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# A review and inventory of U.S. hydrogen emissions for production, distribution and storage

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In response to the growing global interest in hydrogen as an energy carrier, this study provides the first attempt to develop a baseline inventory of U.S. hydrogen emissions from production, distribution and storage. The scope of this study was limited to pure hydrogen emissions and excludes emissions from low purity hydrogen streams and carriers. A detailed literature search was conducted, utilizing various greenhouse gas emissions inventory protocol principles and guidelines, to consolidate a list of activity data and emission factors. The best available activity data and emission factors were then selected, via a Multi-Criteria-Based Decision Making Method named Technique for Order Preference by Similarity to Ideal Solution, or modelled using best-engineering estimates. The study estimated total U.S. hydrogen emissions of 0.063 MMTA with emission bounds ranging from 0.02 to 0.11 MMTA. Given the total estimated H<sub>2</sub> production capacity of 7.97 MMTA, the study estimates a total U.S. hydrogen emission rate for production, distribution and storage of 0.79% (0.26%-1.32%). To reduce the uncertainty in the estimated total hydrogen emissions, future work should be conducted to measure facility-level hydrogen emission factors across multiple sectors. The inventory framework developed in this study can serve as a living document that can be updated and enhanced as more empirical data is obtained. This study also provides detailed insights regarding key emission or leakage sources and causes from each supply chain stage. The insights and conclusions from this study can provide direction for hydrogen production companies and safety professionals as they develop hydrogen emission mitigation measures, and controls.

# KEYWORDS

hydrogen emissions, hydrogen emissions inventory, hydrogen supply chain, hydrogen production, hydrogen economy

# 1 Introduction

Hydrogen ( $H_2$ ) is gaining interest as a promising energy carrier in current and future efforts to decarbonize the global energy portfolio and meet greenhouse gas emissions targets (International Energy Agency, 2021).  $H_2$  is considered to be a clean alternative fuel since it contains no carbon atoms and does not absorb infrared

radiation in the atmosphere (Derwent, 2018; Roueff et al., 2019; Hauglustaine et al., 2022). As a result, there is a growing global and national interest in the utilization of  $\rm H_2$  to store electricity produced by intermittent renewable energy sources, and as a fuel for electricity production, transportation, and building heating (Field and Derwent, 2021; International Energy Agency, 2021; National Renewable Energy Laboratory, 2022).

The International Energy Agency (2021) reported that, since 2019, over 16 countries have published national hydrogen strategies aimed at defining the role of H2 within their future energy economy. In 2020, the U.S. Department of Energy's Office (DOE) of Fossil Energy's released a strategic plan to accelerate research, development, and deployment of H<sub>2</sub> technologies in the U.S. This plan involves significant research initiatives across all segments of the H<sub>2</sub> supply chain (US DOE, 2020). In 2023, the U.S. also announced plans to fund seven regional clean H2 hubs across the country, capable of producing over three million metric tons of clean H2 annually, aiming to significantly decarbonize various end-use sectors, and create numerous job opportunities (The White House, 2023). These projects aim to meaningfully contribute to the United Nations' Sustainable Development Goal 7 "to ensure access to affordable, reliable, sustainable and modern energy for all" (United Nations, 2025). Consequently, H2 is projected to play key role in supporting a future clean energy economy, however an obstacle to its future widespread use revolves around its potential negative impact as an indirect greenhouse gas (Ocko and Hamburg, 2022; Esquivel-Elizondo et al., 2023).

 $\rm H_2$  acts as an indirect greenhouse gas through its involvement in chemical reactions that increase the lifespan or concentration of other potent greenhouse gases, including methane (Derwent et al., 2006; Derwent, 2018). The oxidation of  $\rm H_2$  in the troposphere results in the depletion of hydroxyl (OH) radicals that act as a sink for methane (Derwent, 2018; Paulot et al., 2021). This consequently increases the atmospheric lifespan of methane.  $\rm H_2$  oxidation reactions also ultimately lead to the formation and increase of other greenhouse gases, specifically tropospheric ozone, and stratospheric water vapor (Derwent, 2018; Derwent et al., 2020; Field and Derwent, 2021; Paulot et al., 2021). The indirect greenhouse gas effect of  $\rm H_2$  has led to the emergence of research efforts to quantify current and potential  $\rm H_2$  emissions across the supply chain.

The Inventory of U.S. Greenhouse Gas Emissions and Sinks 1990-2022 (EPA, 2024), published by the United States Environmental Protection Agency, has served as the foundational document detailing emissions inventories for a comprehensive list of greenhouse gases nationally. Since its initial release in the early 1990s, multiple studies have been conducted to update its findings with more precise empirical data and results (Lamb et al., 2015; Zimmerle et al., 2015; Zimmerle et al., 2019). EPA (2024) now operates as a living document updated annually, regularly incorporating new research findings. There is growing interest in emulating this inventory approach for H<sub>2</sub> emissions however due to limitations in sensor technology H2 emissions from infrastructure across the supply chain (i.e., production, distribution and storage, end-use) is difficult to empirically quantify. Current commercially available sensor technologies are suitable for measuring higher H<sub>2</sub> concentrations to ensure safe operating conditions but lack the sensitivity to quantify overall site-level emissions (Najjar, 2019; Ocko and Hamburg, 2022; Esquivel-Elizondo et al., 2023). While research into sensors capable of detecting atmosphericconcentrations of H2 has increased significantly in recent years, the focus remains primarily in safety and process monitoring applications, making many unsuitable for site-level emission rate quantification (Chauhan and Bhattacharya, 2019; Ramaiyan et al., 2023; Liu et al., 2024; Swager et al., 2024; Benitto et al., 2025). To date only one study has empirically measured H2 emissions at an industrial park in the European union with a relatively high degree of uncertainty (Westra et al., 2024), highlighting the need for broader measurement efforts. Larger scope H2 measurement campaign efforts will likely be underway soon with the recent announcement of a new sensor capable of filling this research gap (Momeni et al., 2025). However, in the current absence of empirical measurement data, emerging H2 emissions research has mainly focused on identifying emission sources, and developing emission factor estimates via expert opinions, calculations via proxies, laboratory experiments, and modelling (Cooper et al., 2022; Fan et al., 2022; Frazer-Nash Consultancy, 2022; Esquivel-Elizondo et al., 2023).

Emerging H2 emissions studies have either taken a global approach toward H<sub>2</sub> emissions research (Cooper et al., 2022; Fan et al., 2022; Esquivel-Elizondo et al., 2023) or have focused on specific areas in Europe (Frazer-Nash Consultancy, 2022). The primary focus of these studies has been the production, distribution and storage segment with end-use being acknowledged as the least understood segment with little to no data available regarding activity data or emission factors (Cooper et al., 2022; Fan et al., 2022). Cooper et al. (2022) defined specific global H<sub>2</sub> supply chain cases and estimated or modelled emission factors mainly utilizing calculations via proxies and modelling. Although, Cooper et al. (2022) proposed emission factors for specific sources within the production, distribution and storage segment, their estimation efforts did not extend to some H2 emission sources within these segments that are prominent within the U.S. supply chain (e.g., Underground Hydrogen Storage [UHS], Fuel Cell Electric Vehicles [FCEV] Stations). Frazer-Nash Consultancy (2022) focused on the United Kingdom supply chain and proposed a wide range of emission factor estimates for various sources. Esquivel-Elizondo et al. (2023) and Fan et al. (2022) build on this foundation and provide an overview of the available knowledge on emission factor estimates across global value chains. Arrigoni and Bravo Diaz (2022) and Pranda and de Reals (2023) provided emission factor estimates across the U.S. supply chain based on expert opinion. These studies have provided a foundation for H<sub>2</sub> emissions research; however, they primarily focus on developing and presenting emission factor estimates and do not collate or apply the relevant activity data needed to estimate emissions. Thus, no study to date has attempted to present an initial comprehensive estimate of  $H_2$  emissions across the U.S. supply chain.

One key distinction between the current research on  $\rm H_2$  emissions and the previous work on a comparable gas within EPA (2024), for instance methane, is the lack of a detailed inventory of emission sources (i.e., activity data) and emission factors. A detailed methane inventory existed as part of earlier versions of EPA (2024) before the methane emissions measurement campaigns (Marchese et al., 2015; Mitchell et al., 2015). While the methane emissions contained in earlier versions of EPA (2024) were incomplete and ultimately underreported, the inventory still represented a powerful tool to

#### 1. Definition

#### Goal:

Estimate annual hydrogen emissions from the U.S. production, distribution, and storage sector

# Supply Chain Stages:

- ProductionMerchant and Byproduct hydrogen
- Captive hydrogen
- Green Hydrogen

#### Distribution and Storage

- Liquefaction
- C-H2 Transmission Pipelines
- · Underground Hydrogen Storage
- FCEV Refueling Stations

#### 2. Data Collection

#### Data Sources:

- Journal Articles
- Conference Articles
   Government Reports
- Company Reports
- EPA Greenhouse Gas
   Reporting Program
- Media Releases

#### 3. Inventory Activity Data (AD)

# Emission Factors (EF) 4. Select Best Available Data

- Production EF selection
   TOPSIS
- Distribution and Storage
   Modelling

#### 5. Calculate Emissions

#### **Emissions:**

Multiplying activity data by an emission factor

#### **Activity Data:**

A quantitative measure of the level of activity that results in hydrogen emissions or losses taking place during a given period of time (e.g., production capacity, length of pipeline etc.)

