## **An Environmental Cost Basis for Regulating Aviation NO<sup>x</sup> Emissions**

by

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B.S. Aeronautical and Mechanical Engineering Clarkson University, 2013

### SUBMITTED TO THE DEPARTMENT OF AERONAUTICS AND ASTRONAUTICS IN PARTIAL FULFILLMENT OF THE REQUIREMENTS FOR THE DEGREE OF

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Signature of Author: **Example 20** Figure 20 Figure

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Submitted to the Department of Aeronautics and Astronautics on May 16, 2019 in Partial Fulfilment of the Requirements for the Degree of Master of Science in Aeronautics and Astronautics

#### **Abstract**

Nitrogen oxides and carbon dioxide are two by-products of combustion in aircraft engines, and have different impacts on the environment. Nitrogen oxides ( $N O<sub>x</sub>$ ) are both an air quality concern and an indirect contributor to radiative forcing, while carbon dioxide ( $CO<sub>2</sub>$ ) is a longlived greenhouse gas.

The International Civil Aviation Organization has been responsible for evaluating and setting commercial aircraft  $NO<sub>x</sub>$  emissions standards since 1981. Each of the historical standards has been more stringent than the previous and, when implemented, requires newly certified engines to produce less  $NO<sub>x</sub>$  per unit rated thrust. Each iteration has been defined as a function of engine overall pressure ratio, which then links the engine cycle, and implicitly fuel burn and  $CO<sub>2</sub>$  emissions, to allowable  $NO<sub>x</sub>$  levels

These regulations have historically been evaluated and implemented with a focus on reducing adverse air quality impacts around airports, but the thermodynamic tradeoff with  $CO<sub>2</sub>$ requires additional analysis to quantify net climate impacts. This paper introduces a social cost basis for evaluating aviation  $NO<sub>x</sub>$  emissions regulations, and quantifies air quality damage, climate damage, and fuel costs associated with allowable emission levels. The result is monetized environmental and fuel costs associated with certain emission standards.

Results show higher overall pressure ratio engines operating at the current  $NO<sub>x</sub>$  regulatory limit are allowed more environmental damage per unit rated thrust than lower overall pressure ratio engines, therefore allowing uneven social costs per unit thrust (i.e. fuel and environmental costs combined) across the engine design space. This is a consequence of the definition of the regulation today, where higher pressure ratio engines are allowed higher  $NO<sub>x</sub>$  emissions. Alternative regulation definitions are evaluated which consider the engine cycle and combustor together. Achieving constant social costs requires the regulation to decrease in slope at higher pressure ratios, corresponding to the diminishing marginal efficiency improvements, instead of increasing slope in that region.

Thesis Supervisor: Steven R.H. Barrett Title: Associate Professor of Aeronautics and Astronautics

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#### <span id="page-14-0"></span>**Introduction**

In 1981, the International Civil Aviation Organization (ICAO) adopted the first set of  $NO<sub>x</sub>$ emission standards for commercial aircraft, restricting NO<sub>x</sub> emissions during landing and takeoff (LTO) with an aim to reduce adverse air quality impacts near airports. This policy was defined in terms of total NO<sub>x</sub> emissions below 3,000 ft per unit rated thrust ( $D_p/F_{oo}$ ), as a function of engine overall pressure ratio (OPR,  $\pi_{\infty}$ ). In the years since, four more stringent standards have been adopted, shown in [Figure 1,](#page-19-0) the latest being in 2011 (ICAO, 2008).

 $NO<sub>x</sub>$  has historically been regarded as a local air quality concern around airports due to its role in the formation of particulate matter (PM<sub>2.5</sub>) and ozone (O<sub>3</sub>). More recent research has shown that PM<sub>2.5</sub> and O<sub>3</sub> attributable to cruise  $NO<sub>x</sub>$  emissions are also a significant contributor to premature mortalities. Eastham and Barrett (2016) estimated that ~6,800 premature mortalities per year can be attributable to  $O_3$ , while PM<sub>2.5</sub> contributes ~9,200 from current aircraft operations. Barrett et al. (2010) estimated that ~80% of aviation premature mortalities are attributed to cruise and  $\sim$ 20% to LTO. Yim et al. (2015) found aviation-attributed PM<sub>2.5</sub> and O<sup>3</sup> exposure results in ~16,000 premature mortalities, with 75% attributed to cruise operation. The studies indicate that aviation air quality impacts from cruise are greater than LTO operations.

 $NO<sub>x</sub>$  emissions also have climate impacts arising from short term ozone production (warming), long-term methane destruction (cooling), long-term ozone depletion (cooling), and formation of nitrate aerosols (cooling) (Grobler et al., 2019). It is estimated the aviation industry currently accounts for 5% of global anthropogenic radiative forcing (RF) (Lee et al., 2009).

 $NO<sub>x</sub>$  emission levels are a function of engine combustor design (residence time, mixing rates, equivalence ratios, and temperatures), and also a function of the overall engine cycle design which sets the combustor inlet and exit temperatures. Aircraft fuel burn and  $CO<sub>2</sub>$ emissions are determined by the engine cycle, with higher OPR and higher temperature cycles providing higher efficiency, and therefore lower fuel consumption and  $CO<sub>2</sub>$  emissions. However, as described by the Zeldovich mechanism, higher temperatures increase the rate of thermal NO formation, which is the dominant  $NO<sub>x</sub>$  production mechanism in gas turbine engines (Kundu et al., 2013). Kyprianidis et al. (2017) showed for rich-burn quick-quench lean-burn (RQL) combustors  $NO<sub>x</sub>$  increases exponentially with OPR.

Due to this relationship between the engine cycle and  $NO<sub>x</sub>$  formation, each iteration of the ICAO NO<sub>x</sub> standard has set allowable NO<sub>x</sub> emission levels as a function of OPR, with the result being that a given combustor will have similar margins to the standard when used in engines with different OPR. This approach creates a tradeoff between engine efficiency and  $NO<sub>x</sub>$ -related environmental impacts that must be considered in the design of aircraft engines.

Freeman et al. (2018) discuss this  $NO_x/CO_2$  tradeoff in the context of climate impacts, and aim to find a climate "breakeven" point for total RF given different scenarios. It is shown that by decreasing  $NO_x$  emissions by 20% (per updated policy) and increasing  $CO_2$ emissions by 2% (per assumed technological tradeoffs required to meet that policy today) as compared to a baseline case, total warming actually increases. This highlights a risk of further restricting  $NO<sub>x</sub>$  emissions without understanding the resulting engine cycle impacts and resultant  $CO<sub>2</sub>$  emissions.

There are limitations to this paper that should be expanded upon. First, the impacts of emission levels are limited to a climate analysis, rather than including air quality and additional fuel costs. There is also an assumed fuel burn increase as a result of limiting  $NO<sub>x</sub>$  emission levels. Airlines, and therefore engine manufactures, would likely not choose to increase fuel burn rates to offset  $NO<sub>x</sub>$  emissions, but would instead change the engine cycle or combustor design to meet  $NO<sub>x</sub>$  standards.

