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%V_{Bur} index and steric maps: from predictive catalysis to machine learning

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Steric indices are parameters used in chemistry to describe the spatial arrangement of atoms or groups of atoms in molecules. They are important in determining the reactivity, stability, and physical properties of chemical compounds. One commonly used steric index is the steric hindrance, which refers to the obstruction or hindrance of movement in a molecule caused by bulky substituents or functional groups. Steric hindrance can affect the reactivity of a molecule by altering the accessibility of its reactive sites and influencing the geometry of its transition states. Notably, the Tolman cone angle and %V_{Bur} are prominent among these indices. Actually, steric effects can also be described using the concept of steric bulk, which refers to the space occupied by a molecule or functional group. Steric bulk can affect the solubility, melting point, boiling point, and viscosity of a substance. Even though electronic indices are more widely used, they have certain drawbacks that might shift preferences towards others. They present a higher computational cost, and often, the weight of electronics in correlation with chemical properties, e.g. binding energies, falls short in comparison to %V_{Bur}. However, it is worth noting that this may be because the steric index inherently captures part of the electronic content. Overall, steric indices play an important role in understanding the behaviour of chemical compounds and can be used to predict their reactivity, stability, and physical properties. Predictive chemistry is an approach to chemical research that uses computational methods to anticipate the properties and behaviour of these compounds and reactions, facilitating the design of new compounds and reactivities. Within this domain, predictive catalysis specifically targets the prediction of the performance and behaviour of catalysts. Ultimately, the goal is to identify new catalysts with optimal properties, leading to chemical processes that are both more efficient and sustainable. In this framework, %V_{Bur} can be a key metric for deepening our understanding of catalysis, emphasizing predictive catalysis and sustainability. Those latter concepts are needed to direct our efforts toward identifying the optimal catalyst for any reaction, minimizing waste, and reducing experimental efforts while maximizing the efficacy of the computational methods.

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1. Introduction

In modern times, the field of chemistry has introduced synthetic catalysts with vast potential for numerous applications. Such catalysts are key in addressing the pressing challenge of energy demands,^{1,2} in particular, sustainable sources, as well as the removal of toxic gases.³ However, throughout history, natural catalysts like enzymes have been employed unknowingly by humans to produce food and beverages. While nature has had billions of years to evolve highly efficient enzymes for essential reactions,⁴ chemists have a shorter timeframe to

develop efficient catalysts. In both natural and synthetic contexts,⁵ successful molecular catalysts are built by designing functional catalytic pockets,^{6,7} specific regions where reactions occur. These pockets are created by arranging atoms in three-dimensional space. On the other hand, in synthetic chemistry, two primary strategies for catalyst development exist: high-throughput screening of a wide catalyst range,⁸ and informed design with iterative enhancements.^{9,10} Moreover, the main challenge in catalyst design involves refining ineffective prototypes into systems meeting stringent industrial standards. Even slight changes in a catalyst's active site can significantly alter its performance. Achieving selectivity often means favouring one reaction pathway by only 1–2 kcal mol⁻¹, which is less than a hydrogen bond.¹¹ A small drop in activation energy by the same measure can greatly boost catalytic activity.

In the complex realm of transition metal complexes, choosing the most effective catalyst presents a formidable challenge, typically addressed through trial and error or intuition rather

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than methodical reasoning. A conventional approach to tackle this issue involves employing molecular descriptors that can systematically rearrange the catalyst space, as illustrated in Fig. 1a.¹² This methodology facilitates the virtual design of novel catalysts. The purpose of these descriptors is to establish quantitative connections between characteristics of the catalytic environment (akin to the concept of enzymatic catalysis, referring to the region surrounding the active metal centre) and its observed experimental behaviour.

The categorization of catalyst behaviour using molecular descriptors,¹³ like steric parameters and maps or any measure related to electronics, can expedite the design of improved catalysts. Particularly, concentrating on descriptors capable of capturing the shape of catalytic pockets, those simple tools can serve as distinct identifiers for characterizing transition metal