#### Emission Factor:

A measure of the mass of hydrogen emission or loss relative to a unit with

FIGURE 1
Stepwise methodology of the inventory process for this study.

inform the design, implementation, and dissemination of the methane emissions measurement campaigns.

Since no comparable H<sub>2</sub> emissions inventory currently exists, the goal of this study is to review the hydrogen emissions literature in order to develop a baseline H<sub>2</sub> emissions inventory for the U.S. H2 supply chain infrastructure. Majority of the H2 emissions literature has focused on the production through to the distribution and storage segment and highlights the lack of current knowledge regarding activity data, emission sources, and emission factors associated with H<sub>2</sub> end-use (Arrigoni and Bravo Diaz, 2022; Cooper et al., 2022; Fan et al., 2022). Thus, to provide an initial baseline H<sub>2</sub> emissions inventory this study limits its scope to the production, distribution and storage segment. A detailed literature review and expert interviews were conducted to compile the best available emission factors, and activity data for a wide range of U.S. H<sub>2</sub> supply chain emission sources within these segments. Then emissions estimates were calculated for those sources. In cases where information was lacking, best-engineering estimates for emission factors and activity data were developed or modelled. This initial inventory framework can inform the design and implementation of future H<sub>2</sub> measurement campaigns and provides a framework that can be updated and enhanced as more precise empirical data is obtained. This study is in alignment with the United Nations' Sustainable Development Goal 7 (United Nations, 2025).

# 2 Methodology

The primary goal of this project was to develop a baseline  $H_2$  emissions inventory based on current U.S.  $H_2$  supply chain infrastructure for the production, distribution and storage segments. Figure 1 presents the stepwise methodology involved in developing the inventory for this study. The steps in Figure 1 are based on various greenhouse gas emissions inventory protocol principles and guidelines (EPA, 2004; GPC, 2014; IPCC, 2024). The scope of this study is defined (Section 2.1). The data collection and inventory process is explained (Section 2.2). An overview of the method for selecting or estimating emission factors is provided (Section 2.3, 2.4). In Section 3,

the results of each supply chain stage is provided and more detail regarding emission factor selection and estimation process, including key in emission sources are provided. The overall results of the study (activity data, emission factors, and emissions estimate) is provided in Table 11. Finally, Section 4 details discussions, overall contribution, limitations and future work of this study.

# 2.1 Scope of the study

Hydrogen is an important chemical commodity in various industries including oil refining, methanol production, and ammonia production (Joseck et al., 2016). The  $\rm H_2$  supply chain can be divided into three basic segments: Production, distribution and storage, end-use. The scope of this study was limited to pure  $\rm H_2$  emissions and excludes emissions from low purity  $\rm H_2$  streams (natural gas or C- $\rm H_2$  blends and low purity  $\rm H_2$  byproducts) and  $\rm H_2$  storage or carrier materials (ammonia and alkali metal hydrides). This study also primarily focused on emissions from the production, distribution and storage due to the lack of  $\rm H_2$  end-use data availability in the U.S. Table 1 shows the supply chain stages and segments included within the scope of this study.

The *Production* segment deals with H<sub>2</sub> production which occurs primarily through steam methane reforming (SMR). SMR production reportedly represents about 95% of H<sub>2</sub> production in the U.S (National Renewable Energy Laboratory, 2022). The SMR process involves methane reacting with high-temperature steam in the presence of a catalyst to produce hydrogen, carbon monoxide and carbon dioxide. Then carbon monoxide is reacted with steam in the presence of a catalyst (water-gas shift reaction) to form carbon dioxide and hydrogen. In some production facilities carbon capture, utilization and storage (CCUS) is utilized to reduce carbon dioxide emissions (National Renewable Energy Laboratory, 2022). When CCUS is not used, the H<sub>2</sub> produced from natural gas or methane is termed "Gray" Hydrogen. When CCUS is used, the H<sub>2</sub> produced is termed "Blue" Hydrogen.

SMR production can also be classified as Merchant or Captive Hydrogen based on end-use. "Merchant" Hydrogen refers to  $\rm H_2$ 

TABLE 1 Supply chain segments and stages within the scope of this study.

Segments	Stages
	Merchant and Byproduct H <sub>2</sub>
Production	Captive H <sub>2</sub>
	Green H <sub>2</sub>
	Liquefaction
Distribution and Storage	$C-H_2$ Transmission Pipelines Underground $H_2$ Storage
	FCEV Refueling Stations

TABLE 2 Key information included in inventory codebook in supplementary spreadsheet.

Code	Description
Literature Reference	The literature source where the information was retrieved
Source Category	The source, technology, or method of $\rm H_2$ production
Activity	The type of activity data retrieved
Published Activity Data	The published activity data value retrieved
Emission Factor	The emission factor proposed by the literature reference
H <sub>2</sub> Color	If available, the color category based on mode of production associated with the retrieved data
Geographical Information	If available, the region and/or geographic location of the site
Used in Estimation	Whether or not the data was used in the final emissions estimation of this study

produced on-site then sold to customers via pipeline, tank, or truck delivery. "Captive" Hydrogen refers to H2 produced by the consumer for internal use. As a result, merchant hydrogen is expected to have longer average transporting distances, while the transport segment for captive hydrogen is negligible in comparison. "By-product" Hydrogen refers to H2 that is produced from byproduct process streams and can either be sold to customers as Merchant H<sub>2</sub>, or used internally as Captive H<sub>2</sub> (Connelly et al., 2019). Since Captive hydrogen is necessarily produced for internal use, it is included in the scope of this study as a separate supply chain stage from Merchant and Byproduct hydrogen (Table 1). Green Hydrogen typically refers to hydrogen produced via electrolysis powered by renewable energy sources such as wind and solar energy. This method of producing hydrogen is still emerging and represents the vast minority of hydrogen production worldwide (National Renewable Energy Laboratory, 2022).

The Distribution and Storage segment involves the modes through with  $H_2$  is transported from production facilities to

customers.  $H_2$  is either transported in gaseous form as C-H<sub>2</sub> or liquified and transported as L-H<sub>2</sub>. Current modes of C-H<sub>2</sub> transportation in the U.S. include through transmission pipelines, and tanker trucks or trailers (Arrigoni and Bravo Diaz, 2022). UHS facilities connected to the network of  $H_2$  transmission pipelines are used to store the produced  $H_2$  and supplement delivery to customers as demand fluctuates (Zhu et al., 2023). Some of the produced  $H_2$  is also liquefied and stored or transported, via tanker trucks, as L-H<sub>2</sub> to various end-uses (Arrigoni and Bravo Diaz, 2022). Hydrogen is also stored on-site and dispensed through FCEV refueling stations (US DOE, 2023).

# 2.2 Data collection and inventory process

A detailed literature search and review process was conducted to consolidate a list of emission factors and activity data associated with the H<sub>2</sub> supply chain stages within the scope of this study. This literature search and review process took place from February 2023 until October 2024. During the literature search and review process, published peer reviewed studies, conference articles, technical reports from governmental and non-governmental agencies, the EPA Greenhouse Gas Reporting Program (EPA GHGRP, 2023), and media or press releases were reviewed in detail. Data was also acquired from independent energy research and business intelligence groups and from expert interviews with H<sub>2</sub> researchers and professionals. The overall goal of this process was to understand the H<sub>2</sub> supply chain, and compile information regarding available emission factors and activity data. Each literature or data source was thoroughly screened, and relevant information was extracted and manually coded in a Microsoft Excel spreadsheet Inventory codebook (Supplementary Spreadsheet). Table 2 provides brief descriptions of key categories of information, that were retrieved from the literature, and stored in the inventory codebook. Within the Inventory Codebook (see Supplementary Spreadsheet) all the relevant information found pertaining to activity data was stored in Supplementary Table SA1, while all the relevant data pertaining to emission factors was stored in the Supplementary Table SA2. These tabs represent a compendium of U.S. H2 activity data and emission factors. This data collection and inventory process resulted in 813 activity data entries and 27 emission factor entries.

