels, and its mounting parts come from China. Another large contributor to solar PV was the public administration in Malaysia, which reflects the licensing cost for a significant portion of the capital expenditures for solar PV plants in Malaysia. On the contrary, the largest social footprint contributor of biomass electricity came from the forestry in Malaysia, followed by the construction in Malaysia.

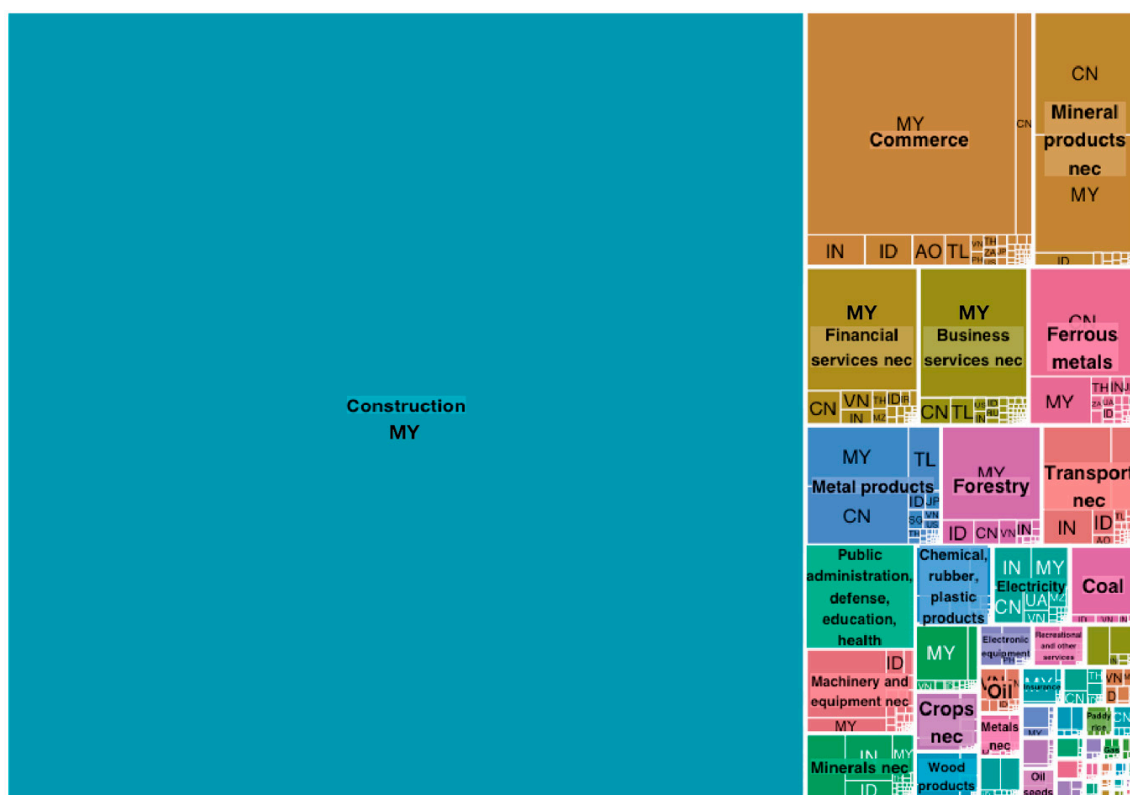


(a) Biomass

Figure 6. Cont.



(b) Solar PV (Large)



(c) Hydro

Figure 6. Process and country contributions: (a) Biomass; (b) Solar PV; (c) Hydro. (MY: Malaysia, CN: China, IN: India, ID: Indonesia).

