

Exploring quinolinone–chalcones: Synthesis, antioxidant potential and industrial applications in biofuels

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Abstract: Biofuels undergo oxidative degradation, which reduces their storage stability and efficiency, hindering their large-scale adoption and integration into the global energy matrix. To address this issue, we synthesized and characterized four novel quinolinone–chalcone derivatives (substituted with Br, Cl, OCH₃ and OCH₂CH₃) as potential antioxidant additives. Structural analysis, density functional theory calculations and machine learning-based hydroxyl radical scavenging rate evaluation were performed to assess the influence of the substituents on stability and antioxidant capacity. Single-crystal X-ray diffraction revealed C–H···O and C–H···π interactions contributing to supramolecular stability, while electronic property analysis indicated enhanced kinetic stability and regions near the sulfonamide group prone to nucleophilic attack. Machine learning predictions showed that the bromine derivative exhibited a hydroxyl radical scavenging rate of $7.97 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$, comparable with the commercial antioxidant butylated hydroxytoluene ($4.34 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$). These findings highlight the potential of quinolinone–chalcone derivatives as efficient antioxidant additives, improving the oxidative stability and sustainability of biofuels. © 2025 The Author(s). *Biofuels*, *Bioproducts and Biorefining* published by Society of Industrial Chemistry and John Wiley & Sons Ltd.

Supporting information may be found in the online version of this article.

Key words: biofuels; oxidative stability; quinolinone–chalcones; machine learning

Introduction

The planned synthesis of bioactive compounds is the initial step in product development, and we have found several promising classes of compounds described in the literature with potential biological activities. Quinolinones represent a class of bicyclic nitrogen-containing heterocycles that have attracted increasing interest in scientific research owing to their diverse biological and chemical activities.^{1,2} These compounds, structurally derived from quinoline, feature a benzene ring fused to a pyridine ring and are known for their chemical stability and multifunctional properties.^{3–5} In recent years, quinolinones and their derivatives (such as quinolinone–chalcones) have been explored for their potential industrial applications, including potential antioxidant activity,^{6–15} as well as antimicrobial activity.^{16–25} These described properties are desirable in new drugs, new materials and new additives. For biofuels, such properties could prevent fuel oxidation and combat the proliferation of microorganisms that degrade them.^{26,27}

With the increasing demand for sustainable energy sources and the need to reduce dependence on fossil fuels, biofuels have emerged as a promising alternative.^{28–31} However, one of the main challenges in using biofuels is their susceptibility to oxidation, which can compromise fuel efficiency and lead to the formation of undesirable subproducts, such as carboxylic acids and insoluble polymers, resulting in corrosion and clogging of fuel injection systems.^{26,27,32–34} In this context, quinolinones and their derivatives can be studied as potential additives owing to their antioxidant and antimicrobial properties. The antioxidant activity of quinolinones can be associated with their ability to donate electrons or hydrogen atoms, neutralizing free radicals and interrupting oxidation chain reactions.^{35–39}

Recent studies indicate that chalcones (compounds that can be hybridized with quinolinones) can be used as additives in biofuels, significantly improving their oxidative stability, prolonging their shelf life, and enhancing fuel efficiency.^{40–43} A study conducted by Carapina *et al.* (2018)⁴⁴ demonstrated the influence of chalcones on the antioxidation of biodiesel, resulting in a significant improvement in the fuel's oxidative stability. Additionally, Moreira *et al.* (2022)⁴⁵ showed the antioxidant activity of a tri-methoxy chalcone added to a biodiesel blend, with stability results comparable with those for commercial additives such as BHT (butylated hydroxytoluene). Based on these aspects, hybrid structures like quinolinone–chalcones are promising, but there is still much to explore. New studies are needed to fully understand the antioxidant mechanisms of action of these compounds. Additionally, evaluating the environmental and economic

impacts associated with the production and use of these substances is essential to determine their commercial viability. Machine learning (ML) has become important for predictive analysis in molecular chemistry. XGBoost (Extreme Gradient Boosting), known for its efficiency in classification and regression tasks, is widely used to model molecular properties, optimize reactions and predict antioxidant activity.^{46,47} In this study, we employ machine learning techniques with XGBoost through the PySIRC platform⁴⁸ to predict reaction rate constants and evaluate the physicochemical properties of novel quinolinone chalcones. We synthesized four quinolinone chalcone derivatives (QC01 to QC04) to investigate the influence of different substituents (Br, Cl, OCH₃ and OCH₂CH₃) on their supramolecular arrangement, reactivity and potential applications as additives in biodiesel blends.

Experimental and computational procedures

Synthesis of the quinolinone–chalcones

Compounds QC01 [2-(4-ethoxyphenyl)-2,3-dihydro-1-(phenylsulfonyl)quinolin-4(1H)-one], QC02 [2-(4-methoxyphenyl)-2,3-dihydro-1-(phenylsulfonyl)quinolin-4(1H)-one], (QC03) 2-(4-chlorophenyl)-2,3-dihydro-1-(phenylsulfonyl)quinolin-4(1H)-one and (QC04) 2-(4-bromophenyl)-2,3-dihydro-1-(phenylsulfonyl)quinolin-4(1H)-one were obtained from the synthesis of four chalcones (CH01 to CH04). 2'-N-Phenylsulfonylacetophenone was synthesized in the first step (intermediate I01). The reaction of benzenesulfonyl chloride with 2-aminoacetophenone in dichloromethane (Scheme 1) was carried out following previously described conditions.^{49–52} Four 2'-N-sulfonamide chalcones (CH01 to CH04) were synthesized *via* Claisen–Schmidt condensation between I01 and a benzaldehyde, using basic catalysis in an ethanolic medium (Scheme 2).

Four quinolinones were prepared (Scheme 3). The cyclized compounds QC01, QC02, QC03 and QC04 were obtained, starting respectively from CH01, CH02, CH03 and CH04, in high yields and purified by filtration, based on the procedure previously described.^{51,53}

Procedure for preparing sulfonamide chalcones CH01 to CH04

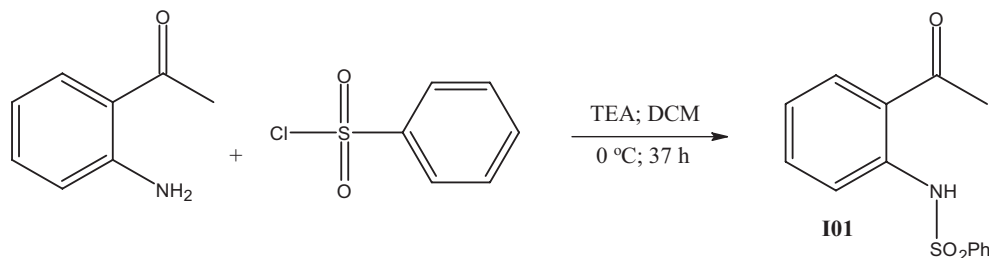
56.8 mmol (3.187 g) of potassium hydroxide was weighed and dissolved in 3.0 mL of water. Ethanol (400.0 mL) was added to this solution and homogenized. An amount of 20 mmol (5.506 g) of I01 was added and dissolved. Then the aldehyde was added (40 mmol for the synthesis of the compounds

CH01, **CH02** and **CH03**, and 21 mmol for the synthesis of the compound **CH04** – 6.007 g of 4-ethoxybenzaldehyde, 5.446 g of 4-methoxybenzaldehyde, 5.623 g of 4-chlorobenzaldehyde and 3.885 g of 4-bromobenzaldehyde, respectively), magnetic stirring started and the reaction timed from that point. The temperature of the system was ambient (near to 25 °C). The reaction progress was tracked using thin-layer chromatography in normal phase, with a hexane and ethyl acetate solution in an 85:15 ratio as the mobile phase. The reactions were quenched by the gradual addition of 37% (w/w) hydrochloric acid in an amount equimolar to potassium hydroxide, while the system remained under stirring. The time elapsed in each reaction was as follows (h:min): **CH01**, 25:30; **CH02**, 2:15; **CH03**, 2:00; and **CH04**, 2:00. The system, after neutralization, was poured into 800 mL of water. It was then extracted with 200 mL of dichloromethane. The organic phase was passed through filter paper containing anhydrous sodium sulfate and left to evaporate slowly. Before fully drying, the crystals were collected and rinsed with ethanol. After drying, they were weighed for yield determination. The yield of each reaction

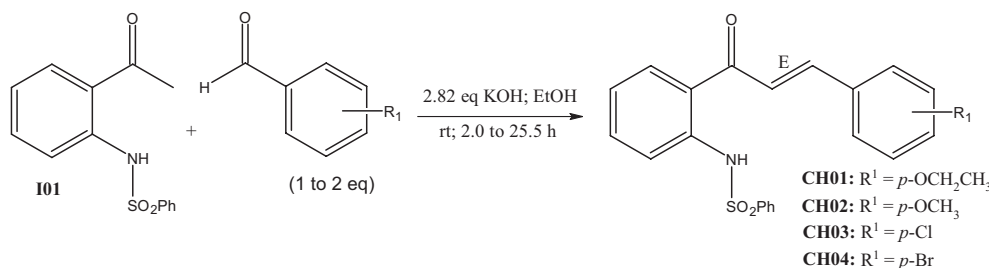
was as follows: **CH01**, 6.20 g (76.1%); **CH02**, 7.36 g (93.5%); **CH03**, 5.64 (70.9%) and **CH04**, 4.61 g (52.0%). For the ^1H NMR, ^{13}C NMR, IR and HRMS for **CH1** to **CH4**, see the supplementary information.

