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Optimization of Fuel Injection in GDI Engine Using Economic Order Quantity and Lambert W Function

4 Abbreviated title: GDI Engine Control and Optimization

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14 Highlights

- EOQ approach for fuel injection event in GDI engine has been evaluated.
- Analogy between EOQ and fuel injection and combustion process has been drawn.
- Components that contribute to the loss of energy in the system have been modelled using EOQ.
- A fuel injection control strategy has been proposed using EOQ and Lambert W function.

19 ABSTRACT

20 The present work evaluated the suitability of Economic Order Quantity (EOQ), commonly used in supply 21 chain management and process optimization, for combustion in Gasoline Direct Injected (GDI) engines. It 22 identified appropriate sub-models to draw an analogy between the EOQ for melon picking and fuel injection in GDI engines. It used experimental data from in-cylinder combustion processes for validating 23 24 the model. It used peak cylinder pressure and indicative mean effective pressure for validating the model; 25 the R² value for linear correlation between the experimental value and estimated value is 0.98. This work 26 proposes that the EOQ based on Lambert W function could be employed for optimizing the fuel quantity in 27 GDI engines for real-world fuel economy.

- 28 KEYWORDS: Fuel consumption, Economic Order Quantity, Lambert W function, Gasoline Direct Injection
- 29

30 NOMENCLATURE

31		
32	ANN	Artificial Neural Network
33	DISI	Direct Injection, Spark Ignition
34	ECU	Electronic Control Unit
35	EOQ	Economic Order Quantity
36	GDI	Gasoline Direct Injection
37	MVT	Marginal Value of Time
38	PFI	Port Fuel Injection
20	CMD	Soutor Moon Diamotor

- 39 SMD Sauter Mean Diameter
- 40 **THC** Total hydrocarbon 41

43 **1. INTRODUCTION**

44 Improving the fuel economy of the internal combustion engines has been one of the main goals of 45 automotive industry along with meeting emission targets set by the legislators, ever since the contribution 46 of engine-out CO₂ from combustion engine was recognized as a contributor towards greenhouse gas 47 inventory. The role of fuel injection and electronic control systems for increased fuel economy, reduced 48 emission levels and overall improvement of thermodynamic efficiency of the internal combustion engines 49 cannot be over emphasized (Heintz, et al., 2001). Therefore, it can be seen that successive legislative 50 targets forced the vehicle manufacturers to move from carburetion to fuel injection; and within the fuel 51 injection, from low pressure injection to high pressure injection; and crude fuel calibration procedure to 52 various forms of complex fuel injection optimization strategy (Salazar & Ghandhi, 2006). Moving away 53 from carburetion or port fuel injection system to direction injection system for gasoline application provides 54 ample opportunity for metering the quantity of the fuel precisely for every cycle per cylinder in a multi-55 cylinder engine. In a carbureted engine, metering the quantity of fuel may not be precise; however, the 56 time available for fuel evaporation before it reaches the cylinder or before the start of the combustion is 57 comparatively long and therefore, the combustion takes place almost in a pre-mixed mode (Khan, et al., 58 2009). However, a significant proportion of unburned fuel could escape the combustion process because 59 of the excessive wall wetting in the manifold. This excessive wall wetting increases the emission levels 60 beyond the current requirements. Similarly, in a Port Fuel Injection system (PFI), fuel is injected in the port 61 upstream of the intake valve; therefore, less surface area is exposed for wall wetting when compared with 62 the carbureted system. In addition, metering the quantity of the fuel per cylinder in PFI system could be 63 more precise when compared to carbureted engines (Zhao, et al., 1997). Similar to carbureted engines, 64 the combustion in port injection engines also takes place in a pre-mixed mode, and therefore smoothness 65 of the cylinder pressure leads to smooth power output. However, pre-evaporation and pre-mixing 66 introduces another limitation on workable compression ratio for the given fuel octane rating (Zhao, et al., 67 1997) in addition to emission levels, even though the emission levels are significantly lower than that of 68 carbureted engines.

69 In contrast to PFI engines, modern gasoline direct fuel injection engines, where the fuel is injected directly 70 into the combustion chamber enable the designers to use higher compression ratio to improve the overall 71 thermal efficiency of the engines. This higher compression ratio is achievable because of the charge 72 cooling effect, which lowers the charge temperature due to the evaporation process that takes place within 73 the combustion chamber (Singh, et al., 2014). One of the inherent limitations of this strategy is the reduced 74 time available for the evaporation and mixing process which directly influences the mode of combustion 75 and also emission formation mechanisms, especially nano-scale particulate matter formation in gasoline 76 direct injection engines (Samuel, et al., 2010). Recently introduced EURO VI emission standards for light-77 duty vehicles includes the levels of nano-scale particulate matter from gasoline direct injection engines and 78 therefore, it is one of the major challenges of the automotive manufacturers using GDI engines. The 79 opportunity for operating the engine at higher compression ratio with the ability to precisely meter the fuel 80 cycle by cycle demands a better optimization strategy in order to overcome the drawbacks of reducing 81 injection and evaporation timing (Whelan, et al., 2012). Hence, various manufacturers are in search of 82 complex optimization algorithms to optimize the fuel injection strategy for improved fuel economy and 83 reduced emission levels.