While this thesis also evaluates the  $CO<sub>2</sub>$  and NO<sub>x</sub> tradeoff for gas turbines, it specifically aims to quantify social costs associated with the current LTO  $NO<sub>x</sub>$  regulation, and is the first study to incorporate climate damages, air quality damages, and the marginal cost of fuel production into a social cost per unit thrust analysis of the current LTO  $NO<sub>x</sub>$  regulation. This thesis shows that the regulation as it is defined today allows for uneven environmental damage per unit thrust across a range of OPRs by allowing engines to do more damage as OPR is increased. Additionally, this thesis assesses alternative potential definitions of the regulation which would result in more evenly distributed social costs (i.e. environmental and fuel combined) across the engine design space.

The analysis is organized as follows. The materials and methods section assesses known emission rates, reported for uninstalled, sea level static engines in certification, and outlines techniques used to quantify unknown emission rates, primarily in cruise operation. Then, the social cost of engine operation is quantified, defined as the sum of the marginal cost of fuel production, and climate and air quality damages associated with emissions. Finally, the split between time in LTO and time in cruise are used to determine full-flight impacts.

Several sources of emissions and cost data are used for this analysis. The ICAO Aircraft Engine Emissions Databank (EDB) is used to estimate landing and take-off emissions. The Aviation Emissions Inventory Code (AEIC) and the Boeing Fuel Flow Method 2 (BFFM2) are used to estimate cruise emissions. The historical price of jet fuel and markup rates are used to estimate fuel costs. Damage estimates associated with  $NO<sub>x</sub>$  and  $CO<sub>2</sub>$  emissions are based on an analysis using the Aviation Environmental Portfolio Management Tool. Finally, an uncertainty analysis from all data sources is included in a Monte Carlo simulation, where 95% confidence intervals are presented.

Results display the allowable emissions associated with the CAEP/8 standard, and show the distribution of social costs from those emissions as a function of OPR. Future goals set by the ICAO Committee on Aviation Environmental Protection (CAEP) are also analyzed, and finally an analysis is presented to remove the social cost biases associated with OPR.

## <span id="page-18-0"></span>Materials & Methods

The main unknown parameters of this analysis are  $NO<sub>x</sub>$  and  $CO<sub>2</sub>$  emissions from current and future aircraft engine technology. There are two modes of operation where emissions must be determined, LTO and cruise, making four unknowns: LTO and cruise NO<sub>x</sub> emissions, and LTO and cruise  $CO<sub>2</sub>$  emissions. However, engine performance data is proprietary and not widely available.

Throughout this analysis all estimates of emissions for LTO operation are a result of measured data from the ICAO EDB. Since there is no equivalent database for aircraft emissions at altitude, extrapolation techniques and cycle analyses must be used to represent emissions in the space they operate most of the time. Uncertainty in each set of parameters discussed in this section is included in a Monte Carlo analysis (along with other sources of uncertainty discussed later in this paper), with an aim to evaluate the impact of the current regulation and potential future regulations that are net environmentally beneficial, accounting for uncertainty.

### <span id="page-18-1"></span>Landing and Take-Off Measured Engine Data

In the certification process for commercial aircraft engines, manufacturers are required to report emissions as measured in four different modes of operation.

- 100% available thrust for 0.7 minutes (Take-Off)
- 85% available thrust for 2.2 minutes (Climb)
- 30% available thrust for 4.0 minutes (Approach)
- 7% available thrust for 26 minutes (Taxi) (ICAO, 2008)

The results are published in the EDB, which provides measured data-points where inproduction and historical aircraft engines can be compared directly to the ICAO standards. [Figure 1](#page-19-0) shows the trend of recent engine certifications from EDB v25a in relation to the standards (ICAO, 2018). Engines with Twin Annular pre-Mixing Swirler (TAPS) or Technology for Advanced Low  $NO_x$  (TALON X) combustors in the EDB fall significantly below the standard compared to the rest of the population, and are highlighted in [Figure 1.](#page-19-0) These new combustor technologies have reduced  $NO<sub>x</sub>$  emissions without showing an increase in fuel burn.



<span id="page-19-0"></span>*Figure 1: History of NO<sup>x</sup> emissions standards per the Committee on Aircraft Engine Emissions and the Committee on Aviation Environmental Protection adopted by ICAO since 1981 compared to in-production aircraft data*

#### <span id="page-20-0"></span>Landing and Take-Off  $NO<sub>x</sub>$  Emissions

Since this analysis aims to understand the allowable emissions per the current regulation, we consider hypothetical engines where LTO  $NO<sub>x</sub>$  emissions match the level defined by the currently-applicable CAEP/8 (ICAO, 2008) regulation. This limit is defined as a function of OPR, given as

$$
\frac{D_p(NO_x)}{F_{oo}} = \begin{cases} OPR \le 30, & 7.88 + 1.4080\pi_{oo} \\ 30 < OPR \le 62.5, & -9.88 + 2\pi_{oo} \\ OPR > 104.7, & 32 + 1.6\pi_{oo} \end{cases} \tag{1}
$$

where the constants have units of g/kN. A history of each iteration of this regulation is included in Appendix A.

#### <span id="page-20-1"></span>Landing and Take-Off  $CO<sub>2</sub>$  Emissions

 $CO<sub>2</sub>$  emissions for LTO operations are calculated based on measured fuel burn from the EDB. The emissions index (EI) of a species relates fuel burn to emission level, and is defined as a constant 3.155 kg CO<sub>2</sub> per kg of jet fuel. All in-production engines in the EDB are used to determine the relationship between total fuel burn in LTO as a function of engine OPR. Each of the four modes of operation (take-off, climb, approach, and taxi) are used in a time average, based on times in mode defined in the ICAO LTO cycle.

The best fit line for this relationship is determined to be

$$
\frac{D_p(CO_2)}{F_{oo}} = 7233 + 29670 e^{-0.0711 \pi_{oo}}
$$
 (2)

where the constants have units of g/kN. The data, best fit and 95% confidence intervals are shown i[n Figure 2.](#page-21-1) In the Monte Carlo analysis, the 95% CI is used to define a triangular distribution about the best-fit line, based on residuals from the EDB data.



<span id="page-21-1"></span>*Figure 2: LTO CO<sup>2</sup> emissions as a function of reported OPR for all in-production engines in the ICAO EDB*

#### <span id="page-21-0"></span>Cruise NO<sup>x</sup> Emissions

Without measurements from the EDB, estimates on aircraft fuel flow rates in cruise conditions must be made to determine cruise emissions. In reality, fuel flow rates change throughout the flight; even cruising at a constant speed and constant altitude requires varying thrust levels based on changing aircraft weight. AEIC is a tool used to estimate aircraft operating conditions over entire flights, and takes into account varying weights, varying thrust levels and varying segments of the flight (Simone et al., 2013). Although AEIC can be used in this analysis, the intention of this work is to evaluate the impacts of a

regulation applied at engine certification, and therefore should not be flight path or aircraft specific. Therefore, other assumptions on an average and representative cruise fuel flow for each engine can be made, and the compared with AEIC as a form of validation.