General procedure for preparing quinolinones QC01 to QC04

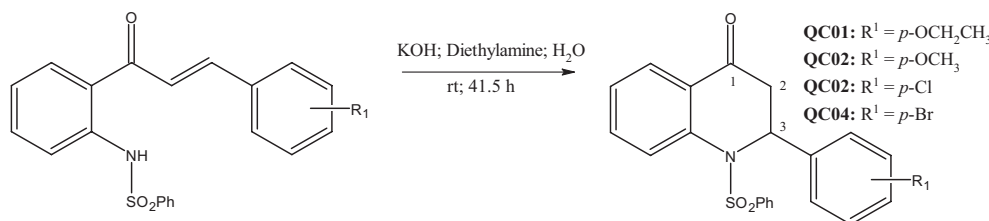
Chalcones **CH01** to **CH04** (200 mg) were weighed, finely ground in a pestle and mortar and transferred to an aqueous solution containing potassium hydroxide (449 mg in 100 mL) and diethylamine (124 μL in 100 mL). The system was kept under agitation for 41 h 30 min, at which point it was no longer possible to verify the presence of chalcone on the thin-layer chromatography plates (normal stationary phase and solution of hexane and ethyl acetate in the proportion of 85:15 as mobile phase); there was just one fluorescent spot under UV light of 366 nm. The precipitate was filtered and washed with water. It was then dissolved in 20 mL of dichloromethane, and the resulting solution was extracted with 20 mL of water. The organic phase was passed through



Scheme 1. Representation of conditions for the synthesis of 2'-N-phenylsulfonylacetophenone (**I01**).



Scheme 2. Representation of Claisen-Schmidt condensation for the synthesis of the chalcones **CH01** to **CH04**.



Scheme 3. Synthesis of quinolinones **QC01** to **QC04**.

filter paper containing anhydrous sodium sulfate and left to dry slowly. The mass of products obtained was: **QC01**, 166 mg (82.7%); **QC02**, 196 mg (98.0%); **QC03**, 171 mg (85.5%); **QC04**, 157 mg (78.5%). For the ^1H NMR, ^{13}C NMR, and HRMS for **QC01** to **QC04**, see the supplementary information.

Although the chemical process initially originates on a small scale in the laboratory, this yield can be scaled up to a large scale, which is necessary for the chemical industry.⁵⁴ Furthermore, the Global Health Innovative Technology determines some criteria for promising molecular successes, such as synthesis in five or fewer steps, acceptable yield and solubility and purity >90%.⁵⁵ The quinolinone–chalcones synthesized in the work meet these criteria and present a relatively simple synthesis, and this class of compounds has acceptable solubility in polar and nonpolar solvents.⁵⁶

Solid-state analysis

The crystallographic analysis of the four quinolinone–chalcone compounds was conducted using X-ray diffraction data collected with a Bruker APEX-II CCD diffractometer, employing MoK α radiation ($\lambda=0.71073$ Å) at 120 K. The structures were solved and refined using SHELX software,⁵⁷ implemented on the OLEX2 platform.⁵⁸ The crystallographic data for these structures have been deposited in the Cambridge Crystallographic Data Centre⁵⁹ under the following deposition numbers: 2393625 (**QC01**), 2393627 (**QC02**), 2393628 (**QC03**) and 2393629 (**QC04**). The representations of molecular interactions for these structures were generated using the Mercury software,⁶⁰ based on the geometric parameters experimentally obtained from X-ray diffraction. Additionally, CrystalExplorer software was employed to quantify the molecular interactions in the quinolinone–chalcones, based on electron density through Hirshfeld surfaces.⁶¹ These Hirshfeld surfaces are derived from the normalized contact distances, including the distance from the surface to the nearest atom within the same molecule (*de*) and the distance from the surface to the nearest atom in a neighboring molecule (*di*).^{62,63} The 2D fingerprint plots are graphical representations of the *de versus di* distribution.⁶⁴

Theoretical analysis and machine-learning procedures

The theoretical analysis for the quinolinone–chalcones was conducted using density functional theory (DFT).⁶⁵ These calculations were performed for both molecular conformations (solid state and gas phase), based on the geometric parameters obtained from X-ray diffraction,

utilizing Gaussian09 software⁶⁶ at the M06-2X/6-311++G(d,p)^{67–69} theory level. From the wave function generated in these calculations, the frontier molecular orbitals were computed, as well as the HOMO (highest occupied molecular orbital) and the LUMO (lowest unoccupied molecular orbital).^{70,71} To further investigate the reactive sites of these quinolinone–chalcones, the molecular electrostatic potential (MEP) map was also calculated.⁷²

The oxidation reactions by free radicals were simulated using *pySiRC*,⁴⁸ a machine-learning platform, selecting the hydroxyl radical ($\cdot\text{OH}$) as a model for oxidative degradation. Predictions of the reaction rate constant (k_{OH}) in B20 blends were generated using the Morgan and MACCS fingerprints combined with the trained model under XGBoost algorithm. In the *pySiRC*⁴⁸ platform, input files can be provided in *.txt* format, containing the SMILES identifiers of the molecules (SMILES for **QC01** to **QC04** are shown in the supplementary information – Table S1). The authors applied XGBoost – a method based on a gradient-boosting decision tree,⁴⁶ one algorithm among others that they used to predict the kinetic constant of organic pollutants in the aqueous phase. This algorithm has a database built for the reaction rate constants of an oxidative process mediated by the $\cdot\text{OH}$ radical, formed by a group of 1374 parameters (k_{OH}) for organic compounds. The referred kinetic parameters are catalogued under standard conditions, 25 °C and 1 mol·L⁻¹ in the aqueous phase. This algorithm was developed under the same Gradient Boosting framework, aiming to be highly efficient, flexible and portable. XGBoost provides a parallel tree reinforcement that solves many data science problems quickly and accurately and has therefore been widely used in recent literature.^{73–76} The performance indices – correlation coefficient (R^2), Pearson correlation coefficient of prediction (r^2), root-mean-square deviation (RMSE), and external validation (Q_{ext}^2) – were calculated using the formulas:

$$R^2 = 1 - \frac{\sum_{i=1}^n (y_{\text{exp}} - y_{\text{pred}})^2}{\sum_{i=1}^n (y_{\text{exp}} - \bar{y}_{\text{exp}})^2} \quad (1)$$

$$r^2 = \frac{\sum_{i=1}^n (y_{\text{exp}} - \bar{y}_{\text{exp}}) \cdot (y_{\text{pred}} - \bar{y}_{\text{pred}})}{\sqrt{\sum_{i=1}^n (y_{\text{exp}} - \bar{y}_{\text{exp}})^2} \sqrt{\sum_{i=1}^n (y_{\text{pred}} - \bar{y}_{\text{pred}})^2}} \quad (2)$$

$$\text{RMSE} = \sqrt{\frac{\sum_{i=1}^n (y_{\text{exp}} - y_{\text{pred}})^2}{n}} \quad (3)$$

$$Q_{\text{ext}}^2 = 1 - \frac{\sum_{i=1}^n (y_{\text{exp}} - y_{\text{pred}})^2}{\sum_{i=1}^n (y_{\text{exp}} - \bar{y}_{\text{exp}}^{\text{tr}})^2} \quad (4)$$

where y_{exp} , y_{pred} , \bar{y}_{exp} and \bar{y}_{pred} are the experimental, predicted and average of the experimental and predicted values of the dependent variable (over the validation set), respectively, and $\bar{y}_{\text{exp}}^{\text{tr}}$ is the average value of the dependent variable for the training set – the sums cover all the compounds in the validation set.⁴⁸ The authors of *pySiRC* have reported this model by yielding a high goodness-of-fit for training ($R^2 > 0.937$) and good predictive accuracy for the test set ($R_{\text{ext}}^2 = Q_{\text{ext}}^2$ in the 0.707–0.823 range).⁴⁸ The hyperparameters of the machine learning model (XGBoost) used in *pySiRC* were: base score (0.5); booster (gbtree); gamma (0); learning_rate (0.3); max delta step (0); max depth (6); Number of gradient boosted trees (150); alpha (0); lambda (1).