# 2.3 Production emission factor selection approach

For the supply chain stages in the Production segment multiple sources were found presenting differing emission factors, thus for this segment, a Multi-Criteria-Based Decision Making Method (MCDM) method named Technique for Order Preference by Similarity to Ideal Solution (TOPSIS) was utilized as the main tool for selecting the best available emission factors.

MCDM methods allow the unbiased selection of an alternative when conflicting options are presented. TOPSIS, is one of the most powerful decision-making techniques that utilizes logical data about different alternatives to select the best available option (Jozi and Majd, 2014). It was developed by Hwang and Yoon (1981) and has been used in different areas including technology

TABLE 3 TOPSIS emission factor selection parameters.

Attribute	Weight ( $W_j$ )	Criteria	Score (x <sub>ij</sub> )
	25%		
		Published within the last 5 years (2019–2024)	3
Year of Study		Published within the last 10 years (≥ 2014)	2
		Published within the last 20 years (≥ 2004)	1
		Other	0
	40%		
		Large-scale direct measurement	3
Management America de		Small-scale lab experiment	2
Measurement Approach		Estimate, calculation, simulation, or model	1
		None of the above	0
	35%		
		Peer-reviewed research literature	3
Source Reliability		Gray literature (e.g., technical reports)	2
		Expert opinion	1
		Other (e.g., media or press release)	0

selection (Halicka, 2020), material selection (Kumar et al., 2014), investment selections (Zeng and Xiao, 2016), among other areas (Shyjith et al., 2008; Amudha et al., 2021). The steps involved in utilizing the TOPSIS method in this study are detailed below.

TOPSIS ranks the closeness or distance of various alternatives from an ideal solution based on the relative importance of key attributes and criteria (Kumar et al., 2014). The first step in utilizing the TOPSIS methodology was identifying the attributes and criteria that were most important for emission factor selection. Table 3 shows the TOPSIS emission factor selection parameters developed by the authors. First important attributes and criteria for emission factor selection were chosen based on recommendations from research literature (Cooper et al., 2022; EPA, 2024) as year of study (or recency), measurement approach, and source reliability. Weights  $(W_i)$  were then assigned to each attribute based on their relative order of importance for emission factor selection, thus, measurement approach ( $W_i = 40\%$ ) was assigned the largest weight followed by source reliability ( $W_i = 35\%$ ) and year of study ( $W_i =$ 25%). A score  $(x_{ij})$  between zero and three was selected as the mode of grading emission factor alternatives based on their attributes and criteria (see Table 3).

When differing emission factor alternatives were obtained from different sources, the TOPSIS method consists of first

normalizing the different alternatives in accordance with Equation 1:

$$A_{ij} = \frac{x_{ij}}{\sqrt{\sum_{j=1}^{n} x_{ij}^2}} \tag{1}$$

where  $A_{ij}$  is the normalized attributed value, and  $X_{ij}$  is the grade of each alternative under each attribute. Following, a weight (denoted as  $W_j$ ) will be assigned to each attribute, depending on its importance and it will be multiplied to each normalized attributed value (Equation 2):

$$V_{ij} = A_{ij}W_j \tag{2}$$

Subsequently, the ideal best  $(S_i^+)$  and ideal worst  $(S_i^-)$  value would be identified to then calculate the distance from the  $S_i^+$  and  $S_i^-$ . Finally, the performance each alternative is calculated following Equation 5:

$$S_i^+ = \left[\sum_{j=1}^m \left(V_{ij} - V_j^+\right)^2\right]^{0.5} \tag{3}$$

$$S_{i}^{-} = \left[ \sum_{j=1}^{m} \left( V_{ij} - V_{j}^{-} \right)^{2} \right]^{0.5} \tag{4}$$

$$P_i = \frac{S_j^-}{S_j^+ + S_j^-} \tag{5}$$

# 2.4 Distribution and storage emission factor estimation approach

A literature review was conducted to consolidate a list of emission factors for the H2 distribution and storage segment (Supplementary Table SA2). When no emission factors were found in the literature, our approach was to model H2 emissions using estimate emission factors. Specific details regarding the emission factor estimation processes for each supply chain stage is presented in Section 3.2. For the liquefaction stage, an emission factor estimate was retrieved from previous L- H<sub>2</sub> literature reviews (Section 3.2.1). For C-H<sub>2</sub> transmission pipelines, comparable natural gas systems and emission factors and the differences in physical properties between methane and hydrogen was used to estimate relative loss and leak rates (Section 3.2.2). For UHS, an emission factor was modelled based on UHS operating characteristics and features (Section 3.2.3). And finally for FCEV Refueling Stations, an emission factor estimate was retrieved from an FCEV Refueling Station emissions study (Section 3.2.4).

# 2.5 Calculating emissions

Emissions were calculated for each supply chain stage in this study using Equations 6, 7 below:

$$E_i = A_i E F_i \tag{6}$$

where  $E_i$  is the emissions (MTA) per supply chain stage i,  $A_i$  is the relevant  $H_2$  activity data in stage i, and  $EF_i$  is the best available emission factor for stage i. From this, the total amount of H<sub>2</sub> emitted (*E*) across a supply chain with *N* stages can be calculated.

$$E = \sum_{i=1}^{N} E_i \tag{7}$$

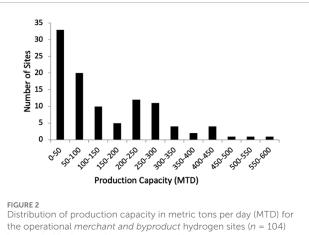
The overall emission rate can be calculated by dividing emissions E by the  $H_2$  production output of the supply chain. It is important to note that for the production and processing segment, the total production capacity was used as the activity data for each supply chain stage in this segment. For Green hydrogen, installed electrolyzer capacity was used to estimate production capacity. Each operational site or electrolyzer was assumed to be active 365 days per year and 24 h per day.

# 3 Results

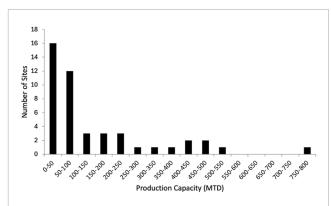
# 3.1 Production

#### 3.1.1 Activity data

Through the literature search process, a detailed list of hydrogen production activity data was consolidated and stored in Supplementary Table SA1. There were 104 Merchant and Byproduct (MB) sites identified with production capacities ranging from 1 MTD to 557 MTD (Supplementary Table SA1, Index 1–1.111). Most of the sites found (97%) were Merchant producers and utilized SMR. The remaining (3%) of sites produced hydrogen through Byproduct process streams. Most of the sites were gray hydrogen producers,



inventoried in the supplementary spreadsheet.

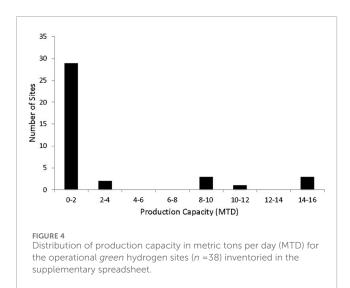


Distribution of production capacity in metric tons per day (MTD) for the operational captive hydrogen sites (n = 46) inventoried in the supplementary spreadsheet.

however two blue hydrogen production sites were identified. The total production capacity for MB sites was 14918.7 MTD, which is approximately 5.44 MMTA. Figure 2 shows the production capacity distribution of MB sites.

This study also found 46 Captive (C) hydrogen sites with production capacities ranging from 4.7 MTD to 787.7 MTD (Supplementary Table SA1, Index 14-14.46). All identified sites were refineries that produce hydrogen and utilize it on-site for various chemical processes. The total production capacity for C sites was 6818.97 MTD, which is approximately 2.49 MMTA. Figure 3 shows the production capacity distribution of C sites.

This study found 38 Green Hydrogen (G) sites with production capacities ranging from 0.5 MTD to 15 MTD (Supplementary Table SA1, Index 10-10.120). The three primary electrolyzer technologies are alkaline electrolysis (ALK), proton exchange membrane (PEM), and solid oxide electrolysis (SOEC) cells (National Renewable Energy Laboratory, 2022). In situations where the production capacity was unclear, the electrolyzer capacity was used to estimate production capacity through the following conversion factors in Equations 8–10 recommended by IEA (2023). When the electrolyzer technology was unclear it was assumed to



be PEM, since the majority of electrolyzers in the U.S. are PEM (Arjona, 2023). Figure 4 shows the production capacity distribution of G sites. The distribution of green hydrogen sites shows that green hydrogen production is still new and emerging in the U.S., majority of sites found (76%) had a relatively small production capacity (2 MTD or less).