4. Discussion

4.1. Normalized Social Footprint per Unit-Electricity vs. Unit-Cost

Figure 7 shows the calculated normalized social footprint for each electricity source, (a) per 1 kWh and (b) per 1 USD of generation cost. According to Figure 7a, the electricity from renewables had a greater adverse social impact per unit-electricity than the conventional electricity mix per kWh, except for hydroelectricity. Biomass (19.5 RH) exhibited 227% of the social impact of conventional electricity (8.6 RH). Solar PVs exhibited differing degrees of social impact: small-scale solar PV (defined as < 100 kW output) had the greatest social impact among the five assessed renewable energy sources with 20.7 RH; mid-scale and large-scale solar PVs (defined as 100–1000 kW and > 1 MW, respectively) had a similar social impact with 16.7 and 16.6 RH, respectively, which are 14% less than biomass but 95% more than the conventional electricity mix per 1 kWh. Hydro, on the other hand, exhibited a significantly reduced social impact with 4.9 RH, which is 43% smaller than the conventional electricity mix.

This result may be mainly attributed to the high cost of renewable electricity. According to ASEAN Centre for Energy (2016), the costs of electricity for each energy source in Malaysia are as follows:

solar PV small (\$0.20/kWh) > solar PV medium (\$0.17/kWh) > solar PV large (\$0.15/kWh)
>> biomass (\$0.10/kWh) >> conventional (\$0.03/kWh) ~ hydro (\$0.03/kWh)

As such, comparing the social impact per unit generation cost, rather than per unit electricity generated, reveals that the solar power plant is much less impactful than all other energy sources. These comparisons are illustrated in Figure 7b, where solar PV has a 60% smaller social impact than conventional electricity per dollar spent, as follows:

conventional (281.5 RH) >> biomass (192.6 RH) > hydro (168.6 RH) >> solar PV large (110.5 RH) > solar PV small (103.5 RH) > solar PV medium (98.2 RH)

This comparison of social footprint per unit electricity vs. unit cost indicates that while the electricity from solar PV and biomass in Malaysia have a greater adverse social impact than the conventional energy mix per unit electricity at present, labor conditions for these renewable electricity sources per unit cost are significantly better than those of conventional electricity generation. This suggests that when the generation costs of these renewable sources eventually drop and reach grid parity, the social impact of electricity generation will be mitigated through the development of these sources.