84 One of the methods is the Genetic Algorithm. Genetic algorithms are proposed for optimizing the fuel 85 injection strategy in gasoline direct injection engines (Tanner & Srinivasan, 2007). Artificial Neural Network 86 (ANN) is another method that can also be applied for the optimization of the fuel injection strategy. ANN is 87 a method commonly used for information processing based on the way biological nervous systems 88 process information (Stergiou & Siganos, 2015) and has been applied to the air-fuel ratio control (Lenz & 89 Schröder, 1998), characterization of DISI emissions and fuel economy (Shayler, et al., 2001) and in 90 powertrain simulation tools (Le berr, et al., 2008). Another way of optimizing the fuel injection strategy is to 91 identify different events and processes, which take part during fuel injection phase and use appropriate 92 phenomenological or semi-empirical models to include the effect of those events and processes in the 93 optimization algorithm. The submodels required for developing optimization algorithm are; fuel spray and 94 impingement model, wall wetting and evaporation model and combustion and heat transfer models.

95 In GDI engines, the fuel droplets may impinge onto the combustion chamber walls because the fuel is 96 directly injected into the combustion chamber at a higher velocity. If these fuel droplets are not completely 97 evaporated on time they will increase the quantity of unburned fuel, and therefore THC emissions. The 98 level of wall wetting in GDI engines is typically higher than those in port fuel injection engines due to higher 99 injection pressures and resulting higher penetration velocity and distance (Serras-Pereira, et al., 2007). A 100 study by Hung et al (Hung, et al., 2007) concludes that for given port flow characteristics, piston and 101 cylinder head, injector spray pattern has higher levels of influence on the quality of air-fuel mixture. They 102 also suggest that fuel impingement on in-cylinder walls can be minimized and fuel-air mixing could be 103 improved by choosing an appropriate spray pattern. In the same line of argument, Mittal et al (Mittal, et al., 104 2010) show that split injection is an effective way to reduce the overall fuel impingement on in-cylinder 105 surfaces.

106 Once the fuel spray is in the combustion chamber the amount of fuel available for combustion is 107 determined by the rate of evaporation and the amount of fuel suspended in the air in vapour phase. This 108 determines the quality of the fuel-air mixture (Gold, et al., 2001; Khan, et al., 2009) that in turn determines 109 the efficiency of the combustion. The next stage is the heat transfer phase; heat lost to the wall during 110 combustion process determines the actual amount of energy available for changing the cylinder pressure 111 (Harigaya, et al., 1993; Hensel, et al., 2009; Morel, et al., 1988; Shayler & and May, 1995). This cylinder 112 pressure for a given combustion chamber volume change during the overall process determine the net 113 energy conversion from fuel to useful work output. Therefore, total quantity of the fuel injected to cylinder pressure could be used to estimate the indicated thermal efficiency of the engine. The overall thermal 114 115 efficiency of the engine is only around 40% for internal combustion engines.

In the light of this brief literature review it can be concluded that identifying suitable optimization strategy 116 117 for fuel injection in GDI engine is still open to research. A closer look at the phases in Economic Order 118 Quantity (EOQ) mainly used for perishable inventory show that an analogy could be drawn by comparing 119 the processes involved in GDI engines. This analogy is based on the fact that the injected fuel is losing its 120 "value" as a function of time from the point of injection to "sold" in form of change in cylinder pressure. 121 Therefore, the purpose of this study was to evaluate the suitability of the Economic Order Quantity based 122 on Lambert W function, which has been successfully applied for perishable inventory, for fuel injection and 123 combustion process in gasoline direction injection engine. The following section will briefly review the 124 approach proposed by (Blackburn & Scudder, 2009) for solving Economic Order Quantity problems and 125 show the relevant processes applicable for the present work.

126

127 **2. Economic Order Quantity**

Economic order quantity (EOQ) is the order quantity that minimizes the total inventory holding costs and ordering costs (Blackburn & Scudder, 2009). Blackburn and Scudder investigated the supply chain strategies for melons, a perishable and fresh product, and proposed a method for minimizing the cost value. The logical approach proposed by Blackburn and Scudder (Blackburn & Scudder, 2009) and the variables considered in their model are summarized in *Figure 1*.

133 The perishable melon as a fresh product, has its peak value at the yield. After being collected, the value of 134 the product is reduced in an exponential manner based on the time spent during different processes. The 135 deterioration of the quality of the product has two phases; the first phase is a fast deterioration phase and 136 then in the second phase the product is brought to the cooling facility where the deterioration is 137 diminished. (Blackburn & Scudder, 2009). They proposed a model based on product's marginal value of 138 time to minimize the lost value of the product during the supply chain. Marginal value of time (MVT) is 139 defined to be the change in value of a unit of product per unit time at a given point in the supply chain 140 (Blackburn & Scudder, 2009). Figure 2 shows the reduction in value of the product (melons) over the time 141 in the supply chain.