ALK: 
$$0.0046 \,\mathrm{MW}/nm^3 H_2/\mathrm{hour}$$
 (8)

PEM: 
$$0.0052 \,\text{MW}/nm^3 H_2/\text{hour}$$
 (9)

SOEC: 
$$0.0038 \,\text{MW}/nm^3 H_2/\text{hour}$$
 (10)

#### 3.1.2 Emission factors selection

# 3.1.2.1 Steam methane reforming

SMR is the representative hydrogen production process for the *Merchant and Byproduct*, and *Captive* hydrogen stages. Emission sources associated with SMR production primarily include leakage through pipework and equipment purging (Esquivel-Elizondo et al., 2023). Hydrogen is typically emitted as part of the start-up and shutdown operational procedures associated with SMR, however in some plants this hydrogen is flared rather than vented (Arrigoni and Bravo Diaz, 2022). Hydrogen is also emitted from equipment purging during maintenance, during fault conditions, or to remove impurities during post-processing (Esquivel-Elizondo et al., 2023). In blue hydrogen production there can also be an additional purging during post-processing to remove residual H<sub>2</sub> in the stream (Frazer-Nash Consultancy, 2022; Esquivel-Elizondo et al., 2023).

Through the literature search process, a list of studies that presented emission factors for the SMR production process were compiled (Table 4). Xia et al. (2019) was the only study found that performed a small scale SMR lab experiment to estimate hydrogen leakage. In their experiment they use the nitrogen leakage detected via an online gas chromatograph to estimate the amount of hydrogen leaked. They estimated hydrogen leakage in their

experiment was less than 1% of the total hydrogen flux. Other studies proposed emission factors through modelling approaches. In their emissions rate model for blue hydrogen supply chain cases Cooper et al. (2022) utilized an emission rate of 0.55% (0.1%-1.0%) of hydrogen throughput. Arrigoni and Bravo Diaz (2022) also proposed that hydrogen losses from SMR are very limited and present a loss rate as low as 0.0001%. Other proposed emission factors for SMR from the literature were 1%-1.5% loss rate Fan et al. (2022) and 0%-0.5% loss rate (Frazer-Nash Consultancy, 2022). The TOPSIS methodology was applied to these emission factors (Table 4) and the emission factor presented in Xia et al. (2019) and Arrigoni and Bravo Diaz (2022) were found to have the highest and second highest performance  $(P_i)$  respectively. Thus, the best available emission factor was selected as a 0.1%-1% loss rate and utilized to estimate emissions for the Merchant and Byproduct, and Captive stages.

#### 3.1.2.2 Green hydrogen

Green Hydrogen production primarily occurs via electrolysis in the U.S. (National Renewable Energy Laboratory, 2022). Frazer-Nash Consultancy (2022) presents the four categories for potential leakage during electrolysis as leakage through casing and pipework, venting during start-up and shutdown, contamination of the vented oxygen (hydrogen crossover), and operational purging to remove impurities. Operational purging is proposed as a main source of leakage, while leakage through casing and pipework is presented as negligible (Frazer-Nash Consultancy, 2022). Other prominent leakage sources include processes such as dryers' regeneration, and hydrogen and oxygen crossover (Arrigoni and Bravo Diaz, 2022), however Frazer-Nash Consultancy (2022) proposes incorporating technology to recombine the hydrogen purged and vented during crossover as a means of mitigating emissions.

Through the literature search process, a list of studies that presented emission factors for the green hydrogen production via electrolysis were compiled (Table 5). Multiple studies (Fan et al. (2022); Esquivel-Elizondo et al., 2023) ascribe a H2 loss rate of 2%-4% of the produced hydrogen to the electrolysis process. This emission factor is based on electrolyzer system efficiency testing results by NREL that present H2 dryer loss rates during electrolysis being up to 4% of the produced hydrogen (Harrison and Peters, 2013). Fan et al. (2022) utilized a similar emission rate of 2.05% (0.1%-4.0%) of hydrogen throughput, in their emission rate model for their green hydrogen supply chain case. Other green hydrogen emission factors in the literature include Frazer-Nash Consultancy (2022) proposing a loss rate with a lower limit (LL) of 0.24%-3.32%, and an upper limit (UL) of 0.52%-9.2% of produced hydrogen in their emissions model. Arrigoni and Bravo Diaz (2022) presents a loss rate of 0.2% of produced hydrogen associated with electrolysis. The TOPSIS methodology was applied to these emission factors (Table 5) and the emission factor presented in Harrison and Peters (2013) and Fan et al. (2022) had the highest performance  $(P_i)$ . Thus, the best available emission factor was selected as 2%-4% loss rate and utilized to estimate emissions for the Green hydrogen stage.

TABLE 4 Comparative table ranking steam methane reforming emission factors utilizing TOPSIS methodology.

Literature Source	Reported <i>EF<sub>i</sub></i> (% weight loss)	Year of study ( <i>W<sub>j</sub></i> = 25%)	Measurement approach ( $W_j = 40\%$ )	Source reliability $(W_j = 35\%)$	P <sub>i</sub>	Rank
		Score (x <sub>ij</sub> )	Score (x <sub>ij</sub> )	Score (x <sub>ij</sub> )		
Xia et al. (2019), Fan et al. (2022)	<1%	3	2	3	1	1
Pranda and de Reals (2023), Arrigoni and Bravo Diaz (2022)	0.0001%	3	1	2	0.20	3
Cooper et al. (2022), Esquivel-Elizondo et al. (2023)	0.55% (0.1%–1.00%)	3	1	3	0.34	2
ConocoPhillips (2015), Barrett and Cassarino (2021), Fan et al. (2022)	1.0%-1.5%	2	1	2	0.00	5
Frazer-Nash Consultancy (2022)	0%-0.5%	3	1	2	0.20	3

The grading Scale for Year of Study, Measurement Approach, and Source Reliability are provided in Table 4. The Equations 1-5 were used to calculate the Pi.

TABLE 5 Comparative table ranking green hydrogen emission factors utilizing TOPSIS methodology.

<u>'</u>						
Literature Source	Reported <i>EF<sub>i</sub></i> (% weight loss)	Year of study ( $W_j$ = 25%)	Measurement approach ( <i>W<sub>j</sub></i> = 40%)	Source reliability $(W_j = 35\%)$	P <sub>i</sub>	Rank
		Score (x <sub>ij</sub> )	Score (x <sub>ij</sub> )	Score (x <sub>ij</sub> )		
Harrison and Peters (2013), Fan et al. (2022)	2%-4%	1	2	2	0.55	1
Cooper et al. (2022)	2.05% (0.10%-4.00%)	3	1	3	0.44	2
Frazer-Nash Consultancy (2022)	LL: 0.24%-3.32%; UL: 0.52%-9.20%	3	1	2	0.35	3
Arrigoni and Bravo Diaz (2022)	0.2%	3	1	2	0.35	3

The grading Scale for Year of Study, Measurement Approach, and Source Reliability are provided in Table 4. The Equations 1–5 were used to calculate the Pi.

# 3.2 Distribution and storage

#### 3.2.1 Liquefaction

#### 3.2.1.1 Activity data

Hydrogen is typically liquified when it is cooled down to a temperature below 20 K at 101.325 kPa. Liquefied hydrogen (L-H<sub>2</sub>) has a higher volumetric density compared to compressed H<sub>2</sub> gas (C-H<sub>2</sub>) and therefore provides a more cost-effective means of storing and transporting large quantities of H<sub>2</sub> over extended distances (Valenti, 2016; Cardella et al., 2017; Yin and Ju, 2020). The U.S. has the largest H<sub>2</sub> liquefaction capacity in the world, with over 15 H<sub>2</sub> liquefaction plants, and a collective liquefaction capacity of at least 326 MTD (Qiu et al., 2021). L-H<sub>2</sub> is typically stored in cryogenic tanks and transported via tanker trucks/trailers to

various end-uses (Arrigoni and Bravo Diaz, 2022). Current end-uses that are supplied with  $L-H_2$  in the U.S. include  $H_2$  refueling stations, FCEV, stationary fuel cell power stations, and applications in aviation such as fuel rocket engines (Cardella et al., 2017; Qiu et al., 2021; Chen et al., 2022).

#### 3.2.1.2 Emissions estimation

Emission sources associated with L- $\rm H_2$  conversion include leakage during start-up and maintenance, leaks from equipment (fugitive emissions, venting, purging, and boil-off), emissions from  $\rm H_2$  engine slip (incomplete combustion of  $\rm H_2$ ), and leaks during transfer operations such as truck loading and unloading (Arrigoni and Bravo Diaz, 2022; Cooper et al., 2022; Esquivel-Elizondo et al., 2023; Pranda and de Reals, 2023).