catalysts and, potentially, metalloproteins, extending their application to biocatalysis.¹⁴ In chemistry, steric indices quantify the spatial bulk of molecules or their constituent groups. They measure alterations in molecular characteristics arising from repulsive forces, typically van der Waals interactions, but also others, between different segments of a molecule. These indices are useful in predicting the reactivity, stability, and physical properties of chemical compounds. Even though there are indices also related to energetics, like molecular electrostatic potential^{15,16} or repulsive energy,^{17–19} the most common steric indices used in chemistry are the steric hindrance index (SHI), the van der Waals volume (VdW), and the Tolman cone angle (CA).²⁰ First, the SHI is a measure of the degree to which a functional group or atom in a molecule is blocked or obstructed by adjacent substituents. It is calculated by comparing the experimental bond angles of a molecule with the ideal bond angles for the same molecule with no steric hindrance, easy to visualize for phosphane ligands.¹⁶ Higher SHI values indicate greater steric hindrance. Second, VdW is a measure of the size of a molecule or a functional group, and it is calculated based on the van der Waals radii of its constituent atoms. A higher VdW value indicates a larger spatial bulk of the molecule. Third, the CA is a measure of the steric hindrance caused by a substituent on a metal centre in organometallic chemistry (Fig. 1b). It is defined as the angle between two lines drawn from the metal centre to the outermost atoms of the substituent. A larger CA value indicates greater steric hindrance. Overall, steric indices are important tools in chemical research, especially in the design and synthesis of new molecules with desired properties. And the aim of this review is to stress the importance of the concepts of steric index of the buried volume of Nolan, Cavallo and coworkers,²¹ %V_{Bur} (Fig. 1c),²² and the associated steric maps of Cavallo and coworkers (Fig. 1d),^{12,14} and how they have been expanded to many applications,²³ mainly thanks to the creation of a user friendly web server.



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Albert Poater

Albert Poater finished the PhD in Chemistry in 2006, supervised by Prof Miquel Duran and Prof Miquel Solà, at the University of Girona. After periods in Chile with Prof Alejandro Toro-Labbé; in Montpellier with Prof Odile Eisenstein and a long postdoc in University of Salerno with Prof Luigi Cavallo, in 2010 he became an independent researcher as Ramón y Cajal in Girona, apart from Visiting Researcher at KAUST in Saudi Arabia, and at LCC-CNRS

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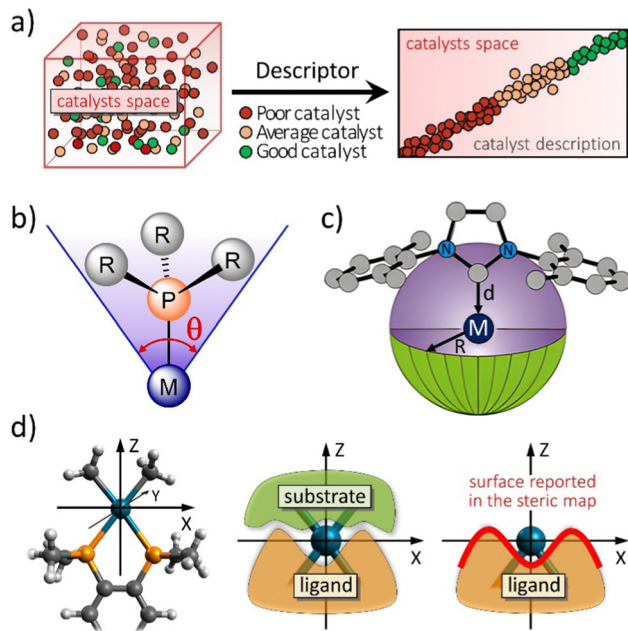


Fig. 1 (a) The challenge to choose the right catalyst from molecular descriptors. Representation of (b) the Tolman cone angle, (c) the buried volume descriptor, and (d) how the steric maps are built (plane XY) from the interaction between the ligand and the substrate. Adapted with permission.¹² Copyright © 2016, American Chemical Society.

The $\%V_{\text{Bur}}$ parameter is a steric index that has been found to be highly correlated with the binding energy of metal catalysts.^{22,24} This index was first introduced by Cavallo and coworkers in a series of papers in the early 2000s,^{25,26} and has since then been used extensively in the rational design of new catalysts and materials.

The $\%V_{\text{Bur}}$ parameter is a measure of the volume occupied by the ligands around a metal centre, taking into account the valence electrons of the ligands that are buried in the coordination sphere of the metal. It is calculated using quantum mechanical methods, such as density functional theory (DFT), and has been found to be a good indicator of the steric hindrance experienced by the ligands around the metal centre. One of the key applications of the $\%V_{\text{Bur}}$ parameter has been in the *de novo* design of metal catalysts for a range of chemical reactions. The steric hindrance around the metal centre can have a significant impact on the selectivity and activity of the catalyst,²⁷ and so a thorough understanding of these effects is essential for guided design purposes.