Dimensional analysis is used to estimate cruise fuel flow with an energy balance. The rate of energy supplied by the fuel is represented in the numerator, while the denominator represents a rate of work (Cumpsty and Heyes, 2015). The sea level and cruise comparison is

$$
\frac{\dot{m}_f \, \text{LCV}}{\sqrt{c_p \, T_{02}} \, D^2 \, p_{02}} \bigg|_{\text{Sea Level}} = \frac{\dot{m}_f \, \text{LCV}}{\sqrt{c_p \, T_{02}} \, D^2 \, p_{02}} \bigg|_{\text{Cruise}} \tag{3}
$$

where  $D^2$  is a characteristic diameter of the engine, and the lower calorific value (LCV) and specific heat (*cp*) are characteristics of the fuel, therefore are constant between the same engine operating at two conditions. The simplified equation is

$$
\frac{\dot{m}_{f_{\text{Cruise}}}}{\dot{m}_{f_{\text{Sea Level}}}} = \frac{p_{02,\text{Cr}}}{p_{02,\text{SL}}} \sqrt{\frac{T_{02,\text{Cr}}}{T_{02,\text{SL}}}}
$$
(4)

Using the International Standard Atmosphere (ISA) at 35,000 ft to represent cruise and an assumed flight Mach number of 0.8, and using ISA at sea level and static conditions to represent sea level, isentropic flow equations are used to define the pressures and temperatures. The relationship is

$$
\dot{m}_{f_{\text{Cruise}}} = 0.33 \,\dot{m}_{f_{\text{Sea Level}}}
$$
\n(5)

The resulting ratio shows that the fuel flow at cruise can be estimated as 33% of the fuel flow at sea level at an equivalent power level. When compared to AEIC data, discussed further below, sea level climb-out (85% thrust) can be used to represent cruise power levels. A detailed outline of this approach is included in Appendix B.

Using 69 known aircraft-engine pairs (listed in Appendix C), AEIC is used to validate cruise fuel flow from above. This list was compiled from active aircraft with engines listed in the EDB per AEIC source data. To minimize the variation associated with direct AEIC output, representative cruise conditions were chosen corresponding to the 3 gross weights used in the ICAO CO<sub>2</sub> standard. This simplifies the analysis by removing varying aircraft weight and fuel burn rates based on flight paths, airport pairings, and route distances. A high gross weight, mid-weight, and low gross weight representative of cruise are calculated based on reported maximum take-off mass (MTOM) (ICCT, 2013).

$$
M_{\text{Low}} = (0.45 \times \text{MTOM}) + (0.63 \times (\text{MTOM}^{0.924})) \tag{6}
$$

$$
M_{\rm Mid} = \frac{(M_{\rm High} + M_{\rm Low})}{2} \tag{7}
$$

$$
M_{\text{High}} = 0.92 \times \text{MTOM} \tag{8}
$$

The agreement between the non-dimensional analysis and the AEIC output, shown in Appendix B, provides the basis for using a 33% scalar on sea level fuel flow to represent cruise fuel flow for the remainder of this analysis.

Once cruise fuel flow is known, the BFFM2 provides an extrapolation technique for emissions at altitude. Unlike  $CO<sub>2</sub>$ , NO<sub>x</sub> does not have a constant EI that ties it directly to fuel flow. The BFFM2 uses measured data from the EDB for each engine, applies corrections for installation and altitude effects, and uses a log-log extrapolation for determining the  $NO<sub>x</sub>$  EI (Dubois et al., 2006; Baughcum et al., 1996). This method is outlined in more detail in Appendix D. Cruise altitude is assumed to be 7,000 ft below the aircraft ceiling, the same assumption used in AEIC (Simone et al., 2013). Implementing this technique across the 69

aircraft-engine combinations shows a trend between LTO  $NO<sub>x</sub>$  emissions per rated thrust and average cruise emissions rate per rated thrust, shown in [Figure 3.](#page-24-0)



<span id="page-24-0"></span>*Figure 3: Average cruise NO<sup>x</sup> emissions per unit rated thrust as a function of LTO NO<sup>x</sup> for 69 AEIC aircraftengine pairs*

The best fit line is determined to be

$$
\frac{\text{ER}_{\text{cr}}(\text{NO}_x)}{F_{oo}} = 1.266 \times \left(\frac{D_p}{F_{oo}}\right)_{\text{LTO}}\tag{9}
$$

where the constants have units of  $g/kN/s$  and  $ER<sub>cr</sub>$  is used to represent the emissions rate (per second) in cruise operation. To capture the uncertainty with this analysis, error bands capturing the 95% confidence interval have been added around the best fit line, with  $R^2$  = 0.861. These bounds are included in the Monte Carlo analysis as a multiplied of the slope of this relationship.

#### <span id="page-25-0"></span>Cruise CO<sub>2</sub> Emissions

Although a fuel flow rate was used in the previous section to extrapolate LTO  $NO_x$ emissions to cruise, that rate is not indicative of future engine technology. Therefore, to capture current and future technology and  $CO<sub>2</sub>$  emissions, engine cycle equations and representative technology values are used to develop a range of hypothetical engines across the engine design space.

Using gas turbine principles, an expression for net power can be used with isentropic relationships in the compressor and turbine to derive an expression for cycle efficiency with four assumptions. This analysis is outlined in Cumpsty and Heyes (2015), with a high-level review included here and more detailed analysis in Appendix E.

First, the net power is defined as the difference between power produced in the turbine and power used to drive the compressor. This simplified representation assumes net power is used for propulsion. The power supplied by an ideal compressor and produced by an ideal turbine is represented by the mass flow, temperature change across the component, and the specific heat of air. Then, an assumption for efficiency and isentropic equations can be substituted to represent real work.

Combining the definitions for turbine and compressor work yields an expression for cycle specific work. Simplified, it includes only four parameters: compressor efficiency, turbine efficiency, pressure ratio, and operating temperature ratio. Temperature ratio is defined as the ratio between the combustor exit temperature, the hottest point in the engine cycle, and the static inlet temperature. Using the same pressure ratio for both the

compressor and turbine expressions assumes a negligible pressure drop across the combustor.

Specific work is then used to find an expression for cycle efficiency, represented as the ratio between the amount of power produced by the cycle (output), and the work done in the combustor (input), i.e.

$$
\eta_{\text{cycle}} = \left(\eta_{\text{turb}} \frac{T_4}{T_2} \left(1 - \frac{1}{p_R (r - \frac{1}{\gamma})}\right) - \frac{p_R (r - \frac{1}{\gamma})}{\eta_{\text{comp}}}\right) \left(\frac{T_4}{T_2} - \left(1 + \frac{p_R r - \frac{1}{\gamma}}{\eta_{\text{comp}}}\right)\right)^{-1} (10)
$$

Thrust can be calculated from the fuel flow, overall efficiency (combining cycle and propulsive efficiency), velocity and fuel heating value as

$$
F = \frac{\eta_{\text{overall}} W_f \text{ LHV}}{\text{Velocity}} \tag{11}
$$

Rated thrust can be assumed from a known lapse rate in air density. Finally, rated thrust and the emissions index of  $CO<sub>2</sub>$  can be used to estimate cruise emissions per rated thrust per second of flight without an explicit assumption on fuel flow, defined as

$$
\frac{\text{ER}_{\text{cr}}(\text{CO}_2)}{F_{oo}} = \frac{\text{EI}_{\text{CO}_2}}{\text{LR}[(\eta_{\text{cycle}} \times \eta_{\text{prop}}) \text{LHV/ Velocity}]}
$$
(12)

To represent current engine technology and future engine technology, a range of each of the above assumptions are made and listed in Table 1. These ranges are chosen based on representative values with input from industry experts. These values are used in the Monte Carlo analysis to estimate cruise  $CO<sub>2</sub>$  emissions as a function of OPR.