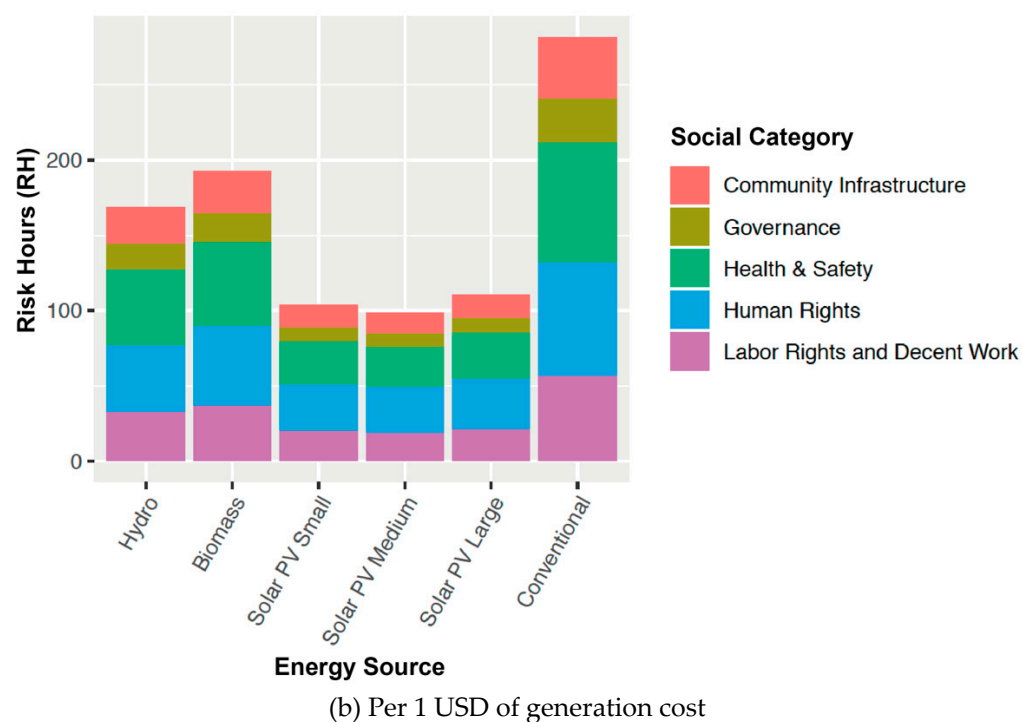
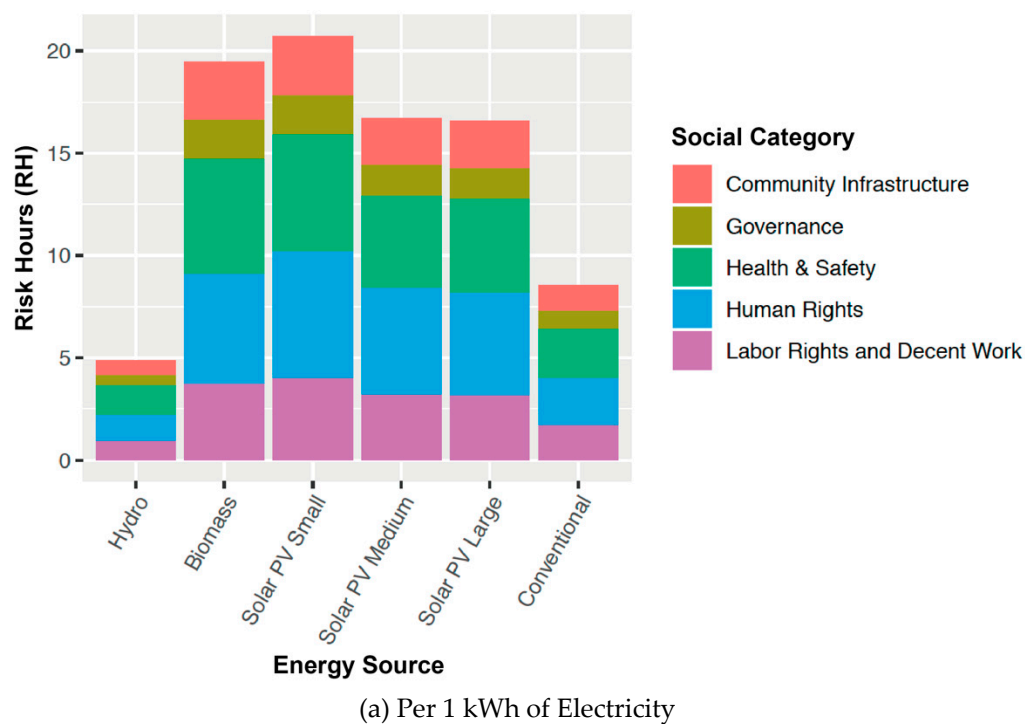


Figure 7. Normalized social footprints: (a) per unit electricity, (b) per unit cost.

4.2. Geographical Social Hotspots of Renewable Electricity

Figure 6 shows that a great proportion of the social footprint of renewable electricity came from outside of the country as a result of the global supply chain. Figure 8 illustrates the social footprint contribution proportions of China, the largest overseas contributor, and of Malaysia.