A number of studies have shown a high correlation between the $\%V_{\text{Bur}}$ parameter and the binding energy of ligands to metal centres, a key factor in determining the reactivity of metal catalysts.²⁸ In general, those studies found that ligands with larger $\%V_{\text{Bur}}$ values correspond to weaker binding to the metal centre, while ligands with smaller $\%V_{\text{Bur}}$ values result in stronger binding. This is translated to the reactivity of the catalysts.²⁹ These findings hold significant implications for the design of new catalysts. For example, if a ligand is required to have strong binding to the metal centre, then ligands with smaller $\%V_{\text{Bur}}$ values may be preferred. Conversely, if a ligand is

required to have weaker binding to the metal centre, then ligands with larger $\%V_{\text{Bur}}$ values may be more suitable. Interestingly, the metal does not influence significantly the $\%V_{\text{Bur}}$ parameter itself, but merely adjusts it based on the bonded ligands, and obviously on the radius with the linking atom of the ligand under study.³⁰ This is one of the strengths of the $\%V_{\text{Bur}}$ parameter, since it remains consistent regardless of the computational approach used on the metal, facilitating a uniform assessment of ligand–metal binding.

The scope of $\%V_{\text{Bur}}$ parameter covers the whole range of reactions, including hydrogenation,^{31,32} oxidation,^{33–35} cross-coupling reactions,³⁶ polymerization,^{29,37,38} among others.

To sum up, the $\%V_{\text{Bur}}$ parameter developed by Cavallo and coworkers has proven to be instrumental in the development of novel catalysts and materials. Its strong correlation with the binding energy of ligands to metal centres has allowed for the development of more efficient and selective catalysts, while its applications in materials science have led to the development of materials with specific properties. As research in this area continues to advance, it is likely that the $\%V_{\text{Bur}}$ parameter will gain even greater significance in the formulation of new catalysts and materials for a range of applications.

Despite the broad scope of the $\%V_{\text{Bur}}$ parameter and associated steric maps, as evidenced by over 2000 papers on these topics, this review focuses on their specific applications. To provide context, an overview of these applications is presented. Importantly, from the moment $\%V_{\text{Bur}}$ was initially developed, the aim was for it to serve not only as a descriptive tool but also as a predictive one. This predictive potential aligns with machine learning (ML) approaches, and ultimately with predictive catalysis.³⁹ The number of catalysts available is huge, creating a nearly infinite set to consider, and thus unaffordable for human analysis, *i.e.* how they could be used is not easy to order. Thus, integrating concepts like steric indices, ML, and predictive catalysis is essential to enhance past efforts and experience.⁴⁰

2. From catalysis to machine learning via predictive catalysis

After the pioneering work and ideas about computing machinery and intelligence of Alan Turing,⁴¹ the term machine learning was coined in 1959 by Arthur Samuel,⁴² from the American company IBM. In addition to artificial intelligence (AI), curiously he was also a pioneer in the field of computer games. Then during the 60s and 70s research was done on ML for pattern classification,⁴³ especially through the efforts of Duda and Hart.⁴⁴ In the 80s the horizons expanded towards neural networks. And currently, modern ML consists of the classification of data based on models which have been developed, and from which to make predictions for future outcomes.

In detail, ML is based on the study of computer algorithms that automatically improve (artificial intelligence) by using data and checking how previous decisions evolve.⁴⁵ Machine learning algorithms have the ability to build a model based on sample data, in order to make predictions or decisions without



solvation models are used to simulate the behaviour of molecules in solution, which is important for understanding biochemical processes and designing new materials; (13) Machine learning in computational chemistry: ML methods are being increasingly used in computational chemistry to develop more accurate models for predicting molecular properties and reactions; (14) Big data and data mining: large databases of chemical and biological data are being analysed using ML and data mining techniques to identify patterns and correlations that can be used to develop more accurate predictive models; (15) Artificial intelligence and deep learning: AI and deep learning techniques are being applied to computational chemistry to develop more efficient algorithms and models.

3. Chemical indices related to reactivity

Several indices have been developed and are used in chemistry to quantify various properties of molecules, including reactivity, stability, and physical properties. In the following sections, we will discuss some of the most important indices used in chemistry, with particular emphasis on the steric indices.