The applicability domain (AD%) indicates the similarity between query compounds and the database, supporting the model's reliability.⁴⁸ The *Tanimoto* index⁷⁷ was used to evaluate the similarity between the compounds in the work, according to the following equation:

$$T_c(A, B) = \frac{c}{a + b - c} \quad (5)$$

where a and b are the number of structural features for compound A and B compared, or bits set to 1, in each molecule, and c is the number in common.

The k_{OH} parameters were calculated for major diesel (1-decene) and biodiesel compounds (e.g. methyl 9-octadecenoate, BD M9OD 19.98%; methyl palmitate, BD MPAL 12.87%; and methyl 8,11-octadecenoate, BD M8OD 10.22%)⁷⁸ and tested against other antioxidant candidates, including compounds (**QC01**, **QC02**, **QC03** and **QC04**), previously tested additives and commercial additives (BHT, tert-butyl hydroquinone, TBHQ; butylated hydroxyanisole, BHA; propyl gallate, PG; pyrogallol, PY; and gallic acid, GA).⁷⁹

Results and discussion

Solid state description

Evidence of the successful synthesis of compounds **QC01** to **QC04** was the nonobservance in ¹H NMR spectra of the signal from sulfonamide hydrogens in the δ 11.3 region. Further evidence was provided by two diastereotopic α -carbonyl hydrogens (H2a and H2b, *Scheme 3*), with a geminal coupling of 17.9 Hz. These hydrogens are coupled

with vicinal hydrogen (H3) with coupling constants near to 6.0 and 1.7 Hz, for *cis* and *trans* respectively (Figs *S1–S12*). All of these structures were confirmed through single-crystal X-ray diffraction analysis. The four analyzed compounds are structurally similar (*Tanimoto* index greater than 0.68, *Table S2*), differing by the substitution of the OCH₂CH₃ group at the *para* position of the aromatic ring 2 in compound **QC01** and the OCH₃ group in compound **QC02**. Compounds **QC03** and **QC04** feature chlorine (Cl) and bromine (Br) substitutions, respectively, also at the *para* position of aromatic ring 2 (Fig. 1). The compounds **QC01**, **QC02** and **QC04** crystallize in the monoclinic space group *P2₁/c*, with four molecules per unit cell and similar volumes (1977.0 (7) Å³, 1835.8 (4) Å³ and 1806.7 (5) Å³, respectively). Compound **QC03** crystallizes in the triclinic space group *P1̄*, with two molecules per unit cell and a smaller volume compared with **QC01**, **QC02** and **QC04** (*Table 1*).

The structural conformation of a molecule can be linked to its biological properties or activities, such as the planar conformation of chalcones, which has been described as having antifungal⁸⁰ and cytotoxic activities.⁸¹ The presence of different substituents can result in distinct conformations for each molecule. To analyze these conformational differences, the four molecules were superimposed using the quinolinone core as an anchoring point. Molecules **CQ01** and **CQ02** have similar conformations; however, the presence of an OCH₂CH₃ group compared with an OCH₃ group causes a 7.08° difference in aromatic ring 2. Next, we observe the superimpositions of **CQ02** with **CQ04**, where the OCH₃ group compared with Br results in a 6.60° difference in aromatic ring 2. Furthermore, the superimposition of **CQ01** with **CQ04** shows a 13.48° difference in aromatic ring 2 owing to the presence of the OCH₂CH₃ group compared to Br. On the other hand, the **CQ03** molecule, which contains Cl, also shows conformational differences when compared with the others: **CQ03** with **CQ02** (Cl to OCH₃ – 7.08° in aromatic ring 1), **CQ03** with **CQ04** (Cl to Br – 20.18° in aromatic ring 1), and **CQ03** with **CQ01** (Cl to OCH₂CH₃ – 9.05° in aromatic ring 2) (Fig. 2). We observe that different substituents can lead to multiple conformations for each molecule. This variation can reflect the diverse intermolecular interactions and the specific supramolecular arrangement of each molecule, which can be directly linked to their physicochemical properties.

The supramolecular arrangement of **CQ01** is stabilized by C–H...O and C–H... π interactions, while **CQ02** is stabilized only by C–H...O interactions. For **CQ01**, the C20–H20...Cg3 interaction contributes to stability along (100) direction (Fig. 3c), C8–H8B...O1, C8–H8A...O1 and C12–H12...O3 interactions contribute along (010) direction

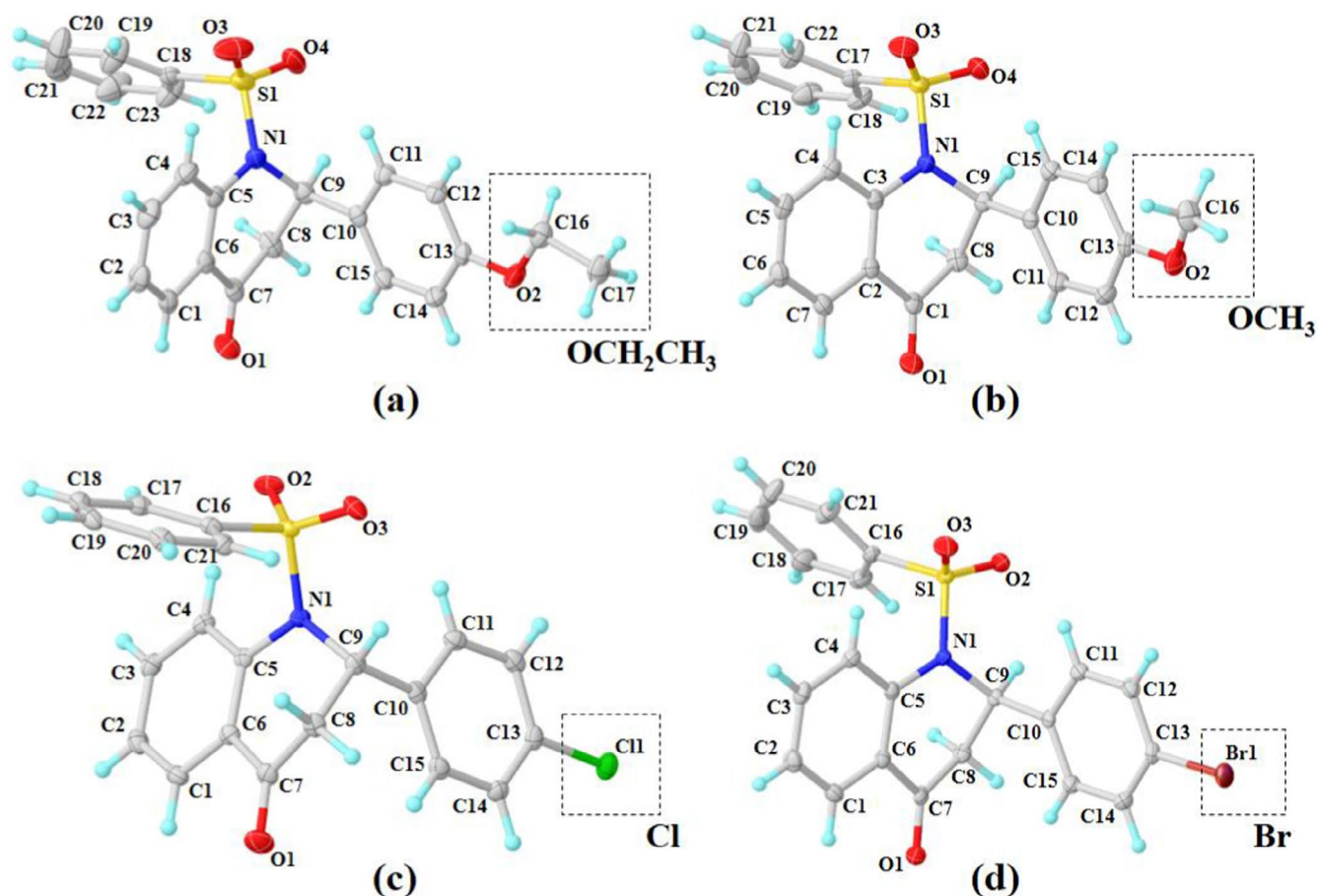


Figure 1. ORTEP representation for **QC01** (a), **QC02** (b), **QC03** (c) and **QC04** (d) with ellipsoids at 50% of probability.