[Figure 4](#page-27-1) shows the 95% confidence interval for  $CO<sub>2</sub>$  emissions at cruise based on the assumptions for component efficiencies, temperature ratio, and propulsive efficiency. Additionally, for validation, output from representative engine cycle models provided by a major gas turbine manufacture have been included. The red data represents 1990's technology, while the blue data represents a 2010's engine.



<span id="page-27-1"></span>*Figure 4: Cruise CO<sup>2</sup> emissions as a function of OPR with two representative cycle models at three high power settings included for reference*

#### <span id="page-27-0"></span>Climate and Air Quality Environmental Damages

Once all four unknown emission rates are estimated across a range of OPRs, the impacts of those emissions are quantified. In this analysis, the social cost of engine operation considers the marginal cost of fuel production, and climate and air quality damages associated with emissions.

RF impacts from  $NO_x$  are attributed to four pathways: short lived  $O_3$  production, long term O<sub>3</sub> depletion, long term CH<sub>4</sub> depletion, and nitrate aerosol formation. Additionally, NO<sub>x</sub> contributes to local air quality damage via  $PM_{2.5}$  and  $O_3$  production.  $CO_2$  is a greenhouse gas and therefore directly impacts climate and RF, but is not considered in our air quality analysis.

Grobler et al. (2019) describes the pathways, climate and air quality impacts for each emissions species, and monetizes those impacts with uncertainty bounds. Each damage value is presented in US dollars per tonne of emissions species. Appendix F contains values and 95% confidence intervals for each pathway, for multiple discount rates. Including only these impacts provides an environmental cost basis for regulating aviation  $NO<sub>x</sub>$  emissions, but a further social cost is also considered.

#### <span id="page-28-0"></span>Marginal Production Cost of Fuel

Fuel consumption is also included in quantifying the social costs of engine operation. While the market price of jet fuel represents a transfer that does not constitute a social cost, the production of fuel consumes resources that could be expended elsewhere. Therefore, the marginal cost of jet fuel production is included in the social cost metric, which is computed from the market price of jet fuel and markup rates from the literature. Khan et al. (2013) showed oil price markup between the years 1980 and 2010 ranged from 4.45 to 4.75. Considine (2001) presents a Lerner Index of 62.3%, 43.3%, and 45.9% for jet fuel, which is equivalent to a 2.65, 1.76, and 1.84 markup respectively.

Using historical data from ThomsonOne (JETA Y-IL) to find the price of jet fuel for 2013- 2018, and the markup rates from above, the marginal production cost of a unit of jet fuel can be estimated for a 5 year period. The Lerner Index and historical price of Jet A are described in more detail in Appendix G.

#### <span id="page-29-0"></span>Costs per Unit Operating Time

A weighted-average of emission rates is found based on the time a flight spends between cruise and LTO operation. Short-haul flights have a higher percentage of LTO operation, while long-haul flights are heavily weighted towards cruise.

Examining the global 2013 AEIC flight inventory, the average cruise operation was found to account for 73.2% of total flight time, with a minimum of 9.1% and a maximum of 95.2%. This is based on an LTO cycle of 2,897 s (0.8 hr) as outlined in Settler et al. (2011), which matches the assumption used while developing AEIC. This is considered more representative of actual operations than the ICAO cycle. The distribution of flight times is included in Appendix H for reference.

Finally, to calculate the LTO and cruise operation costs, the following equations are used with emissions per rated thrust per second of operation, the marginal cost of fuel, and the damage distributions from Grobler et al. (2019).

Nominal values for each of the above analyses are used to calculate a set of nominal results, and are listed in Table 1. Combined social costs are presented as a function of OPR, and are discussed in the first section of results.

Next, Sobol sequences are used to combine all the previous uncertainty, which mathematically sample the desired distribution shapes uniformly (Sobol et al., 2001). A Monte Carlo analysis with 10,000 cases is performed, and used to evaluate the allowable social costs per the current CAEP/8 limit. Distributions from individual emission sources are shown in Appendix I, while the distributions from environmental damage pathways are shown in Appendix J.

Finally, alternate  $NO<sub>x</sub>$  limits are evaluated and compared to the current regulation. Specifically, the mid- and long-term goals set by ICAO's Working Group 3, and a case where social costs are constant across the OPR range. These studies use the 10,000 Monte Carlo runs.

<span id="page-31-0"></span>

## *Table 1: Summary of Nominal Values and Monte Carlo Distributions*

Note: L, M and U represent the lower, middle and upper characteristics for a triangular distribution.

## <span id="page-32-0"></span>**Results**

First, results from the nominal analysis are shown, followed by a Monte Carlo analysis.

#### <span id="page-32-1"></span>Costs per Unit Operating Time: Nominal

Using nominal values from Table 1[, Figure 5](#page-32-2) shows the tradeoff between the  $NO<sub>x</sub>$  and  $CO<sub>2</sub>$ emissions rate per rated thrust from cruise and LTO operation as a function of OPR, with contours of constant social cost. This shows that as OPR is increased, the social cost decreases initially, reaches a minimum, and then begins to climb again. For consistency, the difference between the constant social cost contours in LTO and cruise is the same magnitude (5x10<sup>-4</sup> \$/s/kN), but have different slopes because the two modes of operation have different damages associated with the emissions. All values shown in these results reflect a discount rate of 3%.



<span id="page-32-2"></span>*Figure 5: NO<sup>x</sup> and CO<sup>2</sup> emissions for a range of OPRs (10 to 70) with lines of constant social costs in cruise and LTO respectively for reference*

[Figure 6](#page-34-1) shows the resulting damages associated with climate, air quality, and marginal cost of fuel broken down into their respective contributions. The shape of the overall social cost displays a minimum point that occurs approximately at an OPR of 20, and biases at the high and low OPRs where social costs are higher. This implies the regulation as it is defined today allows for an uneven distribution of social cost across the design space. Climate damages and fuel costs make-up the majority of combined social cost at lower OPR, while air quality damage from  $NO<sub>x</sub>$  makes-up the majority at higher OPR. This trend is especially important to understand as newer engines move to higher OPR values.

As OPR is increased, near-constant values for the climate and fuel cost terms reflect the diminishing returns in efficiency, while the CAEP/8 regulation shape allows the contributions from air quality to climb. The near-constant fuel cost contribution also demonstrates the trends from an environmental cost basis (with climate and air quality contributions only) and a social cost basis are similar.

This analysis and the emissions tradeoffs shown are not necessarily indicative of thermodynamic tradeoffs in the physics of gas turbine engines; this is only an analysis on allowable emissions per the currently-applicable CAEP/8 standard. For example, higher  $NO<sub>x</sub>$ emissions shown here do not require higher temperatures; they are a product of the regulation being defined as a function of OPR. Furthermore, the minimum in damages at an OPR of 20 does not imply that this is an environmentally optimal engine design for all engines; it is a product of the regulation as it is currently defined.