3.1. Steric indices

Very generally, the steric indices are measures of the spatial hindrance or crowding around a particular atom or group of atoms within a molecule. They may consider the size and shape of the substituent groups, the number of bonds, and the number of lone pairs of electrons surrounding the atom of interest. A higher steric index indicates more crowding and hindrance, which can affect the reactivity and physical properties of the molecule. The steric hindrance index, the van der Waals volume, and the cone angle are three commonly used steric indices. The steric hindrance index measures the degree to which a functional group or atom in a molecule is blocked or obstructed by adjacent substituents, the van der Waals volume index measures the size of a molecule or functional group, and the Tolman cone angle measures the steric hindrance caused by a substituent on a metal centre in organometallic chemistry.

3.2. Electronic indices

Indices that measure the electronic properties of molecules, which play a crucial role in determining their reactivity, stability, and physical properties. The electronegativity, electron affinity, and ionization potential are three important electronic indices. Electronegativity measures the tendency of an atom to attract electrons, electron affinity measures the energy released when an atom gains an electron, and ionization potential measures the energy required to remove an electron from an atom or molecule.

3.3. Aromaticity indices

While aromaticity is exclusive to certain molecular systems with ring structures and lacks a univocal characterization, its fundamental role is undeniable, especially given its significant

influence on molecular stability, reactivity, and electronic properties. Therefore, a variety of indices exist, which can be organized into the following groups: Energetic,⁸⁴ geometric,⁸⁵ magnetic,^{86–88} electronic,^{89,90} and reactivity-based measures.⁹¹

3.4. Spectroscopic indices

Indices that measure the spectral properties of molecules, which are important in identifying and characterizing chemical compounds. The infrared (IR) spectrum, the ultraviolet-visible (UV-Vis) spectrum, and the nuclear magnetic resonance (NMR) spectrum are three commonly used spectroscopic indices. The IR spectrum measures the absorption of infrared radiation by molecules, the UV-Vis spectrum measures the absorption of ultraviolet and visible radiation by molecules, and the NMR spectrum measures the magnetic properties of nuclei in molecules.

3.5. Other indices

Other important indices in chemistry include thermodynamic indices, kinetic indices, and thermophysical indices. Thermodynamic indices, such as entropy and enthalpy, are measures of the energy and heat changes in chemical reactions. Kinetic indices, such as reaction rate and activation energy, are measures of the speed and mechanism of chemical reactions. Thermophysical indices, such as melting point and boiling point, are measures of the physical properties of chemical compounds.

Overall, these indices play a crucial role in chemical research and are used in a wide range of fields, from drug development to materials science and catalysis. By providing important insights into the behaviour of chemical compounds, these indices help chemists design and synthesize new molecules with specific properties and improve our understanding of the fundamental principles of chemistry. To sum up, the above indices and others are essential tools in chemistry for quantifying various properties of molecules and compounds. Next, the most used indices are discussed with a brief description:

The acid dissociation constant, commonly abbreviated as pK_a , is a measure of the strength of an acid in solution. It is defined as the negative logarithm of the acid dissociation constant (K_a), which is the equilibrium constant for the dissociation of an acid in water. A lower pK_a value indicates a stronger acid, and a higher pK_a value indicates a weaker acid. The pK_a value is an important index in determining the pH of a solution and the reactivity of acids in chemical reactions.

Basicity, also known as the basic dissociation constant (pK_b), is a measure of the strength of a base in solution. It is defined as the negative logarithm of the basic dissociation constant (K_b), which is the equilibrium constant for the dissociation of a base in water. A higher pK_b value indicates a stronger base, and a lower pK_b value indicates a weaker base. The basicity of a compound is an important index in determining its ability to accept protons in chemical reactions.

The dipole moment, is a measure of the polarity of a molecule. It is defined as the product of the charge on each atom and the distance between them. A higher dipole moment



indicates a more polar molecule, which can affect its physical and chemical properties, such as solubility, reactivity, and melting point.

Electronegativity, is a measure of the ability of an atom to attract electrons towards itself in a chemical bond. It is a fundamental concept in chemistry and is used to explain various chemical phenomena, such as the polarity of molecules, the strength of chemical bonds, and the reactivity of chemical compounds. Electronegativity is measured on the Pauling scale, where higher values indicate higher electronegativity.

Magnetic aromaticity indices, being the Nucleus Independent Chemical Shift (NICS) the most extended in the general chemistry community.^{86,87} NICS measures the magnetic susceptibility at specific points in the space (ghost atoms) to assess aromaticity. However, some authors have warned about the limitations of NICS and its occasional misinterpretations.^{92,93} Therefore, they advocate for the importance of relying on multiple indices based on different aspects of aromaticity.