TABLE 6 The mass flow rates of H₂ and CH₄ assuming choked flow through an orifice at STPa

Gas	M <sub>m</sub>	k	$m_{ m choked}$
$H_2$	2.016	1.41	0.9746
$CH_4$	16.04	1.31	2.6795

The mass flow rate  $(m_{choked})$  calculated using Equation 11.

 $^{a}$ STP = standard temperature and pressure,  $T = 288.15 \, Kandp = 101 \, kPa$ 

Similar to other H<sub>2</sub> segments, the available emission factors for L-H<sub>2</sub> production are based on estimates, or models rather than empirical data (Cooper et al., 2022; Ocko and Hamburg, 2022; Esquivel-Elizondo et al., 2023). After a detailed literature search, Pranda and de Reals, 2023 were the only primary source that proposed an emission factor for U.S. L-H<sub>2</sub> production and is thus cited in most of the recent literature (Arrigoni and Bravo Diaz, 2022; Esquivel-Elizondo et al., 2023; Pranda and de Reals, 2023 from Air Liquide presented estimates for losses, mainly from the liquefaction process and boil-off during transfer operations, ranging from 10%-20% of the delivered L-H<sub>2</sub> (Arrigoni and Bravo Diaz, 2022). Applying this emission factor to the U.S. L-H<sub>2</sub> production capacity (326 MTA; Qiu et al., 2021) we estimate emissions from H<sub>2</sub> liquefaction, storage and transport can range anywhere from 11899 MTA to 23798 MTA.

# 3.2.2 Transmission pipeline

## 3.2.2.1 Activity data

The United States Pipeline and Hazardous Safety Administration reported that there are 1598.8 miles (2573 km) of H<sub>2</sub> transmission pipeline in the U.S. (PHMSA, 2024). This study's inventory of pipeline activity data is recorded in Supplementary Table SA1, Index 25-25.33. These transmission pipelines serve a large network of refineries, chemical manufacturing plants and hydrogen producers mainly along the Gulf Coast and in other geographic locations such as California and Indiana (Penev et al., 2019).

#### 3.2.2.2 Estimating emissions

The H<sub>2</sub> transmission pipeline emission factor was calculated using the assumption of choked flow through an orifice at standard temperature and pressure conditions (STP). The mass flow rate (m choked) of hydrogen and methane were calculated individually under the choked flow assumption using Equation 11 (Pritchard and Mitchell, 2015). Then, the mass flow rate ratio ( $m_R$ ) between the two gases was calculated using Equation 12. In Equations 11, 12 both gases are assumed to have the same upstream pressure and temperature. Equation 13 was used to calculate the specific gas constants (R) for hydrogen and methane, respectively. Table 6 provides the results for the mass flow rates of H2 and CH4 assuming choked flow through an orifice at STP. Thus the  $m_R$ between the two gases is 0.36. The  $m_R$  was then multiplied by the methane emission factor (10.9 kg/mile/year; GRI/EPA, 1996) for natural gas Transmission and Storage pipeline leaks to develop an emission factor for H2 transmission pipeline leaks. The developed emission factor was then multiplied by the relevant activity data resulting in an emissions estimate for H2 transmission pipelines of 6.3 MTA (Table 7).

$$m_{choked} = A_e p_0 \sqrt{\frac{k}{RT_0}} \left(\frac{2}{k+1}\right)^{\frac{(k+1)}{2(k-1)}}$$

$$m_R = \frac{m_{choked_{H_2}}}{m_{choked_{CH_4}}}$$

$$R = \frac{R_0}{M_m}$$
(12)

$$m_R = \frac{m_{choked_{H_2}}}{m_{choked_{CH_4}}} \tag{12}$$

$$R = \frac{R_0}{M_{\rm vir}} \tag{13}$$

where  $A_e$  is the pipeline exit area,  $p_0$  is the upstream pressure,  $T_0$  is the upstream temperature,  $R_0$  is the universal gas constant.

# 3.2.3 Underground Hydrogen Storage

#### 3.2.3.1 Activity data

There are three UHS facilities in the U.S: (1) Clemens, (2) Moss Bluff, (3) Spindletop (Zhu et al., 2023). These UHS facilities in Texas all use salt caverns to store H<sub>2</sub>, especially in periods of low demand. Then, to minimize production inefficiencies due to fluctuations in demand, the stored H<sub>2</sub> is used to supplement supply to refineries and other petrochemical customers in the gulf coast (Panfilov, 2016). These three facilities have a collective net storage capacity or total cavern storage volume of 2,046,000 m<sup>3</sup> (Panfilov, 2016; Zivar et al., 2021; Zhu et al., 2023). In each facility, cushion gas is always injected into the salt cavern and remains there for the entire period the cavern is used for storage (Tarkowski and Uliasz-Misiak, 2022; Miocic et al., 2023). The purpose of cushion gas is to maintain a defined minimum pressure in the cavern, thus allowing effective injection and extraction of H2 gas. Therefore, the usable H<sub>2</sub> capacity of a UHS salt cavern equals the net storage capacity minus the cushion gas capacity.

#### 3.2.3.2 Estimating emissions

Previous research has identified the primary pathways for H<sub>2</sub> leakage in Salt Cavern Hydrogen Storage (SCHS) as leakage losses through the wellbore, caprock, and cavern (Zhu et al., 2023). Wellbore losses can occur due to corrosion overtime, mechanical failure, or hydrogen embrittlement (Zhu et al., 2023). Other sources of emissions are associated with surface equipment during operation (e.g., compressors, pipework), including deliberate releases during scheduled or emergency shutdowns, and maintenance (Fan et al. (2022); Frazer-Nash Consultancy, 2022; Esquivel-Elizondo et al., 2023).

After a detailed literature search, the authors were unable to find any empirical research detailing H2 emission factors associated with SCHS Systems. Previous studies acknowledge major uncertainty associated with SCHS emissions due to a lack of data (Fan et al. (2022); Frazer-Nash Consultancy, 2022; Esquivel-Elizondo et al., 2023; Zhu et al., 2023). Consequently, a best-engineering estimate was developed for U.S. SCHS based on the available body of existing knowledge and research. Table 8 details the properties utilized to estimate emissions (Table 9) for the U.S. UHS sites. A detailed estimation methodology is provided in the following paragraphs.

Previous published literature has provided key details regarding the operating conditions (e.g., operating pressure, depth, and storage capacity) of the U.S. SCHS sites (Panfilov, 2016; Zivar et al., 2021; Miocic et al., 2023; Zhu et al., 2023; Ruiz Maraggi and Moscardelli, 2024). For instance, Panfilov, 2016 reports the usable H<sub>2</sub> capacity of the Clemens cavern as 2520 metric tons. Based on the operating pressure of Clemens cavern (70-137 bar) provided in Zhu et al.,

TABLE 7 Hydrogen transmission pipeline activity data, emission factor, and emissions estimate.

l	H <sub>2</sub> source	Activity data, <i>A<sub>i</sub></i> (km)	CH <sub>4</sub> pipeline leaks <sup>a</sup> (kg/km/year)	$m_R$	H₂ pipeline leaks estimate, EF; (kg/km/year)	$H_2$ pipeline emissions, $E_i$ (MTA)
	Transmission Pipelines	2573	6.8	0.36	2.46	6.34

<sup>&</sup>lt;sup>a</sup>GRI/EPA, 1996.

TABLE 8 Properties of underground hydrogen storage sites.

UHS sites	Storage volume (m³)	Average operating pressure (kPa)	Estimated H <sub>2</sub> density (kg/m <sup>3</sup> )	Useable H <sub>2</sub> capacity (MT)
Clemens	580000	10250	8	2520 <sup>a</sup>
Moss Bluff	566000	10350	8	2268 <sup>b</sup>
Spindletop	906000	13500	10	4735 <sup>b</sup>

The usable H<sub>2</sub> capacity in MT, is being calculated based on recommendations from research literature.

TABLE 9 Calculation of total and per site hydrogen emissions for underground hydrogen storage sites.

UHS sites	Injection/Extraction rate (MTD)	Total injection activity, <i>A<sub>i</sub></i> (MTA)	Emissions <i>E<sub>i</sub></i> lower limit (MTA)	Emissions <i>E<sub>i</sub></i> upper limit (MTA)
Clemens	138.60	25294.50	5	15
Moss Bluff	124.73	22763.48	5	14
Spindletop	260.42	47527.38	10	29
Total	523.76	95585	19	57

2023, we estimate the usable  $\rm H_2$  to account for about 50% of its net storage capacity (Karakilcik and Karakilcik, 2020), with the cushion gas accounting for the remaining capacity. Hence, in our estimation calculations we assumed the usable  $\rm H_2$  capacity to cushion gas ratio was also 50% in the Moss Bluff and Spindletop caverns (Table 8). This assumption is consistent with previous research that specifies cushion gas typically occupies at least 20%–30% of the salt cavern volume (Crotogino et al., 2010; Karakilcik and Karakilcik, 2020; Tarkowski and Uliasz-Misiak, 2022).