<span id="page-34-1"></span>*Figure 6: Nominal social cost analysis broken into marginal production cost of fuel, climate damage and air quality damage*

#### <span id="page-34-0"></span>Costs per Unit Operating Time: Monte Carlo

The distribution of total social costs per second of operation for operations at the level of the CAEP/8 standard is shown in [Figure 7](#page-35-0) with the median, 50%, and 95% confidence intervals drawn. These results also show that there is an uneven distribution of damage over the engine design space, where going to higher OPRs allows total environmental damage to increase. The distribution of OPRs at which minimum damage occurs is shown in [Figure 8.](#page-36-0)



<span id="page-35-0"></span>*Figure 7: Distribution of combined environmental and social costs per rated thrust per second of operation, allowable per the CAEP/8 limit*

A breakdown of [Figure 7](#page-35-0) into climate damage, air quality damage, and the marginal cost of fuel as a function of OPR for LTO and cruise separately is shown in Appendix K. Climate damages and fuel costs make up the majority of combined social cost at lower OPR, while air quality damage from  $NO<sub>x</sub>$  make up the majority at higher OPR.

A sensitivity analysis is also conducted for these results, breaking out the uncertainty contributions from each of the inputs listed in Table 1. The distribution of monetized environmental damages associated with emission levels per Grobler et al. (2019) account for ~80% of the uncertainty at an OPR of 40. Full results are listed in Appendix L.
[Figure 8](#page-36-0) shows the empirical cumulative distribution function at which OPR of minimum social costs occur for each of the 10,000 Monte Carlo simulations. This figure shows 50% of simulations have social cost climbing by an OPR of 22, and 90% of Monte Carlo cases have social cost climbing by an OPR of 36. These results show that new engines, which are higher OPR designs, have higher allowable social costs than previous lower OPR engine designs, if they operate near the CAEP/8  $NO<sub>x</sub>$  limit.



<span id="page-36-0"></span>*Figure 8: Distribution of OPR associated with minimum social cost showing a bias towards low OPR engine design and an OPR of 30, where the current regulation changes slope*

With 10,000 runs in the Monte Carlo analysis, a new trend also becomes obvious: as a result of the regulation changing slope at an OPR of 30, there is a biased minimum damage point. This can be seen in [Figure 8](#page-36-0) as a step change at an OPR of 30.

In 2007, ICAO also announced mid- and long-term goals for  $NO<sub>x</sub>$  emissions, for the years 2016 and 2026 respectively. These are shown on [Figure 9](#page-37-0) as dotted lines, and are defined as a 45% and 60% reduction at an OPR of 30 from the CAEP/6 standards (Dickson, 2015).



<span id="page-37-0"></span>*Figure 9: Comparison of historical ICAO standards, ICAO goals, and constant social cost contours*

For comparison, the total allowable social costs from hypothetical engines operating at those emission levels are displayed in [Figure 10,](#page-38-0) and show a more even distribution of social costs across engine OPR on the same cost-scale as compared to [Figure 7.](#page-35-0) CAEP/8 results are shown in grey with dashed lines in the background for reference, and results from the midand long-term goals are shown in teal.



<span id="page-38-0"></span>*Figure 10: Social costs of NO<sup>x</sup> emissions at the levels of the ICAO mid-term and long-term goals, with CAEP/8 results shown in the background for reference*

Finally, instead of assuming new definitions of the standard will be based on previous versions, an alternative approach to defining the  $NO<sub>x</sub>$  standard is evaluated in an effort to reduce the trends in total allowable damage. The solid black lines shown in [Figure 9](#page-37-0) represent lines of constant social cost, and are defined with the goal of removing the OPR bias in the 50<sup>th</sup> percentile of the Monte Carlo runs.

Comparing the results from the ideal constant social cost analysis to the mid- and long-term goals set by CAEP shows the benefit of moving towards those goals from the current definition. By scaling previous versions of the regulation down at all OPRs, instead of a constant reduction, the social cost is more even across the design space.

The overall social cost for the mid-contour shown in [Figure 9](#page-37-0) is displayed in [Figure 11.](#page-39-0) The 50<sup>th</sup> percentile range is now constant across OPR, and when compared to the original results, the absolute range of damages is smaller than before.



<span id="page-39-0"></span>*Figure 11: Total allowable combined environmental and social cost from a constant social cost contour*

## **Conclusions**

Using the environmental or social cost basis derived in this thesis for  $NO<sub>x</sub>$  regulations would encourage engine manufacturers to employ technology to reduce  $NO<sub>x</sub>$  emissions. A  $NO<sub>x</sub>$ standard defined on this basis, with combustor technology that can achieve a set level of  $NO<sub>x</sub>$ emissions as a function of OPR, would yield an intersection of two curves that defines the maximum OPR that could be used with the given technology while meeting the standard. If combustor technology is improved to further reduce  $NO<sub>x</sub>$  emissions, this intersection will move to higher OPR, allowing an increase in engine efficiency while holding social costs constant. Thus, a  $NO<sub>x</sub>$  standard designed in this way provides incentives for development of improved technologies for minimizing NO<sub>x</sub> emissions, since doing so would allow manufacturers to build more fuel-efficient engines which are also more attractive to their customers.

This research could be taken a step further to suggest  $NO<sub>x</sub>$  regulations should not be defined as a function of OPR at all, but instead defined as a function of SFC. Since OPR is used as a way to determine  $CO<sub>2</sub>$  emissions in this work, it makes sense that  $CO<sub>2</sub>$  emissions (and therefore fuel flow rates) could be used directly.

A challenge of using a social cost analysis for actually setting regulatory limits is that the inputs are subject to change over time. Fuel costs can change rapidly with market fluctuations, new engine technology is developed, and impacts from emissions change as background levels change. Considering the multi-decade lifespan of an aircraft engine certified for a particular  $NO<sub>x</sub>$ standard, the social cost estimates applicable to that engine should be developed with damages applicable to these future times in mind.

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Another aspect of this research with high uncertainty is how to monetize damages. All primary results are shown with a discount rate of  $3\%$ . Since CO<sub>2</sub> is a long-term greenhouse gas, low discount rates increase the relative contribution of  $CO<sub>2</sub>$  damage in this analysis, while short term air quality impacts are valued more highly with high discount rates. Results for 2% and 7% discount rates are included in Appendices N and O, respectively. Results show that higher discount rates diminish the overall costs, but lower discount rates are more constant across the range of OPRs from reducing the  $NO<sub>x</sub>$  air quality damage that dominates at high OPR.

This research could also be improved with more reliable emissions data at altitude. Several campaigns have been performed to capture emissions data, but the uncertainty levels are too high to use here. Additionally, these campaigns have been done on only a handful of aircraft, and therefore cannot be used to represent the world-wide fleet. Without cruise emission data widely available this analysis relies upon several different calculation and extrapolation techniques. One solution would be to require manufactures to report cruise-equivalent points in certification (like the EDB exists for LTO). Research shows cruise operation makes up the majority of aviation related air quality and climate impacts, therefore understanding those emission rates is critical to being able to evaluate impacts accurately.

Despite these unknowns and challenges, this thesis demonstrates a new way to evaluate current and proposed regulations, by demonstrating why a holistic approach is needed to fully understand and quantify the environmental impacts associated with regulations.

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## Appendices

#### A. History of CAEP NO<sub>x</sub> Regulations

Per Annex 16 to the Convention on International Civil Aviation, Volume 2 Aircraft Engine Emissions, the following emissions standards for  $NO<sub>x</sub>$  apply to subsonic commercial aircraft.