(Fig. 3b), while C1–H1...O4 and C8–H8A...O4 interactions contribute to growth along (001) direction (Fig. 3a). For **CQ02**, C20–H20...O1 and C16–H16C...O1 interactions contribute to growth along (100) direction (Fig. 4a), while C8–H8A...O4, C8–H8B...O1 and C14–H14...O3 interactions contribute to growth along the (010) and (001) directions (Fig. 4b). Additionally, for **CQ03**, C15–H15...O1 and C14–H14...O1 interactions contribute to stability along the (100) direction (Fig. 5c) and C8–H8A...Cl1, C19–H19...Cl1 and C19–H19...O2 interaction contribute to growth along the (010) direction (Fig. 5a), while C1–H1...Cg1, C12–H12...O3 and C9–H9...O3 interactions contribute to growth along the (100) direction (Fig. 5b). For **CQ04**, C19–H19...O1 and C15–H15...Br1 interactions contribute to growth and stability along the (100) and (010) directions (Fig. 6a), while C8–H8A...O2, C8–H8B...O1 and C9–H9...O1 interactions contribute to growth along the (001) direction (Fig. 6b).

Note that, for the four compounds analyzed, the molecular arrangements are stabilized only by weak interactions (Table 2), but in the literature we find several authors describing these interactions as considerable for the stability

of the supramolecular arrangement, in addition, several authors have linked certain interactions, such as C–H... π , to the enhancement of specific biological activities.^{82–87} Additionally, C–H...O interactions have been identified as crucial for antibacterial targets⁸⁸ and for stabilizing complexes associated with antioxidant activity.⁸⁹

The contacts were quantified by type through the two-dimensional 2D fingerprints obtained from Hirshfeld surfaces using CrystalExplorer software.⁹⁰ Hirshfeld surfaces are constructed based on electron density to determine regions of possible intermolecular interactions.^{61,62} In constructing these surfaces, the distances from the internal atoms to the surface (d_i) and from the external atoms to the surface (d_e) are considered. These distances are incorporated into the equations that define the Hirshfeld surface, allowing for detailed visualization and analysis of intermolecular interactions.^{61,62} The 2D fingerprints generated from the Hirshfeld surfaces show pseudo-mirrored peaks at $1.0 (d_e, d_i) - 1.2 (d_e, d_i)$, suggesting contacts involving hydrogen atoms (H...H). For all four compounds analyzed, these interactions are predominant owing to the organic nature of

Table 1. Crystallographic data and structure refinement for QC01, QC02, QC03 and QC04.

	QC01	QC02	QC03	QC04
Formula	C ₂₃ H ₂₁ NO ₄ S	C ₂₂ H ₁₉ NO ₄ S	C ₂₁ H ₁₆ NO ₃ SCI	C ₂₁ H ₁₆ NO ₃ SBr
Formula weight	407.47 g mol ⁻¹	393.44 g mol ⁻¹	397.86 g mol ⁻¹	442.32 g mol ⁻¹
Temperature	120 (2) K	120 (2) K	120 (2) K	120 (2) K
Wavelength	0.71073	0.71073 Å	0.71073 Å	0.71073 Å
Crystal system	Monoclinic	Monoclinic	Triclinic	Monoclinic
Space group, Z	P2 ₁ /c, 4	P2 ₁ /c, 4	P $\bar{1}$, 2	P2 ₁ /c, 4
Unit cell dimensions	a = 9.884 (2) Å b = 23.047 (5) Å c = 8.8652 (18) Å α = 90° β = 101.763 (4)° γ = 90°	a = 10.0696 (14) Å b = 21.739 (3) Å c = 8.6674 (12) Å α = 90° β = 104.625 (2)° γ = 90°	a = 8.2750 (18) Å b = 10.011 (2) Å c = 11.853 (3) Å α = 65.602 (3)° β = 79.928 (3)° γ = 77.577 (3)°	a = 10.3865 (16) Å b = 19.972 (3) Å c = 8.9477 (14) Å α = 90° β = 103.249 (2)° γ = 90°
Volume	1977.0 (7) Å ³	1835.8 (4) Å ³	869.2 (3) Å ³	1806.7 (5) Å ³
Calculated density	1.369 g m ⁻³	1.424 g m ⁻³	1.520 g m ⁻³	1.626
Absorption coefficient	0.194 mm ⁻¹	0.206 mm ⁻¹	0.363 mm ⁻¹	2.413 mm ⁻¹
Goodness-of-fit (S)	1.027	1.008	1.039	1.050
Final R indices [I > 2σ(I)]	R ₁ = 0.0520 wR ₂ = 0.1164	R ₁ = 0.0477 wR ₂ = 0.0958	R ₁ = 0.0322 wR ₂ = 0.0792	R ₁ = 0.0237 wR ₂ = 0.0573
R indices (all data)	R ₁ = 0.0783 wR ₂ = 0.1311	R ₁ = 0.0896 wR ₂ = 0.1108	R ₁ = 0.0398 wR ₂ = 0.0833	R ₁ = 0.0297 wR ₂ = 0.0596

the molecules (QC01, 49.7%; QC02, 44.3%; QC03, 31.3%; and QC04, 32.3%) (Fig. 7). Peaks around 1.2–1.4 (d_o) with 1.2–1.4 (d_i) indicate non-classical interactions (C–H...O), which were identified by geometric parameters in all structures and quantified at around 22%, followed by C–H... π interactions with similar quantifications. For compound QC03, C–H...Cl interactions were quantified at 13.4%, and for QC04, interactions involving Br were quantified at 12.7%, corroborating the interactions described by geometric parameters. These interactions represented over 90% of the total structural stability for all molecules.

Molecular modeling analysis

The theoretical calculations through DFT were performed for both the conformations of the molecules in the solid state and the conformations of the molecules in the gas phase. These theoretical parameters provide a more comprehensive comparison of the relationship between the chemical structures of these quinolinone–chalcones and their properties.⁹¹ The conformation of these molecules in different phases provides insights and can support future research as well as biological and industrial applications.⁹² The representation of the LUMO and HOMO orbital arrangements for molecules QC01, QC02, QC03 and QC04 is shown for the solid state [Fig. 8 (a)] and for the gas phase

[Fig. 8 (b)]. For the analyzed molecules, the HOMO orbital is characterized as the bonding orbital with nucleophilic properties, while the LUMO orbital is characterized as the antibonding orbital with electrophilic properties, with energy values indicated in the figure. The difference between these energy values (Egap) can be an important indicator of certain molecular properties, as higher Egap values may suggest greater kinetic stability, higher excitation energies, and lower chemical reactivity.⁹³ For the solid state, the Egap values are: QC03 (657.40 kJ mol⁻¹) > QC04 (655.59 kJ mol⁻¹) > QC02 (623.74 kJ mol⁻¹) > QC01 (615.60 kJ mol⁻¹), while for the gas phase: QC03 (662.65 kJ mol⁻¹) > QC04 (657.14 kJ mol⁻¹) > QC02 (620.69 kJ mol⁻¹) > QC01 (618.51 kJ mol⁻¹). Note that for both phases, the Egap values are similar, with QC03 having the highest Egap and QC01 the lowest.

The MEP map is an important physicochemical tool that contributes to predicting reactive sites targeted in chemical reactions. It was calculated for both the conformations of the molecules in the solid state and the conformations of the molecules in the gas phase (Fig. 9). The MEP map illustrates that the most negative region (color red) is situated on the oxygen atoms of the sulfonamide and carbonyl group (for both states, solid and gas phase) (Fig. 9). Additionally, positive regions (color blue) are close to the hydrogen atoms and the aromatic ring attached to the sulfonamide group. Based on the analysis of the MEP map and the interactions