Previous studies have also provided guidance on the withdrawal rates, injection/withdrawal cycles, and emission rates associated with SCHS. The maximum salt cavern filling and discharge rates are governed by 1 MPa/day, which corresponds to maximum withdrawal rates of approximately 10% of the cavern storage capacity per day (Crotogino et al., 2010; Karakilcik and Karakilcik, 2020). Salt caverns can experience up to 10 injection/withdrawal turnovers per year (Karakilcik and Karakilcik, 2020; Tarkowski and Uliasz-Misiak, 2022). Additionally, previous UHS models have proposed emission rates from surface operations (equipment leaks, compressor seals, depressurization) to typically range from 0.02%–0.06% of the stored hydrogen mass per year (Fan et al. (2022); Frazer-Nash Consultancy, 2022; Esquivel-Elizondo et al., 2023). Hence, to develop an emissions

estimate for U.S. SCHS, we assumed each site is operating year-round, with 10 injection/withdrawal turnovers per year. We used a salt cavern filling and discharge rate of 5.5% of the cavern storage capacity per day. This would mean every 18.25 days there is an injection/withdrawal turnover. We also build upon previous UHS models and incorporate a daily  $\rm H_2$  emissions rate of 0.02%–0.06% of the stored  $\rm H_2$ . This results in an estimate for the maximum emissions from U.S. SCHS ranging from 19 MTA to 57 MTA (Table 9).

# 3.2.4 FCEV refueling stations

#### 3.2.4.1 Activity data

Majority of the market for FCEV refueling stations are in California (US DOE, 2023). The setup and infrastructure utilized in FCEV refueling stations vary, depending on the form of hydrogen delivered to the station (i.e., C-H<sub>2</sub> vs. L-H<sub>2</sub>), and the class of vehicle being served (i.e., passenger vehicles, fuel cell busses, and class 8 trucks; (California Energy Commission, 2024). Losses in FCEV refueling stations are primarily due to purging in the case of C-H<sub>2</sub>, and boil off when dealing with L-H<sub>2</sub> (Arrigoni and Bravo Diaz, 2022; California Energy Commission, 2024) reported 47 operational stations, and 38 planned stations.

<sup>&</sup>lt;sup>a</sup>Retrieved from Panfilov (2016).

bCalculated using a Usable H<sub>2</sub> to Cushion Gas Ratio of 50%.

TABLE 10 Emission calculations for fuel cell electric vehicle refueling stations.

Vehicle category	Passenger vehicle	Transit bus	Class 8 truck
Fuel consumption (kg/100 km)	0.76 <sup>a</sup>	8.67 <sup>b</sup>	7.75 <sup>c</sup>
Yearly distance travelled (km/yr)	17,012 <sup>d</sup>	69,090 <sup>d</sup>	100,030 <sup>d</sup>
Number of vehicles in the U.S.	18,459 <sup>e</sup>	173 <sup>f</sup>	30 <sup>f</sup>
Estimated $H_2$ demand, $A_i$ (MTA)	2387	1035	233
Refueling Station $H_2$ loss rate $EF_i$	6% (2%-10%)	6% (2%-10%)	6% (2%–10%)
$H_2$ emissions estimate, $E_i$ (MTA)	143.19 (47.73–238.66)	62.08 (20.69–103.47)	13.95 (4.65–23.26)

This table presents vehicle specific information found in the literature research. The according sources are denoted and referenced below.

#### 3.2.4.2 Estimating emissions

To estimate emissions for FCEV refueling stations, the estimated  $\rm H_2$  demand was used as the activity data for calculating emissions. The estimated  $\rm H_2$  demand was calculated by multiplying the fuel consumption by the yearly distance travelled of representative or comparable vehicles, and the number of FCEVs in the U.S (Table 10).

Only 2 studies were found that propose loss rates from FCEV refueling stations (Arrigoni and Bravo Diaz (2022); Genovese et al., 2020; Arrigoni and Bravo Diaz (2022)). Genovese et al., 2020 proposed a loss rate of 2%–10% of the hydrogen from an L-H<sub>2</sub> refueling station, based on mass flow meter readings, over a 5-month period. Arrigoni and Bravo Diaz (2022) presents a loss rate of 3%–8.5% of the delivered hydrogen, depending on the form of hydrogen delivered (C-H<sub>2</sub> vs. L-H<sub>2</sub>). To provide a conservative estimate, a loss rate of 2%–10% of the delivered hydrogen is utilized as the best available emission factor in this study. Multiplying this loss rate by the estimated H<sub>2</sub> demand provides an estimate for FCEV refueling stations emissions in the U.S (Table 10).

#### 4 Discussion

# 4.1 Overall emissions and contributions

The goal of this study was to review the hydrogen emissions literature in order to develop a baseline  $\rm H_2$  emissions inventory for the U.S. production, distribution and storage segments. The scope of this study was limited to pure hydrogen emissions and excludes emissions from low purity  $\rm H_2$  streams. This study also provides detailed insights regarding key emission or leakage sources and causes from each supply chain stage.

Figure 5 presents a U.S. map showing the geographic distribution of the operational hydrogen production facilities found in this studies literature search. The results of this study's inventory detail the current U.S.  $\rm H_2$  production capacity to be about 7.97 MMTA (Table 11). This number is slightly lower

than but is still consistent with previous reporting of U.S. hydrogen production being about 10 MMTA (US DOE, 2018; International Energy Agency, 2024; U.S. Energy Information Administration, 2024). Hydrogen production companies are a difficult population to obtain data from. A potential reason for the small discrepancy between this inventory's production results and previous reporting could be a lack of clarity or distinction between decommissioned sites, planned sites, and operational sites. This study is unique because it presents an inventory of production facilities to corroborate its results and is careful to distinguish between decommissioned, planned, and operational sites when calculating the U.S. H<sub>2</sub> production capacity results. This study also estimated total hydrogen emissions from the U.S. production, distribution and storage segment to be about 0.0629 MMTA (Table 11), with emission bounds ranging from 0.02 to 0.11 MMTA. Thus, this study estimates the hydrogen leakage or emission rate to be about 0.79%, with emission rate bounds ranging from 0.26% to 1.32%. Substituting the International Energy Agency (2024) reported U.S. hydrogen production output of 10 MMTA as the production activity data in this study only reduces the resulting emission rate by 0.05%-0.12% which indicates a negligible discrepancy in terms of sensitivity between utilizing the International Energy Agency (2024) activity data or using this study's inventoried production activity data.

This study significantly contributes to the research literature by (i) consolidating a detailed list of hydrogen activity data (Supplementary Table SA1) and emission factors (Supplementary Table SA2) and (ii) estimating emissions across the relevant supply chain stages. The insights and conclusions from this study can provide direction for hydrogen production companies and researchers as they develop hydrogen emission mitigation measures, controls, and goals. The inventory framework developed in this study can also be utilized to inform the design and implementation of future facility-level hydrogen emission measurement studies and can serve as a living document that can be updated and enhanced as more empirical data is obtained.

<sup>&</sup>lt;sup>a</sup>H2 live, 2024.

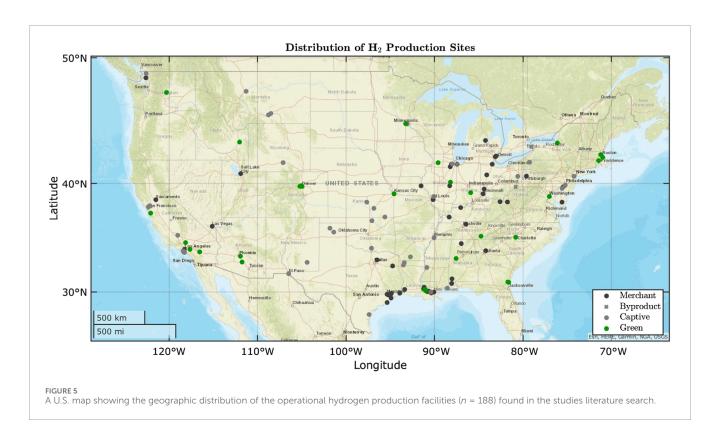
bNew Flyer, 2024.

<sup>&</sup>lt;sup>c</sup>Hyundai, 2024.

<sup>&</sup>lt;sup>d</sup>U.S. Energy Information Agency, 2025.

eArgonne National Laboratory, 2024.

<sup>&</sup>lt;sup>f</sup>California Energy Commission, 2024.