The variables in the model presented by (Blackburn & Scudder, 2009) are summarized in the flow chart **Figure 1** and also in **Figure 2**. The variables corresponding to the first part of the supply chain are: total annual harvest (*D*), the transfer batch size in cartons (*Q*), maximum value of a carton of product at time t=0

145 (*V*), picking rate in cartons per hour (*P*), deterioration rate in value of product per hour (α), batch transfer 146 time in dollars (*K*) and transfer time in hours from field to the cooling facility (t_r).

147 The variables that were selected during the second part of the supply chain corresponding to the time 148 when the product is in the cooling facility are: deterioration in value of product per time (β), time (t_j) and 149 the cost of the transportation to the retailer (C_i).

150 The cost equation proposed by (Blackburn & Scudder, 2009) using these variables was:

$$TC = \frac{K \cdot D}{Q} + D \cdot V - \frac{D}{Q \cdot \alpha} \cdot \left(p \cdot V \cdot e^{-\beta \cdot t_j} \cdot e^{-\alpha \cdot t_r}\right) \cdot \left(1 - e^{-Q \cdot \alpha/p}\right) + c \cdot D + C_j$$
(1)

And the minimum cost equation ignoring the variables D and C_i was:

$$minTC = \frac{1}{Q} \left[K - \left(p \cdot e^{-\beta \cdot t_j} \cdot e^{-\alpha \cdot t_r} \cdot V / \alpha \right) \cdot \left(1 - e^{-Q \cdot \alpha} / p \right) \right]$$
(2)

152 (Blackburn & Scudder, 2009) Also proposed that the optimal Q in the form of Lambert W function satisfies:

$$Q = \left(\frac{p}{\alpha} - k\right) \cdot e^{Q \cdot \alpha/p} - \frac{p}{\alpha}$$
(3)

153 Where the constant *k* corresponds to:

$$k = e^{\alpha \cdot t_r + \beta \cdot t_j} \cdot K/_V$$
(4)

154 **2.1 Analogy EOQ with fuel injection event**

A closer look at the phases in EOQ and the process involved in GDI combustion show that an analogy could be drawn. This analogy is based on the fact that the injected fuel is losing its "value" as a function of time from the point of injection to "sold" in form of change in cylinder pressure.

158 The first part of deterioration is considered to be the fuel that remains in the piston after the wall wetting 159 event. For this first part, evaporation during injection and wall wetting models have been developed. This 160 part is comparable to the phase where the melon is picked from the vine until it is brought to the cooling 161 facility in the melon supply chain. Second part of deterioration phase in the cooling facility is comparable 162 the heat transfer and energy loss in the combustion chamber. The final output from this analogy is the 163 quantity of fuel to be injected in order to optimize the peak pressure and the mean effective pressure 164 values. Figure 3 shows the analogy between economic order quantity and the fuel injection event 165 observed using the reduction of the product value over the time.

166 3. Lambert W function in EOQ

167 The Lambert W function, W[z], is the inverse function of $z=w[z]e^{w[z]}$ (Corless, et al., 1996), where "e" is the 168 natural exponential number and "z" a complex number. The real part of the solution to the Lambert W 169 function in terms of x is shown in *Figure* 4.

170 The Lambert W function does not differ too much from the inverse trigonometric functions. This function is 171 a multi-valued function on a given domain, and a principal branch needs to be defined. When "x" is real, as 172 can be seen in *Figure* 4, it has two solutions in the interval -1/e < x < 0. The branch that satisfies W[z] ≥ -1 173 is named $W_0[z]$, and is defined to be the principal branch (solid line in **Figure 4**), while the secondary real 174 branch that satisfies $W[z] \leq -1$ is designated $W_{-1}[z]$ (dashed line in **Figure 4**) (Stewart, 2005). Recently, 175 Disney and Warburton (Warburton, 2009; Disney & Warburton, 2012) introduced Lambert W function for 176 solving EOQ problems successfully. Following Disney & Warburton's study (Disney & Warburton, 2012), 177 Lambert W function has been found to be very useful for the EOQ problems with perishable inventory by 178 improving their lower bound for the optimum order quantity. Therefore, Lambert W function is used in this 179 study in order to determine the optimum fuel quantity based on the scheme used for Melon picking by 180 (Disney & Warburton, 2012).