The Hammett acidity function, commonly abbreviated as H_0 , is a measure of the strength of an acid in organic chemistry. It is defined as the logarithm of the ratio of the equilibrium constants of two acids, one being a reference acid and the other being the acid of interest. The H_0 index is an important index in determining the reactivity of organic acids in chemical reactions since 1932.⁹⁴ By extension, in 1937 it was introduced an empirical equation relying on two parameters to establish a connection between reaction rates and equilibrium constants for reactions involving aromatic compounds,⁹⁵ that since then it is a reference in catalysis.^{96,97} and can also be found in combination with steric maps to combine the electronic and steric properties of the ligands.⁹⁸

Henry's law constant, is a measure of the solubility of a gas in a liquid. It is defined as the ratio of the concentration of the gas in the liquid to its partial pressure in the gas phase at equilibrium. The Henry's law constant is an important index in environmental chemistry and is used to determine the fate and transport of gases in the environment.

Hydrophobicity, is a measure of the tendency of a molecule or functional group to repel water. It is a crucial index in determining the solubility and stability of molecules in aqueous environments. Hydrophobicity is often quantified using various indices, such as the octanol–water partition coefficient ($\log P$), which measures the partitioning of a molecule between an organic solvent and water.

The melting point, is a measure of the temperature at which a solid substance melts and becomes a liquid. It is an important physical property of compounds and can be used to identify and characterize them. The melting point is affected by various factors, such as the strength and type of intermolecular forces between molecules, the molecular weight and shape, and the presence of impurities.

Polarizability, is a measure of the ability of an atom or molecule to form a temporary dipole moment in response to an external electric field. It is an important index in determining the reactivity and stability of molecules, particularly those involved in non-covalent interactions, such as van der Waals

forces. Polarizability is influenced by various factors, such as the size, shape, and electronic structure of the molecule.

The reactivity index, is a measure of the relative reactivity of a molecule towards a specific reaction or reagent. It is often used in computational chemistry and theoretical studies to predict the outcome of chemical reactions and to design new molecules with specific properties. The reactivity index is calculated from various parameters, such as the molecular orbital energies and electron densities.

Solubility, is a measure of the ability of a substance to dissolve in a solvent. It is an important index in determining the physical and chemical properties of compounds and is influenced by various factors, such as the nature of the solute and solvent, temperature, pressure, and the presence of other solutes. The solubility index is often quantified using various indices, such as the solubility product constant (K_{sp}) and the dissolution rate. In detail, K_{sp} is a measure of the solubility of a compound in water. It is defined as the equilibrium constant for the dissolution of a solid in water, thus, K_{sp} is an important index in determining the solubility and precipitation of compounds in solution.

Steric hindrance, is a measure of the obstruction of the movement or reaction of a molecule by the presence of other groups or atoms in its vicinity. It is an important index in determining the reactivity and stability of compounds, particularly those involved in organic reactions. Steric hindrance can be quantified using various indices as shown above and that will be in detail discussed in the next section.

The van der Waals radius, is a measure of the effective radius of an atom or molecule, which is defined as the distance between the nuclei of two atoms when they are in their most stable state. It is an important index in determining the strength of intermolecular forces and the solubility of molecules in polar solvents. van der Waals radius is influenced by various factors such as the size and shape of the atom or molecule and the nature of the intermolecular forces. VdW interactions are generally used to quantify the steric hindrance, as well.

4. Steric indices

There are several steric indices used in chemistry to measure the spatial bulk or volume of molecules or functional groups. These indices play an important role in predicting the behaviour of chemical compounds and in the design and synthesis of new molecules with specific properties. The most commonly used steric indices in chemistry are the steric hindrance index, the van der Waals volume, the cone angle and the buried volume ($\%V_{bur}$).

The steric hindrance index, is a measure of the degree to which a functional group or atom in a molecule is blocked or obstructed by adjacent substituents. Steric hindrance can affect the reactivity of a molecule by preventing or slowing down chemical reactions that involve the affected functional group or atom. The SHI is calculated by comparing the experimental



bond angles of a molecule with the ideal bond angles for the same molecule with no steric hindrance. The ideal bond angles are calculated based on the size and shape of the atoms and functional groups involved in the bond. A higher SHI value indicates greater steric hindrance, and therefore, lower reactivity. This steric index is a measure of the bulkiness of a molecule. It is calculated by counting the number of atoms that are close to each other in three-dimensional space, and is used to predict the stability and reactivity of a molecule. The SHI refers to a measure of the size of a molecule in relation to the molecular size of its neighbours. This SHI is used to describe the molecular size and shape of a molecule in a system, and it is an important parameter for the design of chemical reactions and for the study of molecular interactions. However, one application of this steric index is specially in drug design, where it is used to predict the pharmacological activity of a drug and its potential for adverse side effects. By determining the steric index of a drug molecule, researchers can predict how it will interact with its target protein and with other molecules in the body. This information is crucial in designing drugs that are effective and safe for human use.