 For engines of a type or model for which the date of manufacture of the first individual production model was before 1 January 1996 and for which the date of manufacture of the individual engine was before 1 January 2000:

$$
\frac{D_p}{F_{oo}} = 40 + 2\pi_{oo}
$$

 For engines of a type or model for which the date of manufacture of the first individual production model was on or after 1 January 1996 or for which the date of manufacture of the individual engine was on or after 1 January 2000:

$$
\frac{D_p}{F_{oo}} = 32 + 1.6\pi_{oo}
$$

- For engines of a type or model for which the date of manufacture of the first individual production model was on or after 1 January 2004:
- $\circ$  For  $F_{oo} > 89.0 kN$

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \text{OPR} \le 30, & 19 + 1.6\pi_{oo} \\ 30 < \text{OPR} \le 62.5, & 7 + 2\pi_{oo} \\ \text{OPR} > 62.5, & 32 + 1.6\pi_{oo} \end{cases}
$$

 $\circ$  For 26.7 kN  $\lt F_{oo} \lt 89.0$  kN

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \text{OPR} \le 30, & 37.572 + 1.6\pi_{oo} - 0.2087F_{00} \\ 30 < \text{OPR} \le 62.5, \\ \text{OPR} > 62.5, \end{cases}
$$
\n
$$
42.71 + 1.4286\pi_{oo} - 0.4013F_{00} + 0.00642\pi_{oo}xF_{00} \\ 32 + 1.6\pi_{oo}
$$

- For engines of a type or model for which the date of manufacture of the first individual production model was on or after 1 January 2008 or for which the date of manufacture of the individual engine was on or after 1 January 2013:
- $\circ$  For  $F_{oo} > 89.0 kN$

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \text{OPR} \le 30, & 16.72 + 1.4080\pi_{oo} \\ 30 < \text{OPR} \le 62.5, & -1.04 + 2\pi_{oo} \\ \text{OPR} > 82.6, & 32 + 1.6\pi_{oo} \end{cases}
$$

 $\circ$  For 26.7 kN  $\lt F_{oo} \lt 89.0$  kN

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \n\text{OPR} \le 30, & 38.5486 + 1.6823\pi_{oo} - 0.2453F_{00} - 0.00308\pi_{oo} xF_{00} \\
30 < \text{OPR} \le 62.5, & 46.16 + 1.4286\pi_{oo} - 0.5303F_{00} + 0.00642\pi_{oo} xF_{00} \\
\text{OPR} > 82.6, & 32 + 1.6\pi_{oo} \n\end{cases}
$$

- For engines of a type or model for which the date of manufacture of the first individual production model was on or after 1 January 2014:
- $\circ$  For  $F_{oo} > 89.0 kN$

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \text{OPR} \le 30, & 7.88 + 1.4080\pi_{oo} \\ 30 < \text{OPR} \le 62.5, & -9.88 + 2\pi_{oo} \\ \text{OPR} > 104.7, & 32 + 1.6\pi_{oo} \end{cases}
$$

 $\circ$  For 26.7 kN  $\lt F_{oo} \lt 89.0$  kN

$$
\frac{D_p}{F_{\infty}} = \begin{cases} \n\text{OPR} \le 30, & 40.052 + 1.5681\pi_{oo} - 0.3615F_{00} - 0.0018\pi_{oo}xF_{00} \\
30 < \text{OPR} \le 62.5, & 41.9435 + 1.505\pi_{oo} - 0.5823F_{00} + 0.005562\pi_{oo}xF_{00} \\
\text{OPR} > 104.7, & 32 + 1.6\pi_{oo} \n\end{cases}
$$

### B. Dimensional Analysis for Estimating Cruise Fuel Flow

A dimensional analysis is used to find the relationship between fuel flows for two flight conditions from the same engine, in this case cruise and LTO. From Cumpsty and Heyes (2015), the relationship is

$$
\left. \frac{\dot{m}_f \, \text{LCV}}{\sqrt{c_p \, T_{02} \, D^2 \, p_{02}}} \right|_{\text{Sea Level}} = \left. \frac{\dot{m}_f \, \text{LCV}}{\sqrt{c_p \, T_{02} \, D^2 \, p_{02}}} \right|_{\text{Cruise}}
$$

where

Non – Dimensional Fuel Flow Rate = 
$$
\frac{\dot{m}_f \, \text{LCV}}{\sqrt{c_p \, T_{02}} \, D^2 \, p_{02}}
$$

and *D 2* (diameter) is a characteristic of the engine, and LCV (lower calorific value of fuel) and *c<sup>p</sup>* (specific heat at constant pressure) are characteristics of the fuel, therefore are constant between two engine operating conditions. The simplified equation is

$$
\left. \frac{\dot{m}_f}{\sqrt{T_{02}} \ p_{02}} \right|_{\text{Sea Level}} = \left. \frac{\dot{m}_f}{\sqrt{T_{02}} \ p_{02}} \right|_{\text{Cruise}}
$$

Ambient conditions are assumed to match atmospheric standard day, shown below.

Sea Level:

- $\bullet$  Ts = 288.15 K
- $PS = 101,325$  Pa
- $•$  Mach = 0.0
- Cruise, 35,000 ft:
	- $Ts = 218.8 K$
	- $PS = 23,800$  Pa
		- $\bullet$  Mach = 0.8

Using Mach number to get stagnation properties from the above static assumptions, and rearranging the equation above, a ratio between cruise and LTO conditions is expressed as

$$
\frac{\dot{m}_{f_{\text{Cruise}}}}{\dot{m}_{f_{\text{Sea Level}}}} = \frac{p_{02,\text{Cr}}}{p_{02,\text{TO}}} \sqrt{\frac{T_{02,\text{Cr}}}{T_{02,\text{TO}}}}
$$

where pressures are in kPa and temperatures are in R. The resulting ratio is

$$
\dot{m}_{f_{\text{Cruise}}} = 0.33 \dot{m}_{f_{\text{Sea Level}}}
$$

Climb-out (85% thrust) from the EDB is used to represent an equivalent power level at cruise, and therefore 33% of reported fuel flow in the EDB is used as cruise fuel flow. It is compared to fuel flow reported by AEIC in [Figure A-12](#page-47-0) below.



<span id="page-47-0"></span>*Figure A-12: Cruise fuel flow as estimated with dimensional analysis compared to reported fuel flow from AEIC, with an R<sup>2</sup> value of 0.9615*

# C. AEIC Aircraft-Engine List



# *Table A-2: AEIC Aircraft-Engine Pairs and EDB Matches*





### D. Summary of Boeing Fuel Flow Method 2

An outline of the Boeing Fuel Flow Method 2 (Dubois et al., 2006; Baughcum et al., 1996) as applied to extrapolating  $NO<sub>x</sub>$  emissions from LTO to cruise conditions is included below. A more detailed analysis can be found in NASA's Scheduled Civil Aircraft Emission Inventories for 1992, Appendix D or Boeing's "Fuel Flow Method2" for Estimating Aircraft Emissions.

First, apply corrections for installation effects to the measured fuel flow reported in the EDB.

<b>Power Setting</b>	Correction
Take-off	1.010
Climb Out	1.013
Approach	1.020
Idle	1.100

*Table A-3: BFFM2 Installation Corrections*

On a log-log scale, apply a linear fit through the  $NO<sub>x</sub>$  Emissions Indices reported in the EDB vs the corrected fuel flows from above.