181 Rearranging *equation 3* (Disney & Warburton, 2012):

$$\frac{p}{\alpha} - k = \left(Q + \frac{p}{\alpha}\right) \cdot e^{-\alpha Q/p}$$
(5)

$$\frac{\alpha k}{pe} - \frac{1}{e} = \left(-\frac{Q\alpha}{p} - 1\right) \cdot e^{\left(-\frac{Q\alpha}{p} - 1\right)}$$
(6)

182 The *equation 6* is in the form of Lambert W function $(y = xe^x)$, and the solution can be obtained by x = W[y]. Therefore, the exact solution is found in the W_{-1} branch of Lambert W function as follows.

$$-\frac{Q\alpha}{p} - 1 = W_{-1} \left[\frac{\alpha k}{pe} - \frac{1}{e} \right]$$
(7)

$$Q^* = -\frac{p}{\alpha} \left(W_{-1} \left[\frac{\alpha k}{pe} - \frac{1}{e} \right] + 1 \right)$$
(8)

The optimal value implies that Q^* does not exist if one of the values of $\{p, k, \alpha\}$ is negative, or if an even (or zero) number of $\{p, k, \alpha\}$ are negative and $V < (\frac{e^{\alpha \cdot t_r + \beta \cdot t_{j_K \alpha}}}{p})$ (Disney & Warburton, 2012). Notice that in the equation of the optimal value for Q^* a branch of Lambert W function is defined since the optimal order quantity, the deterioration rate, and the picking rate are always positive. Therefore follows W[z] < -1, which only happens on $W_{-1}[z]$, the secondary branch.

189 4. Fuel injection and combustion model based on EOQ

The possibility of developing an analogy between EOQ and fuel injection and combustion process in GDI engines is clear by drawing parallels between the EOQ and the physical process involved in fuel injection and combustion in GDI engines. In order to employ and validate this analogy, the details relating to appropriate mathematical models that could be used to represent the physical processes such as fuel injection and spray model, wall wetting and evaporation and heat transfer are essential and therefore, the following sections provide the details of these models from the published literature.

196 4.1 Fuel Spray model

Direct injection engines have the injectors mounted in the cylinder head and the fuel is injected directly into 197 198 the combustion chamber. As fuel is added during the compression stroke, only a short period is available 199 for the completion of evaporation and mixing process (Pulkrabek, 2003). Fuel injectors in direct injected 200 engines must operate with relatively high injection pressure when compared to port fuel injected engines. 201 A fuel spray model that includes vaporization process should be capable of predicting the occurrence of 202 wall-wetting in order to estimate the amount of fuel available for combustion. This work employs Hiroyasu 203 model since this model is known to give very good correlations with the experimental data (Boot, et al., 204 2007). Hiroyasu model (Hiroyasu, et al., 1993) considers mass of the fuel quantity evaporated during 205 injection and the spray tip penetration, to quantify the mass of fuel that will hit the piston. The fuel mass 206 evaporated will mix with the air and, therefore, is available for combustion. The remaining quantity of fuel 207 (hitting the piston) is estimated using a suitable wall-wetting model for evaluating the final quantity of fuel 208 available for combustion. Hiroyasu model divides the spray into multiple radial and axial packages as 209 shown in *Figure 5*. This model assumes no interaction between the packages; it considers that each 210 package initially consists of droplets of one unique diameter and the ambient gas entrainment is controlled 211 by conservation of momentum only.

Based on (Boot, et al., 2007) these information, two main phases can be identified in the spray development. The first phase is named as pre-breakup area, where the jet travels freely at a constant velocity. Spray tip penetration in the pre-breakup area could be estimated using Bernoulli's equation as shown below:

$$x(t) = C_a \cdot C_v \sqrt{\frac{2 \cdot (p_{inj} - p_a)}{\rho_f}} \cdot t = C_d \sqrt{\frac{2 \cdot (p_{inj} - p_a)}{\rho_f}} \cdot t$$
(9)

The coefficients C_a and C_v are measures for losses in the orifice area and velocity due to cavitation and frictional effects respectively. The introduction of the discharge coefficient is not part of the Hiroyasu model, but it was proposed to account for the dissimilarities in injector orifice dimensions between modern and older designs (Boot, et al., 2007). In this study the discharge coefficient is fixed to 0.39 and p_{inj} corresponds to the injection pressure, p_a to the pressure in the combustion chamber at the start of injection point obtained from the experimental in-cylinder pressure data acquired and ρ_f is the fuel density.

Atomization due to ambient gas entrainment is assumed to occur after a certain break-up time:

$$t_{b,k} = 4.351 \cdot \frac{\rho_f \cdot d}{C_a^2 \sqrt{\rho_a \cdot (p_{inj} - p_a)}} \cdot \left(\frac{6 - k}{5}\right)$$
(10)

223 Where *k* is the radial index based on the assumption that the initial jet periphery is more exposed to the 224 ambient gas than the core, ρ_a corresponds to the in-cylinder air density, and *d* is the injector orifice 225 dimension. As assumed by Boot et al. (Boot, et al., 2007) only the tip penetration along the central axis 226 (*k*=1) is considered since at this location wall wetting is most prevalent. Therefore, the model is assumed 227 to be a 1D model.