# 4.2 Limitations and data gaps

Previous research is clear about the lack of empirical data available to quantify hydrogen emissions across the supply chain, especially in the end-use segment (Arrigoni and Bravo Diaz, 2022; Cooper et al., 2022; Fan et al., 2022). Therefore, it is important to clarify the limitations of this study. Firstly, majority of the emission factors were either retrieved from hydrogen emission reviews, estimated, or modelled. Few emission factors were based on smallscale experiments, and none of the emission factors were based on large-scale hydrogen measurement studies. Secondly, end-use emissions were not included in the results of this study due to the lack of data availability (activity data and emission factors) needed to sufficiently estimate majority of end-use emissions (Arrigoni and Bravo Diaz, 2022; Fan et al., 2022). One exception is emissions from FCEV's, which were estimated based on data available in the literature but not included in the final inventory. The FCEV emissions estimation is based on a FCEV exhaust analysis performed by Pearman et al. (2021). As presented in detail in Section 4.2.2, the total emissions of H<sub>2</sub> from FCEV's was estimated to be 77.83 MTA. Lastly, the production activity data utilized in this study is based on production capacity rather than output and thus does not account for downtime in production due to maintenance or other unexpected occurrences.

#### 4.2.1 End-use emissions

Most hydrogen end-use applications can be assigned to one of three broad categories: industrial facilities, energy or electricity generation, and mobility or transportation. Industrial facilities include ammonia and methanol production, petroleum refineries, chemical or synthetic fuels, and iron or steel production. Electricity

generation includes hydrogen generators, stationary fuel-cells, and building heating. Transportation includes FCEVs and applications in aviation, and shipping. Hydrogen usage and reporting is still emerging in the U.S. market thus the exact usage for each of the applications is unclear. However, previous sources suggest that a vast majority (about 90%) of the produced hydrogen in the U.S. is used for petroleum refining, ammonia production (fertilizer production), and methanol production. While, the remaining (about 10%) are utilized in other applications including treating metals, and processing foods (Connelly et al., 2019; National Renewable Energy Laboratory, 2022). Hydrogen end-use leakage is the least understood compared to the production and distribution segments (Fan et al., 2022). Hydrogen enduse emission studies are primarily focused on quantitative risk analysis and plant surveys to prevent large-scale fires or explosions (Spouge, 2005; Mohammadfam and Zarei, 2015). Quantification of facility level leaks (small and distributed leaks) are less urgent and thus largely absent from the literature (Fan et al., 2022). Previous studies have proposed or assumed a similar emission factor (0.5%) as gray hydrogen production to emissions from industrial facilities such as refining, chemical or synthetic fuels, and iron or steel production (Fan et al., 2022). In various modelling studies, loss rates between 0.5%-3% have been assumed or proposed for electricity generation, and building heating; based on comparable natural gas infrastructure (Alvarez et al., 2012; Fischer et al., 2018; Hormaza Mejia and Brouwer, 2018; Fan et al., 2022). Other hydrogen end-use emission factors are virtually unexplored. Quantifying hydrogen end-use activity data and emission factors represents a major data gap and is an important future work for estimating emissions from the entire

TABLE 11 Overall emissions estimate based on best available emission factors and activity data for the U.S. production, distribution and storage segments.

H <sub>2</sub> supp	oly chain	Activity	data <i>A<sub>i</sub></i>	Emission factors <i>EF</i> <sub>i</sub>		Emission, <i>E<sub>i</sub></i> (MTA)
Source	Stage	Average	Units	Average (Min – Max)	Units	Average (Min – Max)
SMR	Merchant and Byproduct	5,445.32	MTA	0.55% (0.1%–1%)	loss rate	29,949.25 (5,445.32–54,453.19)
SMK	Captive	2,488.93	MTA	0.55% (0.1%–1%)	loss rate	13,689.09 (2,488.93–24,889.25)
Electrolysis	Green Hydrogen	32.52	MTA	3% (2%-4%)	loss rate	975.47 (650.32–1,300.63)
C-H <sub>2</sub>	Transmission Pipeline	2573	km	2.46	kg/km/yr	6.34
L-H <sub>2</sub>	Production and Delivery	326	MTD	15% (10%–20%)	Delivered L-H <sub>2</sub>	17,849 (11,899–23,798)
	Clemens	25294.50	MTA	0.04% (0.02%-0.06%)	loss rate	10.12 (5.06–15.18)
UHS	Moss Bluff	22763.48	MTA	0.04% (0.02%-0.06%)	loss rate	9.11 (4.55–13.66)
	Spindletop	47527.38	MTA	0.04% (0.02%-0.06%)	loss rate	19.01 (9.51–28.52)
	Passenger Vehicles	2,386.58	MTA	6% (2%-10%)	loss rate	143.19 (47.73–238.66)
FCEV Refueling Stations	Transit Buses	1,034.70	MTA	6% (2%-10%)	loss rate	62.08 (20.69–103.47)
	Class 8 Trucks	232.57	MTA	6% (2%-10%)	loss rate	13.95 (4.65–23.26)
Total Emissions (MMTA)						0.063 (0.0206-0.105)
Total Production (MMTA)						7.971
Emissions Rate						0.788% (0.259%–1.318%)

This table was consolidated based on all the results detailed in Section 3: Results and Discussion. A full annotated version is provided in Supplementary Table SA3.

TABLE 12 Overall emissions estimate based on best available emission factors and activity data for the U.S. production, distribution and storage segments.

Vehicle category	Passenger vehicle	Transit bus	Class 8 truck
Fuel consumption (kg/100 km)	$0.76^{a}$	8.67 <sup>b</sup>	7.75°
Yearly distance travelled (km/yr)	17,012 <sup>d</sup>	69,090 <sup>d</sup>	100,030 <sup>d</sup>
Number of vehicles in the $\mathrm{U.S}A_i$	18,459 <sup>e</sup>	173 <sup>f</sup>	30 <sup>f</sup>
Estimated $H_2$ exhaust emission factor, $EF_i$ (kg/yr/vehicle)	2.75	127.39	165.12
$\rm H_2$ exhaust emissions estimate, $E_i$ (MTA)	50.83	22.04	4.95

This table presents vehicle specific information found in the literature research. The according sources are denoted and referenced below. The emission factors only account for the emissions in the exhaust and no leakage losses. The annual fuel consumption is calculated from the fuel consumption and the yearly distance travelled.

<sup>&</sup>lt;sup>a</sup>H2 live, 2024.

<sup>&</sup>lt;sup>b</sup>New Flyer, 2024.

<sup>&</sup>lt;sup>c</sup>Hyundai, 2024.

<sup>&</sup>lt;sup>d</sup>U.S. Energy Information Agency, 2025.

<sup>&</sup>lt;sup>e</sup>Argonne National Laboratory, 2024.

<sup>&</sup>lt;sup>f</sup>California Energy Commission, 2024.

hydrogen value-chain. In Section 4.2.2 we propose FCEV emissions estimation process.

# 4.2.2 Fuel cell electric vehicles emissions

Hydrogen emissions from FCEVs includes hydrogen residues in the exhaust from the fuel cell due to incomplete reactions, permeation through storage vessels, and leakage from high pressure valves, pipelines, and joints (Hao et al., 2020; Pearman et al., 2021). Emissions from FCEVs also occur during start-up, shut-down, and vehicle idling (Hao et al., 2020).

There are three main groups of FCEVs in the U.S: (1) Passenger vehicles, (2) Fuel cell busses, and (3) Class 8 trucks. Majority of the FCEV market and infrastructure (including refueling stations) in the U.S. is currently in California (US DOE, 2023). Argonne National Laboratory (2024) reports 18,459 FCEV passenger vehicles have been sold in the U.S. since 2014. California Energy Commission (2024) reports 173 fuel cell transit busses, and 30 Class 8 trucks in operation in California.

There is a lack of empirical research detailing leakage rates and emission factors for FCEVs. Only one study was found that proposed an average hydrogen exhaust concentration (Pearman et al., 2021). Pearman et al. (2021) utilized a prototype FCEV exhaust analyzer to perform an exhaust analysis on a commercial FCEV. Under simulated urban driving conditions, their graphical analysis indicates an average exhaust hydrogen concentration of about 0.75%. Thus, we applied the exhaust hydrogen concentration proposed by Pearman et al. (2021) through fuel cell stochiometric relationships to estimate an exhaust emission factor for FCEVs.