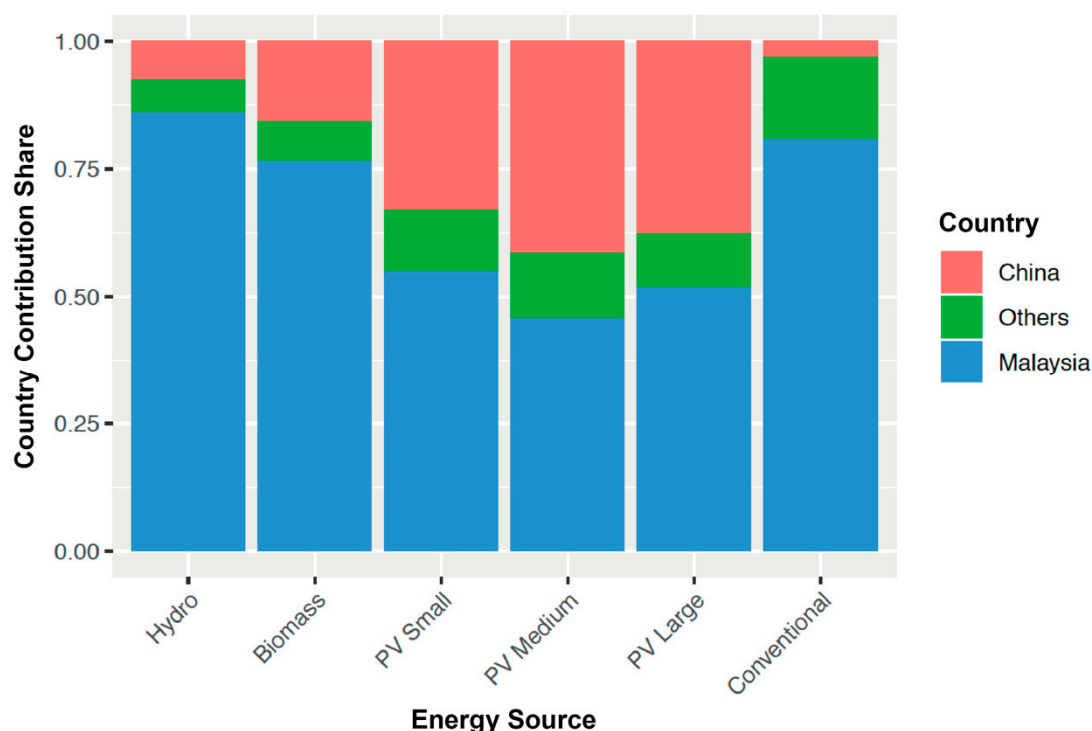


Figure 8. Geographical Social Hotspots of Renewable Electricity.

The results show that renewable electricity has a less adverse social impact domestically, while the effects are transferred to the exporting countries, in this case China. This indicates the possibility to improve the social impact of renewable electricity across the globe by improving labor conditions in a few exporting countries.

5. Conclusions

What do all these data suggest? Based on Figure 7a, it can be concluded that electricity generation through renewable energy technologies causes significantly greater stress among workers, with the exception of hydroelectricity. However, on the other hand, when the social impacts per generation cost were compared, renewables had far lower impacts than the conventional electricity mix as illustrated in Figure 7b. This is a fascinating result, because it suggests that while renewables are not as friendly to humans as to the environment at the moment, they will have the potential to be much more favorable to humans than the conventional energy mix in the future when costs of renewable electricity eventually drop. In fact, it is estimated that when solar PV achieves grid parity in the future, labor conditions through the electricity supply chain will be as much as 60% less than the conventional electricity generation in Malaysia. It was also discovered that these adverse social effects are transferred to few exporting countries, including China, as illustrated in Figure 8. This could provide an opportunity to reduce the social impacts of renewable electricity across the nation, by improving labor conditions in a few exporting countries.

Are renewable energy technologies as friendly to humans as to the environment? Based on the findings, the authors conclude that the answer is, “*not-yet, but eventually.*” This analysis suggests a clear path toward the reduction of the adverse impacts of renewables: to continue the efforts to reduce the cost of renewable energy technologies, while improving the labor conditions in key exporting countries like China.

Supplementary Materials: The following are available online at <http://www.mdpi.com/2071-1050/11/5/1370/s1>, Table S1: Cost Structures and the Country of Origins Model: (a) Biomass; (b) Solar PV; (c) Hydro, Table S2: SHDB product models: (a) Biomass; (b) Solar PV; (c) Hydro.

Author Contributions: Conceptualization, S.T.; methodology, S.T., A.R.K. and C.B.N.; software, C.B.N.; validation, S.T., S.S., A.R.K., S.M. and C.B.N.; formal analysis, S.T.; investigation, S.T. and A.R.K.; resources, S.T. and C.B.N.; data curation, S.T. and C.B.N.; writing—original draft preparation, S.T.; writing—review and editing, S.T. and A.R.K.; visualization, S.T.; supervision, S.S., S.M. and C.B.N.; project administration, S.T. and A.R.K.

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