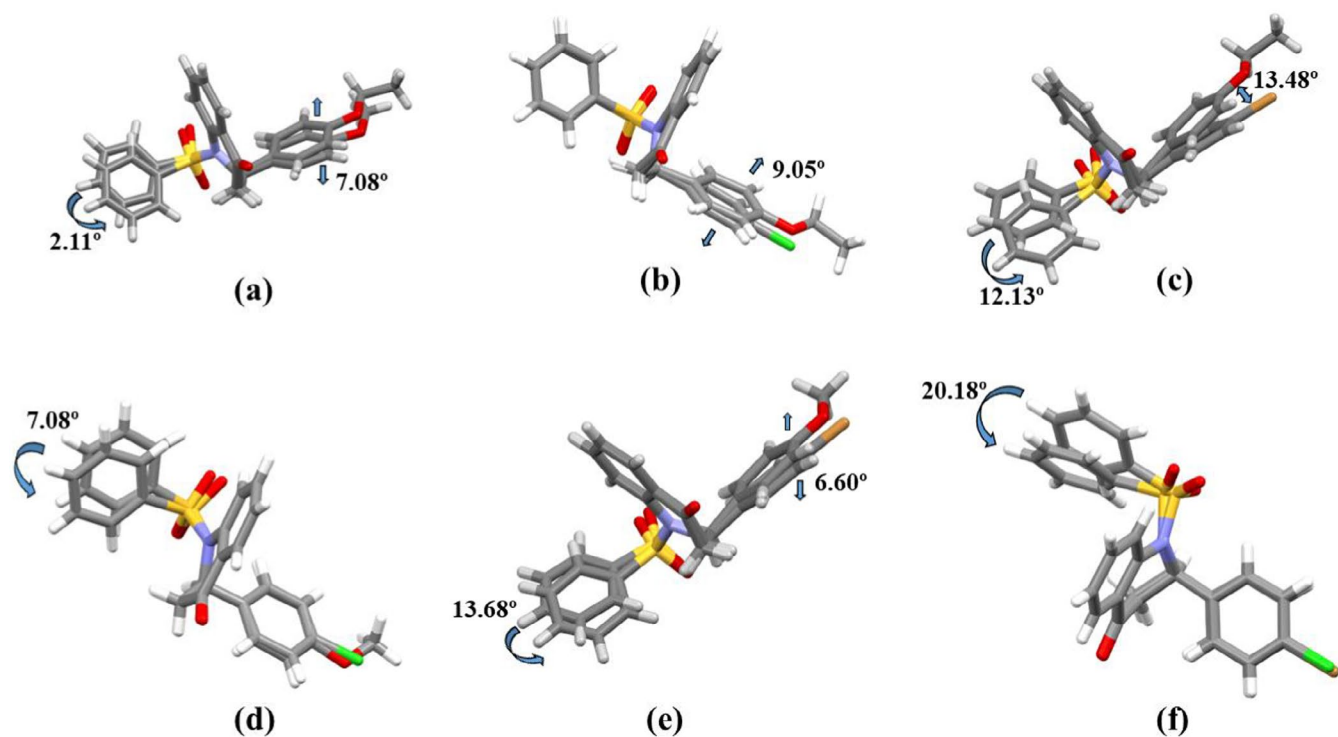


Figure 2. Overlap between compounds **CQ01** with **CQ02** (a), **CQ01** with **CQ03** (b), **CQ01** with **CQ04** (c), **CQ02** with **CQ03** (d), **CQ02** with **CQ04** (e), **CQ03** with **CQ04** (f).

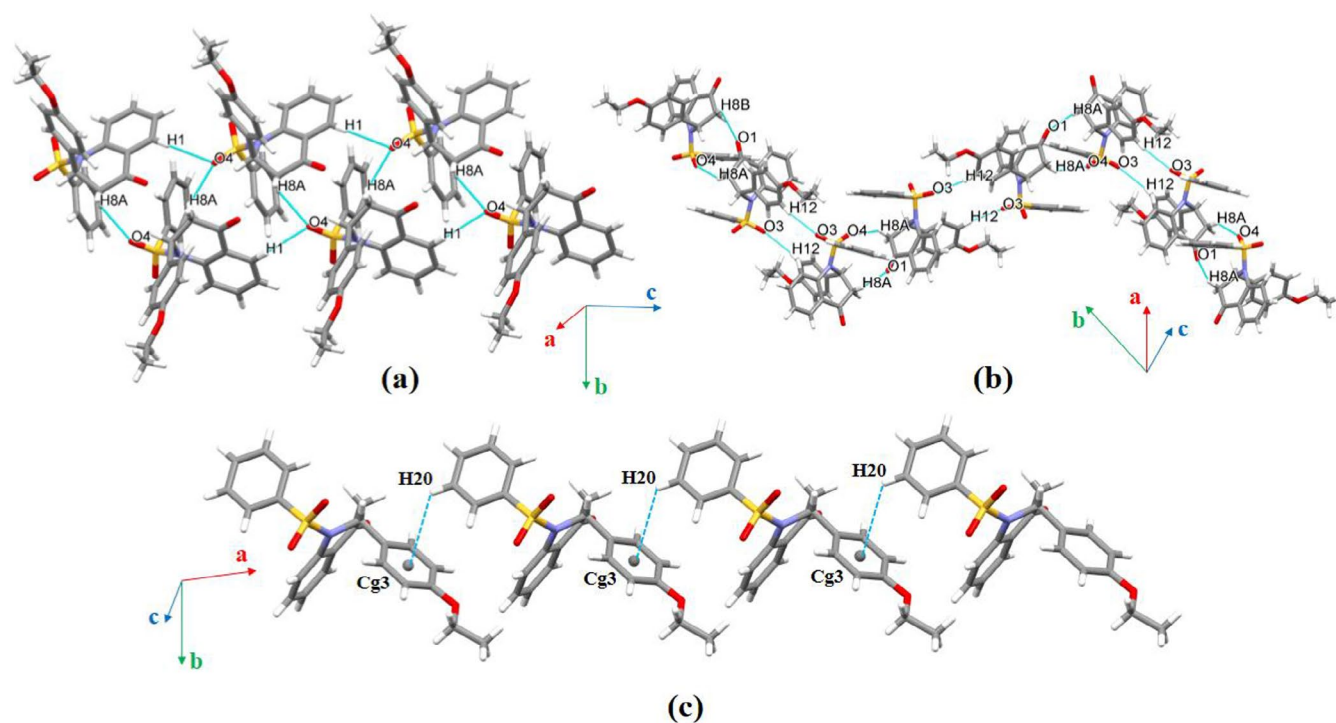


Figure 3. Representations for **QC01** interactions. C1–H1...O4 and C8–H8A...O4 interactions (a); C8–H8B...O1, C8–H8A...O4, C12–H12...O3, C8–H8A...O1 interactions (b); and C20–H20...Cg3 interaction (c). (a-axis (100); b-axis (010); c-axis (001)).

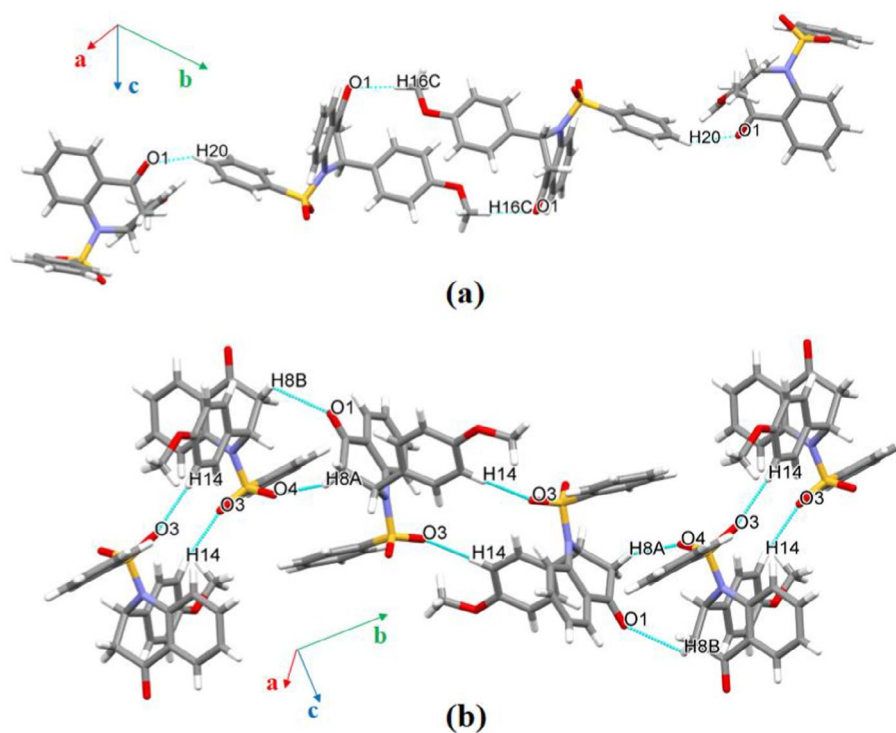


Figure 4. Representations for **QC02** interactions. C16–H16C...O1 and C20–H20...O1 interactions (a); C8–H8B...O1, C8–H8A...O4 and C14–H14...O3 interactions (b). (a-axis (100); b-axis (010); c-axis (001)).