In a given radial package k at $t = t_{b,k}$, the Sauter Mean Diameter (SMD) (Boot, et al., 2007) is:

$$SMD = d \cdot max\left[\frac{SMD^{LS}}{d}, \frac{SMD^{HS}}{d}\right]$$
 (11)

$$\frac{SMD^{LS}}{d} = 4.12 \cdot Re^{0.12} \cdot We^{-0.75} \cdot \left(\frac{\mu_f}{\mu_a}\right)^{0.54} \cdot \left(\frac{\rho_f}{\rho_a}\right)^{0.18}$$
(12)

$$\frac{SMD^{HS}}{d} = 0.38 \cdot Re^{0.25} \cdot We^{-0.32} \cdot \left(\frac{\mu_f}{\mu_a}\right)^{0.37} \cdot \left(\frac{\rho_f}{\rho_a}\right)^{-0.47}$$
(13)

229 Initial diameter of the liquid fuel droplets right after the breakup time in each zone can be calculated using 230 **equations 11-13** and assuming normal distribution. Hence, the number of droplets in each zone can be 231 calculated knowing the SMD and the mass of fuel injected (Jung & Assanis, 2001). Where μ_a and μ_f are 232 the dynamic viscosity of gas in the cylinder and the dynamic viscosity of the fuel respectively, and *Re* is the 233 Reynolds number.

We is the Weber number, which is a dimensionless quantity for analysing the interface between two fluids and is defined as follows:

$$We = \frac{\left[\rho_a \cdot v^2 \cdot D\right]}{\sigma} \tag{14}$$

Where v is the droplet normal impact velocity estimated using the fuel mass rate, D is the characteristic length, in this case the droplet diameter and σ is the fuel surface tension for octane.

Similarly, penetration in the post-breakup area will occur when the ambient gas is entrained into a spray
 packet, at this point its velocity will decrease and the penetration at this stage could be estimated as
 follows (Boot, et al., 2007):

$$x(t)_{k=1} = 2.95 \cdot \left(\frac{p_{inj} - p_a}{\rho_a}\right)^{1/4} \sqrt{d \cdot (t - t_{b,k})} + C_d \sqrt{\frac{2 \cdot (p_{inj} - p_a)}{\rho_f}} \cdot t_{b,k}$$
(15)

A simplified droplet evaporation model as applied by (Boot, et al., 2007) based on (Lefebvre, 1989) was chosen for this study as follows:

$$\frac{dD}{dt} = \frac{4 \cdot k_a \cdot \ln(1 + B_m)}{\rho_f \cdot c_{p,a} \cdot D}$$
(16)

$$\frac{dm_D}{dt} = 2\pi \cdot D \cdot \frac{k_a}{c_{p,a}} \cdot \ln(1 + B_m)$$
(17)

$$\frac{dT_{f,s}}{dt} = \frac{dm_D}{dt} \cdot \frac{L_{f,s}}{c_{p,f,s} \cdot m_D} \cdot \left(\frac{B_T}{B_M} - 1\right)$$
(18)

The subscripts *a*, *f* and *s* correspond to the conditions in the ambient gas, liquid fuel and on the droplet surface temperature and the variables *k*, c_p , L_s , B_T and B_M correspond to thermal conductivity, specific heat, latent heat of evaporation and the heat and mass transfer numbers respectively.

246 **4.2 Wall Wetting Model**

The fuel droplets may impact onto the combustion chamber walls due to the fuel directly being injected at a high speed in GDI engines and this impingement on walls will affect the performance of the engine. If these fuel droplets are not evaporated completely before the start of combustion they will increase unburned fuel mass, and therefore increase the levels of THC and soot emission levels.

The evaporation process in this study is modelled using the method proposed by Curtis et al. (Curtis, et al., 1996). Although the model described by (Curtis, et al., 1996) is developed for a port fuel injected engine, the cylinder wall wetting model part can be implemented in the gasoline direct injection engine of this study considering only one film. The predictions for in-cylinder liquid fuel mass made by (Curtis, et al., 1996) were found to give reasonable prediction, hence a good correlation between the complexity of the model and the output given by the model is found using this method. The equation presented for the mass vaporization rate is:

$$\dot{m}_{v} = Sh\left(\frac{A_{ls}}{B}\rho_{gm}D_{fa}\ln\left(1 + \frac{\Delta MFF}{(1 - MFFs)}\right)\right)$$
(19)

258 Where A_{ls} corresponds to the liquid surface area, ρ_{gm} is fuel density, D_{fa} Corresponds to the mass 259 diffusion coefficient between the fuel and the air, ΔMFF Stands for the difference in mass fraction of fuel in 260 the vapour at the liquid surface and the free stream, MFFs is the mass fraction of fuel in the vapour at the 261 liquid surface, *Sh* is the Sherwood number, which is calculated by:

$$Sh = (1 + 0.023Re^{0.83}Sc^{0.33})$$
⁽²⁰⁾

262 4.3 Heat Transfer Model

263 One of the most important parameters related with engine performance, fuel economy and emission levels 264 is the thermal efficiency. The most significant operating variable which is directly linked to thermal 265 efficiency which can be controlled is heat loss from the combustion and expansion stroke (Andrews, et al., 266 1989). GDI engines suffer significantly during cold start because of the poor evaporation of the fuel and the 267 air-fuel ratio fluctuation due to wall wetting (Lahuerta & Samuel, 2013). This work uses Woschni's heat 268 transfer model (Woschni, 1967) for estimating heat transfer rate in the combustion chamber during 269 compression, gas exchange, combustion and expansion process separately. This study assumes steady 270 state considering that the role of convection is predominant compared with radiation (Hensel, et al., 2009) 271 in gasoline engines inside the combustion chamber. The heat transfer between the air-fuel mixture and the 272 gas side of the cylinder wall is calculated using the Newton's Law of convection:

$$Q = h_c \cdot A \cdot (T_g - T_w) \tag{21}$$

Where h_c corresponds to the convection coefficient which depends on the Nusselt, Reynolds and Prandtl numbers. *A* is the exposed area where the heat transfer is present, T_g is the mean gas temperature and T_w is the wall temperature.