An emission factor for the hydrogen residues in the exhaust was estimated for the three classes of FCEVs: passenger vehicles, fuel cell busses, and class 8 trucks. Since majority of the FCEV market is currently in California, it is used as the reference state in this analysis. The Toyota Mirai, New Flyer xcelsior Charge H<sub>2</sub> bus, and the Hyundai XCIENT Fuel Cell Truck are used as the representative passenger vehicles, fuel cell bus, and class 8 truck respectively in our estimation. The Toyota Mirai represents 88% of FCEV passenger vehicles in California, the New Flyer xcelsior Charge H2 bus represents 60% of FCEV transit buses, while all class 8 trucks in California are a Hyundai XCIENT Fuel Cell Trucks (California Energy Commission, 2024). The hydrogen exhaust emissions were estimated for the passenger vehicle and scaled up to the bus and class 8 truck based on their fuel cell stoichiometry (Equations 14-18) and the relative fuel consumption of the respective vehicles. The methodology employed in these calculations is detailed in the following paragraphs. Our estimation assumed that dry air is used at the inlet of the fuel cell, the entire amount of O2 reacts with the H2 in the fuel cell, and the produced water is vaporized. Based on the stochiometric equation (Equation 14), and the molar conservation requirements for hydrogen, oxygen and nitrogen (Equations 15-17), the molar exhaust ratios can be calculated using Equation 18:

$$H_2 + a(O_2 + 3.76N_2) \rightarrow bH_2O + cH_2 + dN_2$$
 (14)

$$b + c = 1 (H_2) (15)$$

$$a = b/2(O_2) \tag{16}$$

$$d = 3.76 a(N_2) \tag{17}$$

$$X_{i,2} = \frac{x_i}{b+c+d} \tag{18}$$

where  $x_i = b$ , c or d for  $H_2O$ ,  $H_2$  and  $N_2$  respectively.

hydrogen exhaust using the concentration result from Pearman et al. (2021),  $X_{H_2,2} = 0.0075$ , the molar amounts can be calculated to (a = 0.4893, b = 0.9787, c = 0.0212, d = 1.8400) which in turn result in the mole fractions  $X_{H_2O,2} = 0.3446$  and  $X_{N_2,2} =$ 0.6478. Under ambient conditions, the mole fraction of water in the air is between 0.5% and 1.0% which is relatively small compared to the 34.46% in the exhaust. This further corroborates the assumption of dry air as valid.

Then the molar fractions are converted into mass fractions using Equation 19, with  $M_m$  being the molecular weight of the according molecule. The total inlet mass flow can be calculated using Equation 20. Where the mass flow rate of hydrogen at the inlet  $(\dot{m}_{H_2,1})$  is derived by multiplying the yearly distance travelled and the fuel consumption of comparable or representative vehicles (Table 12). For instance, in the case of FCEV passenger vehicles, the average yearly distance travelled of passenger cars in general is used (U.S. Energy Information Agency, 2025).

Then, by multiplying the mass fraction of hydrogen at the outlet with the total mass flow rate (Equation 21), the hydrogen exhaust mass flow rate  $(\dot{m}_{H_{2},2})$  can be calculated. In Table 12, the annualized hydrogen emission factors for each vehicle category is presented based on the previously stated methodology and assumptions. Based on this, the yearly hydrogen exhaust emissions was estimated per vehicle type.

$$Y_{i,2} = \frac{X_{i,2} M_{m_i}}{\sum X_i M_{m_i}} \tag{19}$$

$$Y_{i,2} = \frac{X_{i,2} M_{m_i}}{\sum X_j M_{m_j}}$$

$$\dot{m}_T = \left(1 + a \frac{M_{mO_2}}{M_{mH_2}} + 3.76 a \frac{M_{mN_2}}{M_{mH_2}}\right) \dot{m}_{H_2,1}$$
(20)

$$\dot{m}_{H_2,2} = Y_{H_2,2} \dot{m}_T \tag{21}$$

This estimation methodology results in a H<sub>2</sub> exhaust emissions estimate from FCEV's of 77.83 MTA. Future work should focus on large-scale hydrogen measurement research to empirically measure emissions from the hydrogen transportation sector (FCEV's and FCEV fueling stations). Previous pump to wheel studies aimed at quantifying methane emissions from the transportation sector can be utilized as a framework and guide (Clark et al., 2017).

### 4.3 Future work

This study is the first to present a baseline inventory of hydrogen emissions in the U.S. Its goal is to provide an initial inventory framework that can be updated as more empirical data is obtained. Future work should focus on large-scale hydrogen measurement research to empirically measure, quantify, and confirm emissions across the supply chain (production to end-use). Hydrogen enduses include ammonia and methanol production, oil refining, transportation, power generation and other industrial processes. With the recent announcement of a new sensor capable of sitelevel emission rate quantification (Momeni et al., 2025), future

measurement studies can utilize tracer flux ratio to quantify sitelevel emissions (Marchese et al., 2015; Mitchell et al., 2015). The tracer flux ratio technique involves releasing tracer gas at the hydrogen production site at known flow rates (typically 5 to 40 standard liters per minute). The concentrations of the tracer gas and hydrogen emitted from the site can then be measured downwind of the facility. The measured downwind concentration ratios of the hydrogen gas to the tracer gases, in combination with the known tracer release rate, can then be used to determine the facility-level emission rate of hydrogen (Mitchell et al., 2015). Additionally, physics-based Bayesian plume model inversion methodologies as detailed in (Momeni et al., 2025) can also be utilized to quantify facility-level hydrogen emission rates. These methodologies would be particularly useful for site-level quantification (e.g., SMR, electrolysis, ammonia and methanol production) and for quantifying emission from equipment at the process level (e.g., pipelines, storage, loading/unloading). Future empirical measurement research will reduce the uncertainty depicted by the upper and lower emission rate estimates in this study and provide empirical data for hydrogen processes outside of the scope of this study. Additionally, previous pump to wheel studies aimed at quantifying methane emissions from the transportation sector can be utilized as a framework and guide for measuring hydrogen emissions from FCEV's and FCEV fueling stations (Clark et al., 2017).

# **Author contributions**

JD: Writing - review and editing, Formal Analysis, Visualization, Validation, Investigation, Methodology, Data curation, Writing - original draft, Conceptualization. KM: Formal Analysis, Visualization, Methodology, Supervision, Conceptualization, Project administration, Data curation, Investigation, Writing - review and editing. NV: Investigation, Writing - original draft, Visualization, Data curation, Formal Analysis, Methodology. PD: Data curation, Formal Analysis, Methodology, Investigation, Writing - original draft. CM: Formal Analysis, Methodology, Visualization, Data curation, Writing original draft, Investigation. HM: Formal Analysis, Methodology, Data curation, Visualization, Writing – original draft, Investigation. DO: Writing - original draft, Formal Analysis, Methodology, Conceptualization, Visualization, Data curation, Investigation. NL: Methodology, Formal Analysis, Data curation, Writing original draft, Visualization, Conceptualization. MH: Data curation, Visualization, Conceptualization, Project administration, Writing - review and editing, Methodology. AM: Funding acquisition, Formal Analysis, Visualization, Supervision, Project administration, Conceptualization, Validation, Writing - review and editing, Methodology, Investigation.

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# Conflict of interest

Authors KM, PD, CM, HM, DO, and MH were employed by SLR Consulting.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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# Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fenrg.2025.1650479/full#supplementary-material

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kPa

MT

MTA

Kilopascal

Metric Tons (equivalent to kg)

Metric Tons per Year

# Nomenclature

# Chemistry compound

 $CO_2$ Carbon dioxide MTD Metric Tons per Day

MMTA Million Metric Tons per Year  $H_2$ Hydrogen

Megawatts MW  $H_2O$ Water Nitrogen  $N_2$ Year

Normal Cubic Meter  $nm^3$  $O_2$ Oxygen

# Symbol

Pipeline exit area  $\mathbf{A}_{\mathbf{e}}$ H<sub>2</sub> activity data  $A_i$ 

 $\mathbf{A}_{ij}$ Normalized attribute value

E Total  $H_2$  emissions

Emission per supply chain stage i in MTA  $\mathbf{E}_{\mathbf{i}}$ 

 $EF_i$ Best available emission factor for supply chain stage i

K Isentropic exponent

Mean M

Mass flow rate  $\mathbf{m}_{\mathrm{choked}}$ ṁ mass flow

Molecular weight  $M_{\rm m}$ Mass flow rate ratio  $m_R$ MWMolecular weight

Pipeline upstream pressure

Performance  $\boldsymbol{P}_{i}$ 

 $R_0$ Universal gas constant

Ideal worst Ideal best  $S_i^+$ 

SD Standard deviation

Pipeline upstream temperature  $T_0$ 

 $W_i$ Weighing factor

Hydrogen exhaust concentration  $X_{i,2}$ 

TOPSIS score Xij

Molecular mass fraction at outlet  $Y_{i,2}$ 

# Units

Joule K Kelvin Kilogram kg Kilometer km Mole

Megapascal per day MPa/day