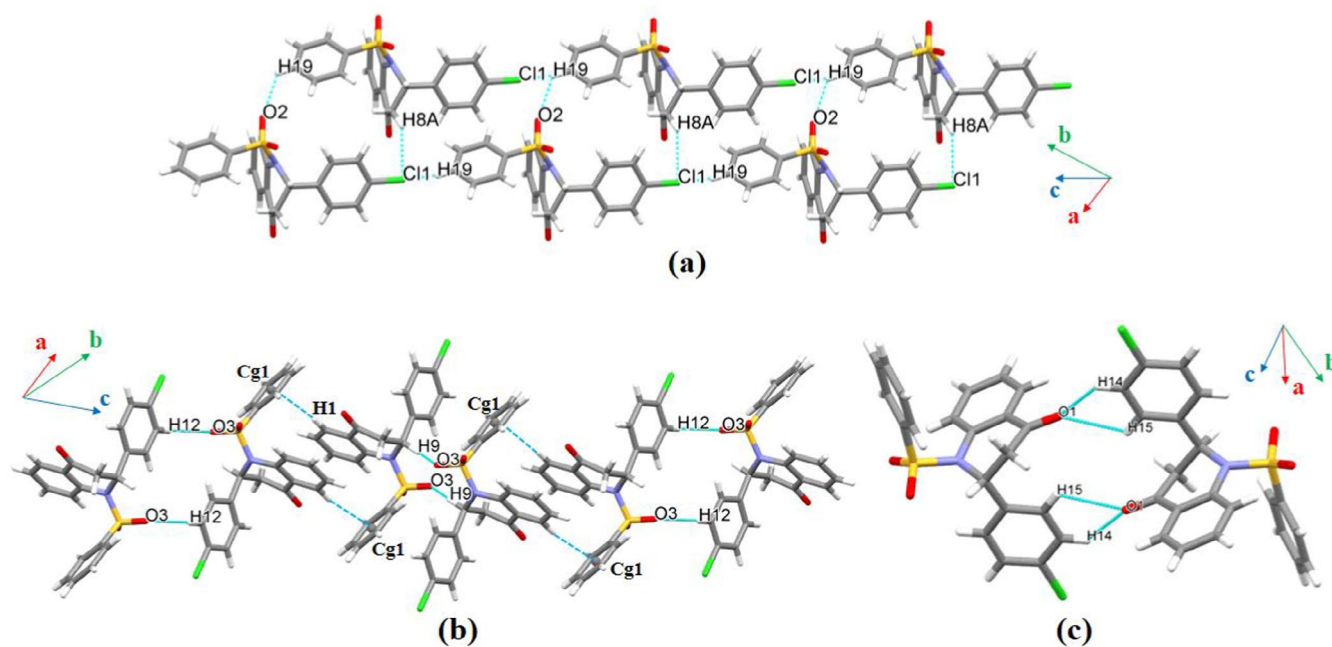


Figure 5. Representations for **QC03** interactions. C19–H19...O2, C8–H8A...Cl1 and C19–H19...Cl1 interactions (a); C1–H1...Cg1, C9–H9...O3 and C12–H12...O3 interactions (b); C14–H14...O1 and C15–H15...O1 interactions (c). (a-axis (100); b-axis (010); c-axis (001)).

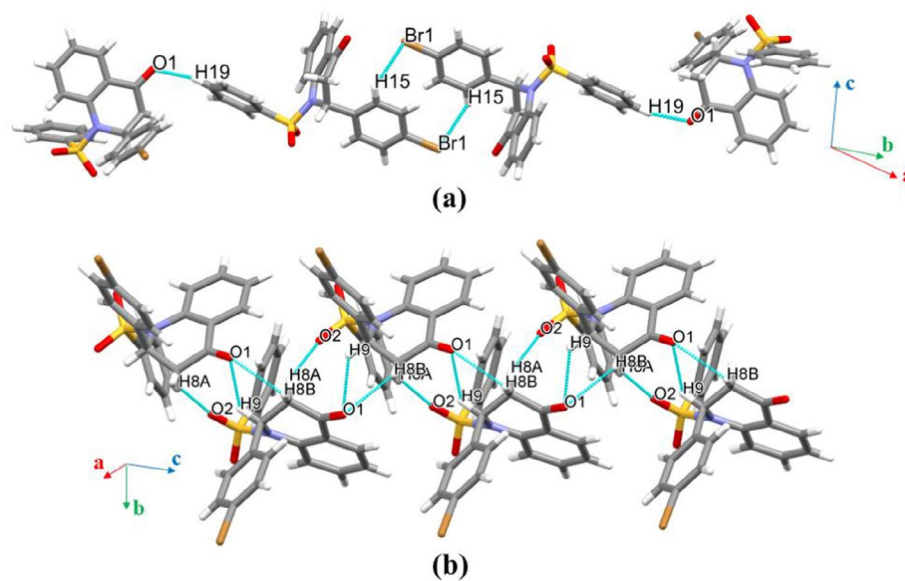


Figure 6. Representations for **QC04** interactions. C19–H19...O1 and C15–H15...Br1 interactions (a); C9–H9...O1, C8–H8B...O1 and C8–H8A...O2 interactions (b). (a-axis (100); b-axis (010); c-axis (001)).

Table 2. Molecular interactions for QC01, QC02, QC03 and QC04.						
QC	Interaction	d (D–H) Å	d (H...A) Å	d (D–A) Å	d (D–H–A) (°)	Symmetry code
01	C1–H1...O4	0.950 (1)	2.707 (5)	3.337 (6)	124.41 (6)	$x, y, -1+z$
	C8–H8A...O4	0.990 (1)	2.683 (6)	3.534 (5)	144.16 (7)	$x, 1/2-y, -1/2+z$
	C8–H8B...O1	0.991 (1)	2.616 (6)	2.956 (6)	100.11 (5)	$x, 1/2-y, -1/2+z$
	C12–H12...O3	0.950 (1)	2.462 (5)	3.394 (5)	166.93 (5)	$1-x, 1-y, -z$
	C8–H8A...O1	0.990 (1)	2.662 (6)	2.956 (6)	97.32 (6)	$x, 1/2-y, 1/2+z$
	C20–H20...Cg3	0.950 (1)	3.375 (5)	4.019 (5)	127.04 (6)	$1+x, y, z$
02	C20–H20...O1	0.950 (1)	2.634 (6)	3.412 (5)	139.44 (6)	$-1+x, 1/2-y, -1/2+z$
	C16–H16C...O1	0.980 (1)	2.649 (6)	3.553 (6)	153.45 (5)	$-x, 1-y, 1-z$
	C8–H8B...O1	0.990 (1)	2.661 (5)	3.084 (5)	105.97 (6)	$x, 1/2-y, -1/2+z$
	C8–H8A...O4	0.990 (1)	2.527 (7)	3.437 (7)	152.82 (7)	$x, 1/2-y, -1/2+z$
	C14–H14...O3	0.950 (1)	2.447 (6)	3.328 (6)	154.25 (5)	$1-x, 1-y, 2-z$
03	C19–H19...O2	0.950 (1)	2.695 (4)	3.331 (4)	124.86 (7)	$1+x, y, z$
	C8–H8A...Cl1	0.990 (1)	2.896 (5)	3.772 (6)	147.93 (6)	$1+x, y, z$
	C19–H19...Cl1	0.950 (1)	2.891 (6)	3.467 (7)	120.12 (7)	$-1+x, -1+y, z$
	C14–H14...O1	0.950 (1)	2.648 (4)	3.252 (6)	121.97 (4)	$1-x, -y, 1-z$
	C15–H15...O1	0.950 (1)	2.691 (5)	3.279 (4)	120.67 (7)	$1-x, -y, 1-z$
	C12–H12...O3	0.950 (1)	2.672 (5)	3.590 (6)	162.76 (6)	$-x, 1-y, -z$
	C9–H9...O3	1.000 (1)	2.515 (6)	3.347 (7)	140.75 (7)	$1-x, 1-y, -z$
	C1–H1...Cg1	0.950 (1)	3.171 (7)	3.586 (6)	125.69 (6)	$1-x, 1-y, 1-z$
04	C19–H19...O1	0.950 (1)	2.469 (6)	3.323 (6)	149.61 (6)	$-1+x, 1/2-y, -1/2+z$
	C15–H15...Br1	0.950 (1)	3.353 (5)	3.664 (7)	101.61 (5)	$2-x, 1-y, 1-z$
	C8–H8A...O2	0.991 (1)	2.401 (4)	3.300 (7)	150.74 (5)	$x, 1/2-y, -1/2+z$
	C9–H9...O1	1.000 (1)	2.567 (5)	3.245 (6)	124.92 (6)	$x, 1/2-y, -1/2+z$
	C8–H8B...O1	0.990 (1)	2.659 (6)	3.082 (5)	105.93 (5)	$x, 1/2-y, -1/2+z$