The convection coefficient can be obtained by using the Woschni's heat transfer model, assuming a correlation based on Reynolds and Nusselt number, the convection coefficient is:

$$h_c = C \cdot B^{m-1} \cdot p^m \cdot w^m \cdot T^{0.75 - 1.62}$$
⁽²²⁾

Where *C* and *m* are empirical coefficients which take the values 0.0035 and 0.8 (Woschni, 1967) and *P* is the pressure and *T* is the temperature of the gas inside the combustion chamber. The average gas velocity *w* is determined by considering four-stroke, water-cooled and direct injection without swirl motion (Heywood, 1988):

$$w = C_1 \cdot \bar{S}_p + C_2 \cdot \frac{V_d \cdot T_r}{p_r \cdot V_r} \cdot (p - p_m)$$
(23)

282 Where V_d is the displacement volume, p_r , V_r and T_r are the fluid pressure, volume and temperature at a 283 reference state, and p_m is the isentropic pressure. Coefficients C_1 and C_2 depend on the phase of the 284 engine cycle, for combustion and expansion are 2.28 and $3.24 \cdot 10^{-3}$ respectively (Nieuwstadt, et al., 2000).Once the convection coefficient is known, heat losses can be estimated using the Newton's Law of 286 convection (*equation 21*). It is important to note that, with the aim of achieving a good relation between 287 complexity of the model and the output given, heat transfer losses have been calculated by using in-288 cylinder mean gas temperature, mean wall temperature and pressure at the start of combustion.

4.4 Fuel Injection Optimization Using Lambert W function based on EOQ

The analogy between variables is shown in *Table 1*. Although the analogy is based on the fact that the injected fuel is losing its "value" as a function of time from the point of injection to "sold" in form of change in cylinder pressure, each variable in EOQ could be mapped against corresponding variable in the fuel injection and associated process in the combustion chamber.

294 Once the mapping of the relationship between the economic order quantity variables and the fuel injection 295 event variables is done, the solution for the optimal fuel injection quantity can be found. Now, the optimal 296 quantity of fuel mass can be estimated as shown in *equation 24*. The decision variable for the fuel 297 injection problem is assumed to be the difference between the original injected fuel mass and the optimal 298 fuel mass to be injected.

299

$$Q^* = -\frac{p}{\alpha} \left(W_{-1} \left[\frac{\alpha k}{pe} - \frac{1}{e} \right] + 1 \right) * InjectionDuration$$
(24)

The data required for applying the optimization equation are obtained using the wall wetting, evaporation and heat transfer models. Once all data is obtained, the equation is solved and the fuel mass quantity to subtract from the original value is known. Therefore, the total quantity of fuel to be injected is estimated and the required new injection time is obtained. It is important to note that the main output from that model is the new injection duration, as the fuel mass flow rate is imposed by the injector, injection pressure and combustion chamber conditions. The present work used MATLAB® for solving wall wetting, evaporation and heat transfer models and for analytical solving Lambert W function.

308 5. Results and discussion

309 **5.1 Engine**

A Euro-IV compliant, 1.6-L, four-cylinder in-line, GDI, turbocharged and intercooled spark ignition engine was used in this study. The specifications of the engine are listed in *Table 2*:

Experiments have been carried out at 3 different speeds (2000, 2400 and 2800 rpm) and at different loading conditions at each operating speed (20, 40, 60. 80, 100 and 120 Nm).

314 **5.2 Model validation**

315 In order to validate the heat transfer and wall wetting models, measured in-cylinder pressure was used. 316 The results are validated using peak cylinder pressure and area under the curve. Since the peak pressure 317 has direct correlation with the location of maximum heat release and the area under the curve represents 318 the indicative work, i.e., cumulative energy release, these two variables were chosen for validation. The selected area is in the range from -30 crankshaft angle to +70 crankshaft angle which includes latest part 319 320 of compression, duration where the impact of evaporation on cylinder pressure is dominant, combustion 321 and early part of expansion. The results of peak pressure and area under the pressure curve after applying 322 the fuel spray model, the wall wetting model and heat transfer model and carrying out the model validation 323 are summarized in the Annex A.

Estimated peak pressure values (for model validation purpose) versus experimental peak pressure values for the engines conditions are shown in *Figure 6a*. Similarly *Figure 6b* shows the area under the curve.