Calculate a Fuel Flow Factor,  $W_{\text{ff}}$ , using delta and theta corrections for temperature and

pressure at altitude, and known cruise fuel flow using

$$
W_{ff} = \frac{W_f}{\delta_{\rm amb}} \theta_{\rm amb}^{3.8} e^{0.2 M^2}
$$

where theta is defined as

$$
\Theta_{\rm amb} = \frac{T_{\rm amb} + 273.15}{288.15}
$$

and delta is defined as

$$
\delta_{\rm amb} = \frac{P_{\rm amb}}{14.696}
$$

Next, calculate the new cruise NO<sub>x</sub> emissions index,  $(EINO_x)_{W_{ff}}$ , from W<sub>ff</sub> and the log-log linear fit, then apply the following relationships and corrections

$$
\text{EINO}_\text{x} = (\text{EINO}_\text{x})_{W_{ff}} e^H \left(\frac{\delta_{\text{amb}}^{1.02}}{\theta_{\text{amb}}^{3.3}}\right)^{0.5}
$$

where

$$
H = -19.0(\omega - 0.0063, \quad \omega = \frac{0.62198(\Phi)P_v}{P_{amb} - (\Phi)P_v}, \quad P_v = (0.14504)10^{\beta}
$$

and

$$
\beta = 7.90298 \left( 1 - \frac{373.16}{T_{\text{amb}} + 273.16} \right) + 3.00571 + (5.02808) \log \left( \frac{373.16}{T_{\text{amb}} + 273.16} \right) + (1.3816 * 10^{-7}) \left[ 1 - 10^{11.344 \left( 1 - \frac{T_{\text{amb}} + 273.16}{373.16} \right)} \right] + (8.1328 * 10^{-3}) \left[ 10^{3.49149 \left( 1 - \frac{373.16}{T_{\text{amb}} + 273.16} \right)} - 1 \right]
$$

A visual representation of this method is included below for reference, where the blue circles are the original EDB values, the pink shows the fuel flow factor look-up on the log-log fit, and the green shows the final Emissions Index.



*Figure A-13: Example of BFFM2 Compared to EDB Data for B744 Aircraft*

#### E. Cycle Efficiency Equations

Cycle equations used to represent cycle efficiency, and thrust as outlined in Cumpsty and Heyes (2015) are included here. First, net power is defined as the difference in work done in the turbine and work done in the compressor, which assumes net power is used for propulsion.

$$
\dot{W}_{\text{net}} = \dot{W}_{\text{turb}} - \dot{W}_{\text{comp}}
$$

Ideal work across the compressor is defined as

$$
\dot{W}_{\text{comp}} = \dot{m}_{\text{air}} C_p (T_3 - T_2)
$$

where  $\dot{m}_{air}$  is the mass flow rate,  $c_p$  is the specific heat of air, and the temperature change across the compressor is represented by  $T_3$  and  $T_2$ . Isentropic temperature change can be represented with pressure ratio for an adiabatic process, and real work is represented by

$$
\dot{W}_{\text{comp}} = \frac{\dot{m}_{\text{air}} C_p T_2 \left( \text{PR} \left( \frac{\gamma - \frac{1}{\gamma}}{1} \right) - 1 \right)}{\eta_{\text{comp}}}
$$

which incorporates a compressor efficiency.

The ideal work across the turbine is defined as

$$
\dot{W}_{\text{turb}} = \dot{m}_{\text{air}} C_p (T_4 - T_5)
$$

with all the same definitions as above. Again, isentropic temperature change can be represented with pressure ratio for an adiabatic process, and real work is represented by

$$
\dot{W}_{\text{turb}} = \eta_{\text{turb}} \dot{m}_{\text{air}} C_p T_4 \left( 1 - \text{PR}^{-\left(\gamma - \frac{1}{\gamma}\right)} \right)
$$

which incorporates turbine efficiency.

Going back to the original definition for net work, and combining the two previous expressions for real work, yields

$$
\frac{\dot{W}_{\text{net}}}{\dot{m}_{\text{air}}C_pT_2} = \left(\eta_{\text{turb}}\frac{T_4}{T_2}\left(1 - \frac{1}{PR}\left(\nu - \frac{1}{pR}\right)\right) - \frac{PR}\left(\nu - \frac{1}{\nu}\right) - 1}{\eta_{\text{comp}}}\right)
$$

where specific work is only a function of four assumptions.

Finally, the cycle efficiency can be defined as

$$
\eta_{\text{cycle}} = \frac{\dot{W}_{\text{net}}}{\dot{m}_{\text{air}}C_p(T_4 - T_3)}
$$

where the numerator represents the net power out and the denominator represents power from the combustor.

Replacing the  $T_3$  terms with

$$
(T_4 - T_3) = T_2 \left( \frac{T_4}{T_2} - \frac{T_3}{T_2} \right)
$$

and

$$
\frac{T_3}{T_2} = 1 + \frac{\text{PR}^{\gamma - 1/\gamma} - 1}{\eta_{\text{comp}}}
$$

eliminates one unknown in the expression for cycle efficiency. The others are eliminated by substituting the specific work term already derived, and cycle efficiency can be represented as

$$
\eta_{\text{cycle}} = \left(\eta_{\text{turb}} \frac{T_4}{T_2} \left(1 - \frac{1}{PR^{\left(\gamma - \frac{1}{\gamma}\right)}}\right) - \frac{PR^{\left(\gamma - \frac{1}{\gamma}\right)} - 1}{\eta_{\text{comp}}}\right) * \frac{1}{\left(\frac{T_4}{T_2} - \left(1 + \frac{PR^{\gamma - \frac{1}{\gamma}} - 1}{\eta_{\text{comp}}}\right)\right)}
$$

based on the same four assumptions of temperature ratio, compressor efficiency, turbine efficiency and propulsive efficiency.

Thrust is calculated as

$$
Thrust = \frac{\eta_{\text{overall}} * W_f * LHV}{Velocity}
$$

using an overall efficiency, which includes both cycle and propulsive, a fuel flow rate, the lower heating value of the fuel, and the velocity.

Next, the  $CO<sub>2</sub>$  emission rate at cruise per unit thrust is defined as

$$
CO2 \text{Emissions Rate per Unit Thrust} = \frac{EI_{CO2}}{(\eta_{\text{cycle}} * \eta_{\text{prop}}) * LHV * \frac{1}{\text{Velocity}}}
$$

using the terms for cycle efficiency and thrust from above.

Finally, a lapse rate between cruise thrust and rated thrust can be estimated as the change in air density. Although more complex assumptions can be made, such as including effects of by-pass ratio on lapse rate, this simple relationship can be used when little information is available.

[Figure A-14](#page-57-0) shows the results of this analysis using 10,000 Monte Carlo runs and a range of assumptions listed in Table 1. For comparison, two representative engine models from a major gas turbine manufacture have been included. The red data represents 1990's technology, while the blue data represents a 2010's engine.