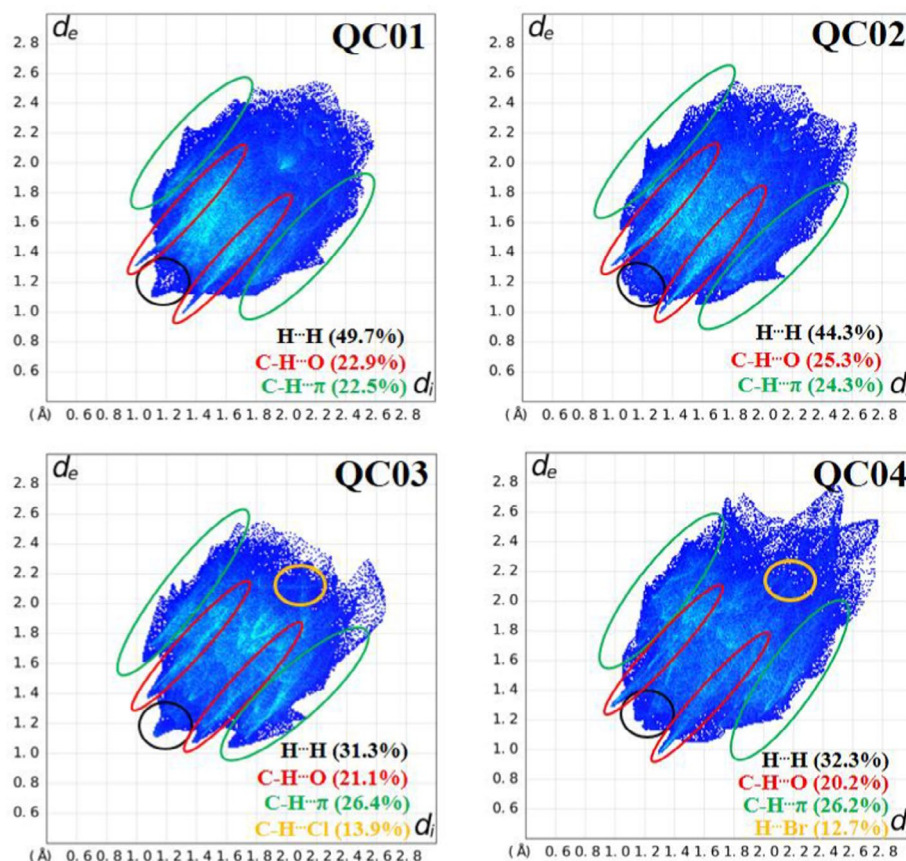


Figure 7. 2D Fingerprints with percentages of the contact types for **QC01**, **QC02**, **QC03** and **QC04**.

observed in the aromatic ring (attached to the sulfonamide group) in the crystal structures, the possibility of a nucleophilic attack occurring in this region can be suggested.

Machine learning analysis

The reaction rate constant (k_{OH}) is the main kinetic parameter for assessing a compound's degradation efficiency *via* hydroxyl radical attack, for which a high k_{OH} value indicates faster oxidation. Here, we examined the highest AD% values for the ML model (XGBoost) and molecular descriptor, with the MACCS fingerprint returning AD% values above 59% for all tested molecules. This suggests that MACCS provides more accurate predictions than the Morgan fingerprint using the *pySiRC* tool. Table 3 presents k_{OH} values for the main diesel component ($1.14 \times 10^{10} \text{ M}^{-1} \text{ s}^{-1}$) and three biodiesel components – BD M9OD ($5.94 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$), BD MPAL ($4.98 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$) and BD M8OD ($5.94 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$), respectively.

Taking the **QC01**, **QC02**, **QC03** and **QC04** (this work) as the reference molecules, their molecular similarity was evaluated using the *Tanimoto* index,⁷⁷ calculated over the

Morgan Fingerprint of 2048 bits⁹⁶ by using the RDKit,⁹⁷ against some target compounds, including the quinolinone–chalcones QC,⁶ CQH,⁹⁵ arylsulfonamide chalcones NSC and HSC,⁹⁴ chalcones Chal01 and Chal05,⁴¹ a tri-methoxy chalcone (TMC20)⁴⁵ and commercial additives (*CA: BHT, TBHQ, BHA, PG, PY and GA,⁷⁹ see Table S2 in the supplementary information). The lower similarity observed against the *CA is within 0.035–0.051 for the quinolinone derivatives with respect to BHT, and for our previous molecules is within 0.057–0.18 to TMC20. This was done focusing on biodiesel additive applications in standardized tests like Rancimat EN 15751:2014⁹⁸ and combustion heat (ASTM D4809).⁹⁹

The values of k_{OH} for the quinolinone–chalcones in the present study are $6.73 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for **QC01**, $5.45 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for **QC02**, $4.89 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for **QC03**, and $7.97 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for **QC04**. These can be compared with oxidation rates previously published by our group,⁶ ranging from a value, close to the present ones, of $7.44 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$, to a k_{OH} equal to $1.07 \times 10^{10} \text{ M}^{-1} \text{ s}^{-1}$ for the promising trimethoxy chalcone (TMC20).⁴⁵ Further comparison can be made to four arylsulfonamide chalcones, with k_{OH} equal to

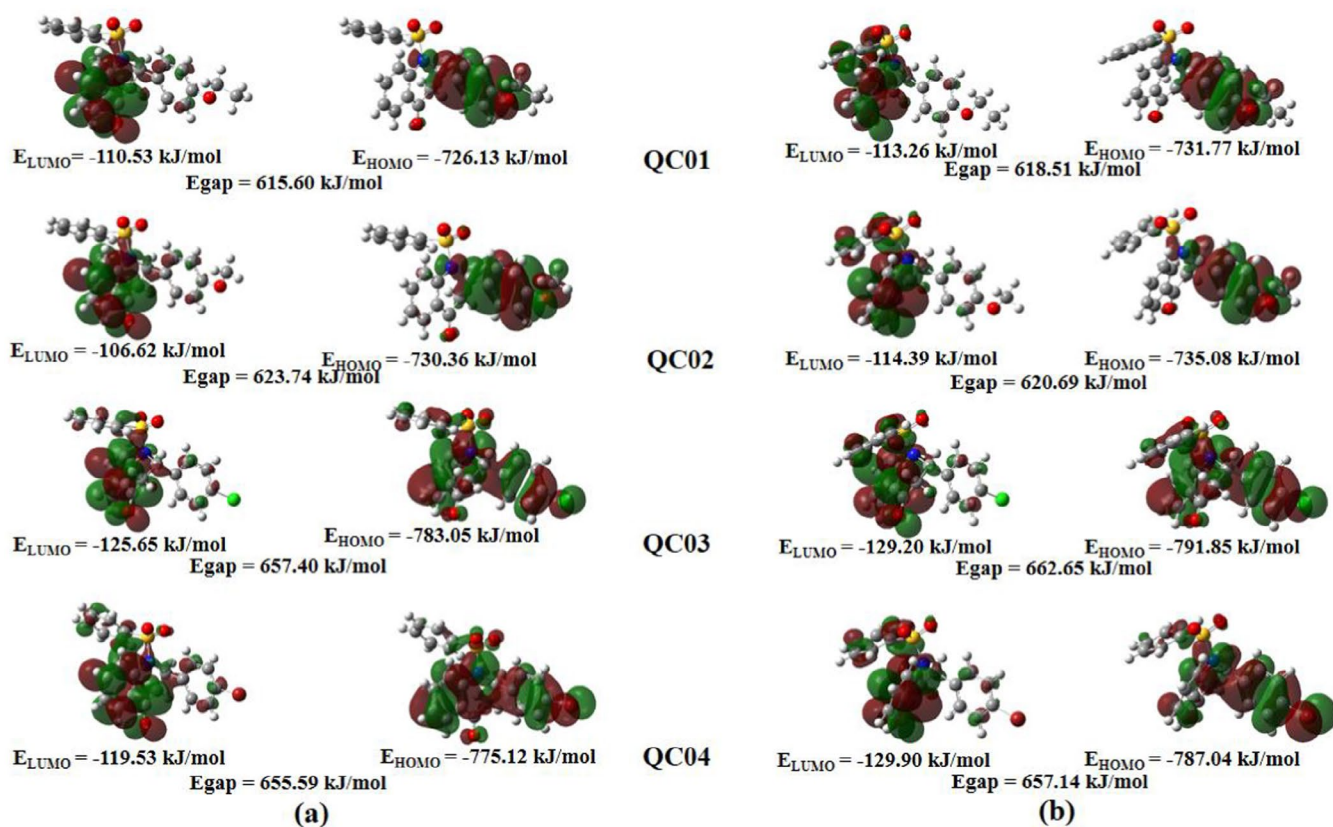


Figure 8. The frontier molecular orbitals for **QC01**, **QC02**, **QC03** and **QC04**. Solid state (a) and gas phase (b).