Experimental value and the estimated values show linear correlations and the R² value is 0.98 for the peak pressure as well as for the indicated mean effective pressure. It shows acceptable level of correlations for validation purposes.

329 **5.3 Fuel injection optimization using Lambert W function results**

- 330 The results obtained regarding the quantity of fuel to be injected are summarized in Figure 8:
- 331

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For the given operating conditions, the possibility of using EOQ for optimizing the fuel quantity was studied. As can be observed in *Table 3* and Figure 7, at each engine condition fuel was optimized using Lambert W function based on Economic Order Quantity. As previously mentioned, the main output from the model is the new injection duration for a given fuel pressure and flow rate through the injector, as the fuel mass flow rate is imposed by the injector, injection pressure and combustion chamber conditions. Therefore, the final values of injection duration is summarized in the following table:

339

The main difference between the original mode of operation and the mode resulting from the optimization is the change of air fuel ratio. After applying the analogy between EOQ and fuel injection process, different amount of fuel is injected for the same air mass in the cylinder, hence the air fuel ratio needs to be controlled if a fixed-air fuel ratio is maintained at a constant value. The final air-fuel ratio obtained for each engine condition can be observed in Table 4. However, if employed, air-fuel ratio could be adjusted using air-flow controls

347 **5.4 Control Strategy**

The implementation of this method in the Electronic Control Unit (ECU) could be done by adding a new stage in the electronic control unit after the initial calculation of the ECU based on fuel pressure, fuel temperature and air mass flow rate, the injection duration is set to meet a total fuel mass quantity to be injected. Lambert W function based on economic order quantity could be used for determining the optimized mass of fuel to be injected for the for the real-world fuel economy once the base map is generated.

One of the main limitations of this study is that the current study used the experimental data to develop and verify the model, however, we couldn't deploy the model through ECU to study the effectiveness of the model since the purpose of the study is to identify the possibility of using EOQ for fuel injection strategy. The application of EOQ using Lambert W function offers promising direction because of the properties of Lambert W functions.

359 The final output is the new injection time that allows the ECU to inject the optimum quantity of fuel.



360

361 6. Conclusions

This work investigated the application of the Economic Order Quantity problem with perishable product to optimize the quantity of fuel to be injected in GDI engine through the use of Lambert W function.

- Evaporation, wall wetting and heat transfer models have been developed in order to see how it affects the fuel consumption. These models have been validated with two in-cylinder pressure based validation models based on peak pressure and area under the pressure curve. The models have been considered suitable for representing the events of fuel spray, wall wetting and heat transfer during engine operation.
- Analogy between Economic Order Quantity and fuel injection has been successfully established.
 It has been demonstrated that the exponential deterioration of the product can be applied to the in-cylinder fuel events from injection until peak pressure.
- 372 3. By applying Lambert W function to the analogy between EOQ and fuel injection, the quantity of 373 fuel to be injected can be optimized, and consequently the injection duration. The present study 374 shows that for the current experimental engine an average fuel saving of 5.71% could be 375 achieved for the engine conditions studied.
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- 478
- 479 ANNEX A
- 480

481 Evaporation, Wall Wetting and Heat Transfer validation results

482 **Peak pressure results:**

	Load (Nm)	Theoretical Peak pressure (bar)	Experimental peak pressure (bar)	% Difference
	20	18.26	17.25	+ 5.53
ε	40	25.27	24.97	+ 1.19
Ъ	60	34.99	32.11	+ 8.22
8	80	42.33	39.43	+ 6.85
20	100	52.80	47.48	+ 10.08
	120	58.76	54.44	+ 7.36
	40	16.14	13.72	+ 14.98
ud	60	19.10	19.78	- 3.57
0	80	28.70	28.21	+ 1.70
240	100	39.93	37.84	+ 5.24
,	120	50.39	46.33	+ 8.04
-	40	16.71	19.96	- 19.48
nq	60	24.18	23.57	+ 2.50
0 L	80	29.70	31.26	- 5.25
580	100	43.11	39.08	+ 9.37
	120	53.75	48.12	+ 10.48

483

484 Area under the pressure curve results:

	Load (Nm)	Theoretical area (bar*deg)	Experimental area (bar*deg)	% Difference
	20	1298.10	1220.30	+ 5.99
ε	40	1816.80	1750.34	+ 3.66
d d	60	2436.42	2261.20	+ 7.19
8	80	2933.74	2786.74	+ 5.01
20	100	3582.02	3316.47	+ 7.41
	120	3961.90	3787.70	+ 4.40
_	40	1134.22	1010.60	+ 10.90
μd	60	1423.89	1447.88	- 1.68
0	80	2054.67	2029.10	+ 1.25
540	100	2707.52	2670.53	+ 1.37
	120	3337.79	3238.00	+ 2.99
	40	1258.49	1434.97	- 13.94
8 E	60	1702.46	1682.46	+ 1.17
28 rp	80	2122.54	2212.54	- 4.24
	100	2950.59	2754.27	+ 6.65



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Fuel injection optimization using Lambert W function Results 488