<span id="page-57-0"></span>*Figure A-14: Cycle efficiency vs OPR with representative cycle model output for comparison*

# F. Climate and Air Quality Damages

## *Table A-4: Cruise Results with Uncertainty (Grobler et al. 2019)*



## *Table A-5: Landing and Take-off Results with Uncertainty (Grobler et al. 2019)*



### G. Lerner Index and Markups

The Lerner Index is a measure of market power, between 0 and 1, where higher values represent greater market power. It is a function of the market price of an item, and the marginal cost of production. This relationship can be rearranged to solve for the markup of a product, defined as the marginal cost over the market price.

$$
L = \frac{P - MC}{P} \rightarrow \frac{MC}{P} = (1 - L)
$$

Using 4.75 (Khan et al., 2013) and 1.76 (Considine, 2001) as bounds, a range of marginal costs are drawn compared to the market price of Jet A from ThomsonOne.



*Figure A-15: The market price and marginal cost of jet fuel between 2013 and 2018*

[Figure A-16](#page-60-0) and [Figure A-17](#page-60-1) show the distribution of costs per the data above, and the Monte Carlo draws used to represent that data.



<span id="page-60-0"></span>*Figure A-16: Distribution of marginal cost of jet fuel production from 2013-2018*



<span id="page-60-1"></span>*Figure A-17: Monte Carlo draws used to represent the marginal cost of jet fuel from 2013-2018*

## H. Time in Cruise for 2013 Flights

AEIC is used to find a distribution of time in cruise for all flights in 2013, the last reference year available in AEIC V3.0. The time in cruise is then combined with a constant assumed time in LTO of 2,897s (0.8 hr) to get the time-split between LTO and cruise used in this analysis.



*Figure A-18: Distribution of time in non-LTO operation for 2013 flights per AEIC*

A triangular fit was used in the Monte Carlo draws to represent this data, shown in Figure A-19.



*Figure A-19: Monte Carlo draws used to represent the time spent in cruise operation*

## I. Distributions of Monte Carlo Inputs

This section reviews Monte Carlo draws for non-uniform assumptions listed throughout the analysis. [Figure A-20](#page-63-0) includes three parts: the original figure for average cruise  $NO<sub>x</sub>$ emissions (also [Figure 3\)](#page-24-0), the distribution of residuals from the data to the best fit line (lower left), and the Monte Carlo draws used for the 10,000 simulations (lower right).



<span id="page-63-0"></span>*Figure A-20: Comparison of actual distribution to Monte Carlo draws for NO<sup>x</sup> Cruise Emissions*

[Figure A-21](#page-64-0) includes three parts: the original figure for average LTO  $CO<sub>2</sub>$  emissions (also [Figure 2\)](#page-21-0), the distribution of residuals from the data to the best fit line (lower left), and the Monte Carlo draws used for the 10,000 simulations (lower right).



<span id="page-64-0"></span>*Figure A-21: Comparison of actual distribution to Monte Carlo draws for CO<sup>2</sup> LTO Emissions*

## J. Distribution of Damage Functions

The draws used in the Monte Carlo analysis for environmental damages are from a 100,000 sample run per Grobler et al. (2019). The distributions are included in Figures A-22 through A-27, from which the 10,000 draws were picked.



*Figure A-22: Distribution of Air Quality Damage from NO<sup>x</sup> Emissions at Cruise*



*Figure A-23: Distribution of Air Quality Damage from NO<sup>x</sup> Emissions in LTO*



*Figure A-24: Distribution of Climate Damage from NOx Emissions in Cruise*



*Figure A-25: Distribution of Climate Damage from NOx Emissions in LTO Operation*



*Figure A-26: Distribution of Climate Damage from CO<sup>2</sup> Emissions in Cruise*



*Figure A-27: Distribution of Climate Damage from CO<sup>2</sup> Emissions in LTO Operation*

## K. Fuel Costs, Climate Damage and Air Quality Damage for DR = 3.0%

For the results presented for the Monte Carlo analysis, and additional breakdown of LTO and cruise contributions are included below in Figures A-28 through A-30 for reference. The cruise contributions are higher than LTO because average cruise fuel flow rates are higher.



*Figure A*-*28: Marginal Cost of Fuel per second of LTO and Cruise Operation*



*Figure A*-*29: Climate Damage per second of LTO and Cruise Operation*



*Figure A*-*30: Air Quality Damage per second of LTO and Cruise Operation*

### L. Sensitivity Analysis for Discount Rate = 3.0%

First-order and total-effect sensitivity are computed to determine the impacts of each unknown parameter, using the approach outlined below. Results show 80% of the uncertainty is associated with the monetized damage values from with emission levels, followed by 15% from the market price of jet fuel. This shows the analysis is stable with respect to emission level estimates, but widely impacted by the cost functions associated with those emissions. The equations for first-order and total-effect sensitivity are

$$
S_i = \frac{\left(\frac{1}{N}\right)\sum_{j=1}^{N} y_A^{(j)} y_{C_i}^{(j)} - f_0^2}{\left(\frac{1}{N}\right)\sum_{j=1}^{N} \left(y_A^{(j)}\right)^2 - f_0^2}
$$

and

$$
S_{T_i} = 1 - \frac{\left(\frac{1}{N}\right)\sum_{j=1}^{N} y_B^{(j)} y_{C_i}^{(j)} - f_0^2}{\left(\frac{1}{N}\right)\sum_{j=1}^{N} \left(y_A^{(j)}\right)^2 - f_0^2}
$$

respectively, where  $y_A$ ,  $y_B$  and  $y_C$  are matrices of results, in this case social costs associated with different Monte Carlo runs.  $y_A$  and  $y_B$  are results from independent Sobol draws, and  $y_C$  are the results from replacing once variable at a time from  $y_A$  into  $y_B$ .




Results are also shown in the figures below, across all OPR ranges. It is easy to see here that almost all (>95%) of uncertainty is from monetizing damages, and less is from the emission calculations. As seen in the results, the production cost of fuel (in blue) is a high contributor at lower OPR, but air quality (combined here with climate) dominates at high OPR.



*Figure A-31: First-orders sensitivity analysis for quantifying contributors to uncertainty* 



*Figure A-32: Total-effect sensitivity analysis for quantifying contributors to uncertainty*

## M. Social Costs for Discount Rate = 2.0%

Finally, the same results discussed in the main body of text are shown again here, with a discount rate of 2.0% instead of 3.0%. All charts keep the same scale as the original results, showing a 2.0% discount rate yields overall higher social costs than 3.0%.



*Figure A-33: Combined social cost analysis for a discount rate of 2.0%*

Also included is the range of OPR corresponding to a minimum social cost in the 10,000 Monte Carlo runs. Here, 50% of results show social costs increasing by an OPR of 25 (as compared to 23) and 90% increasing by 42 (as compared to 36).



*Figure A-34: Distribution of OPR for minimum social cost for a discount rate of 2.0%*

## N. Social Costs for Discount Rate = 7.0%

The same results discussed in the main body of text are shown again here, with a discount rate of 7.0% instead of 3.0%. All charts keep the same scale as the original results, showing a 7.0% discount rate yields overall lower social costs than 3.0%.



*Figure A-35: Combined social cost analysis for a discount rate of 7.0%*

Also included is the range of OPR corresponding to a minimum social cost in the 10,000 Monte Carlo runs. Here, 50% of results show social costs increasing by an OPR of 21 (as compared to 23) and 90% increasing by 32 (as compared to 36).



*Figure A-36: Distribution of OPR for minimum social cost for a discount rate of 7.0%*