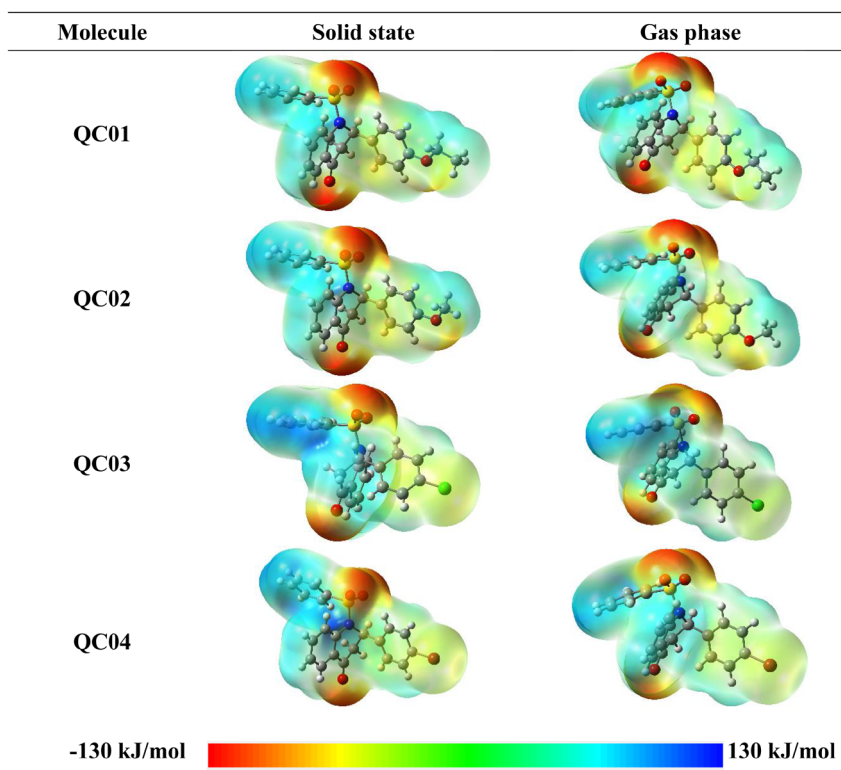


Figure 9. The MEP map surfaces for **QC01**, **QC02**, **QC03** and **QC04**.

Table 3. The reaction rate (k_{OH}) for the dihydroquinolin-4(1H)-one compounds QC01, QC02, QC03, QC04 (this work). Diesel and biodiesel (BD) are represented by their majority compound, respectively. Previously studied compounds, and other commercial additives (*CA). The values were obtained with the ML model XGBoost and separated into two types of molecular fingerprints Morgan and MACCS. AD is the percentage similarity within the applicability domain.

Molecule	Reaction rate coefficient (k_{OH}) ($\text{M}^{-1} \text{s}^{-1}$)				Reference
	Morgan	AD (%)	MACCS	AD (%)	
Diesel ⁷⁸	6.05×10^9	50	1.14×10^{10}	76.92	Duarte <i>et al.</i> ⁴¹
BD M9OD ⁷⁸	5.90×10^9	48.21	5.94×10^9	85	Duarte <i>et al.</i> ⁴¹
BD MPAL ⁷⁸	5.70×10^9	58.7	4.98×10^9	89.47	Duarte <i>et al.</i> ⁴¹
BD M8OD ⁷⁸	5.90×10^9	40.32	5.94×10^9	85	Duarte <i>et al.</i> ⁴¹
QC01	6.98×10^9	18.57	6.73×10^9	61.73	This work
QC02	7.51×10^9	21.88	5.45×10^9	66.23	This work
QC03	5.18×10^9	20.31	4.89×10^9	59.49	This work
QC04	4.33×10^9	20.63	7.97×10^9	59.49	This work
QC ⁶	3.82×10^9	18.42	7.44×10^9	62.82	This work
NSC ⁹⁴	2.25×10^{10}	21.43	9.29×10^9	60.42	This work
HSC ⁹⁴	2.81×10^{10}	21.69	9.29×10^9	60.42	This work
CQH ⁹⁵	4.99×10^9	15.48	4.20×10^9	61.90	This work
Chal01 ⁴¹	4.60×10^{10}	20.29	1.06×10^{10}	70.31	Duarte <i>et al.</i> ⁴¹
Chal05 ⁴¹	1.24×10^{10}	24	8.17×10^9	62.71	Duarte <i>et al.</i> ⁴¹
TMC20 ⁴⁵	9.40×10^9	33.93	1.07×10^{10}	80	Duarte <i>et al.</i> ⁴¹
*CA BHT ⁷⁹	4.16×10^9	100	4.34×10^9	100	Duarte <i>et al.</i> ⁴¹
*CA TBHQ ⁷⁹	7.57×10^9	100	7.63×10^9	100	Duarte <i>et al.</i> ⁴¹
*CA BHA ⁷⁹	7.27×10^9	66.67	4.31×10^9	77.78	Duarte <i>et al.</i> ⁴¹
*CA PG ⁷⁹	1.11×10^{10}	100	1.22×10^{10}	100	Duarte <i>et al.</i> ⁴¹
*CA PY ⁷⁹	7.06×10^9	40.74	1.02×10^{10}	100	Duarte <i>et al.</i> ⁴¹
*CA GA ⁷⁹	4.09×10^9	40.62	1.48×10^9	100	Duarte <i>et al.</i> ⁴¹

$1.06 \times 10^{10} \text{ M}^{-1} \text{ s}^{-1}$ for Chal01, to $8.17 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for Chal05,⁴¹ to $8.17 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for both NSC and HSC.⁹⁴ Passing from TMC20 to Chal01, NSC, HSC, Chal05, QC04, QC and QC01 the oxidation potential is decreasing, but always higher than or equivalent to those obtained for both diesel and BD reference molecules (Table 3).

The k_{OH} rate coefficients were also obtained for additive compounds, previously reported, with the higher values being observed, in increasing order, for TBHQ, $7.63 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$, PY, $1.02 \times 10^{10} \text{ M}^{-1} \text{ s}^{-1}$ and PG, $1.22 \times 10^{10} \text{ M}^{-1} \text{ s}^{-1}$, while values close to the ones in present work are observed for BHT, $4.34 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$, BHA $4.31 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ alongside with $1.48 \times 10^9 \text{ M}^{-1} \text{ s}^{-1}$ for GA, which shows the lower oxidative potential (Table 3). Thus, the rate coefficients obtained for GA, BHT, BHA and CQH are even slightly below the reference molecules. These values suggest that oxidative degradation for the commercial additives (*CA), quinolinone–chalcones, tri-methoxy chalcone and arylsulfonamide chalcones, ordered by decreasing

oxidation potential (PG > TMC20 > Chal01 > PY > NSC and HSC > Chal05 > QC04, QC and QC01), is higher than for diesel and biodiesel references. This aligns with experimental results through the Rancimat method reported by the authors.^{41,45} Compared with existing additives, these oxidative rates are generally similar, suggesting that quinolinone–chalcone derivatives could be promising candidates to slow biodiesel degradation, suitable for enhancing its oxidative stability.

All these results indicate the potential of these compounds as possible additives; however, some aspects still need to be addressed and further investigated. The addition of antioxidants to biodiesel can enhance its oxidative stability, reducing the formation of harmful byproducts and extending its shelf life, which would minimize pollutant emissions and logistical costs.^{100,101} However, considering environmental factors, the biodegradability and toxicity of additives must be carefully evaluated to prevent negative impacts. From an economic perspective, effective additives would increase

the commercial viability of biodiesel and diesel (which can contain biodiesel additives according to national regulations) while also reducing storage losses.¹⁰² Furthermore, their compatibility with existing infrastructure is crucial to avoiding high adaptation costs, given that biodiesel can be used in diesel cycle engines with certain modifications.¹⁰⁰ Thus, the development of sustainable and cost-effective additives can make biodiesel more competitive and strengthen its adoption in the global energy matrix.

Conclusions

This work presented the synthesis and characterization of four new quinolinone–chalcone derivatives (**QC01**, **QC02**, **QC03** and **QC04**), demonstrating their potential applications as antioxidant additives in biofuels. These compounds are unprecedented in the literature, ensuring originality in their application as additives. Structural and supramolecular analyses revealed that different substitutions (Br, Cl, OCH₃ and OCH₂CH₃) on the quinolinone–chalcone influence intermolecular interactions. Specifically, C–H · O and C–H · π interactions were found to contribute significantly to the stability of the crystalline structures, and these interactions are consistent with previously reported biological activities in the literature. Additionally, theoretical calculations based on DFT for both solid and gas phases provided detailed insights into the distribution of molecular orbitals and their corresponding energies. **QC03** was found to have the highest *E*_{gap}, suggesting greater kinetic stability, while **QC01** exhibited the lowest *E*_{gap}, reflecting higher chemical reactivity.

Reaction rate coefficients (*k*_{OH}) showed that the studied compounds are effective in scavenging hydroxyl radicals, with **QC04** standing out by displaying a *k*_{OH} value comparable with those of commercial antioxidants such as BHT. Therefore, the quinolinone–chalcone derivatives synthesized in this study are promising candidates for slowing down oxidative degradation in biofuels. Future research could explore the detailed mechanisms of their antioxidant action and further evaluate the experimental aspects of these compounds, aiming at their application as biofuel additives.

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Conflict of interest statement

The authors declare no conflicts of interest.

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