489 Peak pressure results:

	Load (Nm)	Experimental peak pressure (bar)	Peak pressure after Lambert W function (bar)	% Difference
	20	17.25	17.53	+ 1.61
E	40	24.97	24.13	- 3.47
d	60	32.11	33.32	+ 3.64
00	80	39.43	39.67	+ 0.62
20	100	47.48	49.41	+ 3.90
	120	54.44	54.72	+ 0.51
ι	40	13.72	15.43	+ 11.07
ud	60	19.78	18.17	- 8.85
0	80	28.21	27.03	- 4.37
240	100	37.84	37.62	- 0.58
,	120	46.33	47.19	+ 1.81
ι	40	19.96	16.00	- 24.74
ud	60	23.57	22.97	- 2.61
0	80	31.26	27.89	- 12.08
280	100	39.08	40.47	+ 3.45
	120	48.12	50.41	+ 4.55
Area u	nder the pressure cu	rve results:	JS I	

490

491 Area under the pressure curve results:

	Load (Nm)	Experimental area (bar*deg)	Area after Lambert W function (bar*deg)	% Difference
	20	1220.30	1261.73	+ 3.28
ε	40	1750.34	1759.88	+ 0.54
<u>e</u>	60	2261.20	2353.37	+ 3.92
00	80	2786.74	2801.03	+ 0.51
20	100	3316.47	3412.11	+ 2.80
	120	3787.70	3759.78	- 0.74
-	40	1010.60	1098.79	+ 8.03
ud	60	1447.88	1377.57	- 5.10
õ	80	2029.10	1971.18	- 2.94
240	100	2670.53	2592.90	- 2.99
	120	3238.00	3177.81	- 1.89
-	40	1434.97	1223.22	- 17.23
ud	60	1682.46	1642.24	- 2.45
õ	80	2212.54	2032.06	- 8.88
580	100	2754.27	2818.42	+ 2.28
	120	3371.55	3478.82	+ 3.08

492









512 **Figure 6:** Experimental and Predicted peak cylinder pressure and indicated mean effecting pressure for validation 513 purpose (2000, 2400 and 2800 rpm engine speed and 20, 40, 60. 80, 100 and 120 Nm load conditions).

















Figure 8: Optimized values for injection duration.

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Table 1: Analogy between variables.



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Table 2: GDI test engine technical specifications.

Bore	77 mm
Stroke	85.8 mm
Compression ratio	10.5
Displacement	1598 cc
Rated Power	173 bhp @ 5500 rpm
Rated torque	240 Nm @ 1700-4500 rpm
Maximum fuel injection pressure	120 bar

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528

Table 3: Summary of results regarding fuel saving.

	Load (Nm)	Fuel injected (kg)	Optimized fuel to be injected (kg)	% Fuel saving when EOQ is applied
	20	7.28E-06	6.92E-06	4.95
ε	40	1.05E-05	9.88E-06	5.50
d d	60	1.39E-05	1.31E-05	5.92
00	80	1.71E-05	1.60E-05	6.62
20	100	2.13E-05	1.99E-05	6.71
	120	2.53E-05	2.36E-05	6.89
	40	9.73E-06	9.25E-06	4.91
8 E	60	1.32E-05	1.25E-05	5.32
24 rp	80	1.62E-05	1.52E-05	5.96
	100	2.11E-05	1.99E-05	5.68

	120	2.47E-05	2.32E-05	6.03
_	40	9.62E-06	9.16E-06	4.72
μ	60	1.34E-05	1.27E-05	5.11
0 -	80	1.60E-05	1.51E-05	5.74
580	100	2.10E-05	1.98E-05	5.70
	120	2.47E-05	2.33E-05	5.66

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Table 4: Air-fuel ratio at each engine condition after applying Lambert W function.

0 rpm 2400 rpm 2000 rpm	20 40 60 80 100 120 40 60 80 100 120 40	15.4661 15.5553 15.6242 15.7415 15.7565 15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm 2000 rpm	40 60 80 100 120 40 60 80 100 120 40	15.5553 15.6242 15.7415 15.7565 15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm 2000 rp	60 80 100 120 40 60 80 100 120 40	15.6242 15.7415 15.7565 15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm 2000	80 100 120 40 60 80 100 120 40	15.7415 15.7565 15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm 20	100 120 40 60 80 100 120 40	15.7565 15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm	120 40 60 80 100 120 40	15.7883 15.4583 15.5255 15.6324 15.5858 15.6422
0 rpm 2400 rpm	40 60 80 100 120 40	15.4583 15.5255 15.6324 15.5858
0 rpm 2400 rpn	60 80 100 120 40	15.5255 15.6324 15.5858
0 rpm 2400 r	80 100 120 40	15.6324 15.5858 15.6422
0 rpm 240	100 120 40	15.5858
0 rpm	120	15 6/22
0 rpm	40	13.0433
0 rpr	40	15.4287
o L	60	15.4924
-	80	15.5944
580	100	15.5887
	120	15.5825
	Neo o	