The van der Waals volume, is a measure of the size of a molecule or a functional group. It is calculated based on the van der Waals radii of its constituent atoms. The VdW index is important in determining the shape and size of molecules, and it plays a key role in the development of drugs and other compounds with specific physical and chemical properties. A higher VdW value indicates a larger spatial bulk of the molecule. The VdW index is particularly useful in predicting the solubility, stability, and bioavailability of drugs.

The Tolman cone angle, is a measure of the steric hindrance caused by a substituent on a metal centre in organometallic chemistry.^{20,99} Mathematically, the CA is a measure of the shape of a molecule, calculated by determining the angle between the vector from the central atom to the centre of gravity of the molecule, and the vector from the central atom to the furthest point on the molecule. In detail, it is calculated by drawing a cone-like shape around the molecule and measuring the angle of the cone. This angle is defined as the angle between the axis of the molecule and the axis of its neighbours in a system. This angle provides information about the overall shape of the molecule and can be used to predict its reactivity and stability.

The CA index is used to predict the reactivity and selectivity of metal-catalysed reactions, as it can affect the orientation and accessibility of reactants near the metal centre. A smaller CA value indicates greater steric hindrance, and therefore, lower reactivity. The CA index is particularly useful in designing new catalysts for organic reactions, where the selectivity of the reaction is important. However, commonly, the CA is applied to assess ligands like phosphanes,¹⁰⁰ phosphites,¹⁰¹ and similar ligands. This is especially fitting due to the distinctive conical shapes of these ligands, even in cases of unsymmetrical phosphanes. The Tolman cone angle has been the favoured steric parameter for such ligands, despite some corrections and criticisms over the years,¹⁰¹ and being present book by Crabtree

in the key educational in organometallics.¹⁰² This parameter has proven exceptionally valuable to the chemical community, even when expanded the concept to steric maps.¹⁰³ Although the solid angle concept then appeared, and it was a viable alternative,^{104–107} it was not as intuitive and did not provide distinct advantages over the cone angle, and the bidimensional representations by Immirzi and Musco were not conclusive either,¹⁰⁸ as the cone angle radial profiles of Taverner and coworkers.¹⁰⁹ As a result, the solid angle did not gain significant traction in describing steric properties of phosphanes and related ligands.¹¹⁰

General steric descriptors, link the successful CA for phosphines with the % V_{Bur} of Nolan, Cavallo and coworkers. Actually, those indices are more general than the CA. In detail are proposed by Taft,¹¹¹ Charton^{112,113} and Verloop.^{114,115}

The % V_{Bur} index, developed quantitatively by Cavallo and coworkers, is a steric index that quantifies the volume of a molecule in a specific region of space. This index has found numerous applications in the study of the structure and reactivity of molecules and materials. One particularly interesting application is the correlation of % V_{Bur} with binding energy, which has important implications for the design of new catalysts and materials. However, the most important feature of % V_{Bur} index is that, while the Tolman CA had been proven highly effective for assessing phosphanes, it was clear that it was not the most suitable choice for characterizing the steric attributes of the emerging category of NHC ligands,^{21,116} which possess nearly C_2 -symmetric structures. NHC ligands have gained significant traction as highly efficient alternatives to traditional phosphanes in the last decades. Their application often results in smoother reaction conditions and novel reactivity. This is particularly evident in areas such as iridium-catalysed hydrogenation,^{117–119} palladium-catalysed C–C coupling reactions,^{120–122} ruthenium-catalysed olefin metathesis,^{123–129} and even unforeseen reactivity, exemplified by gold-catalysed reactions.^{130,131} Moreover, the saturated NHC skeletons play a pivotal role in introducing asymmetry to the NHC ring,^{124,132,133} enabling the use of chiral NHCs in asymmetric synthesis.^{134–136} And then for NHC ligands the use of % V_{Bur} could be extended to any type of ligand, unlike the limited CA.

Other steric indices. there are several other steric indices used in chemistry, including the steric strain energy (SSE), the steric energy (SE), and the steric shielding constant (SSC). These indices are used to measure the energy of steric interactions and the degree of shielding of atoms or functional groups from chemical reactions. Then, to switch to multidimensional steric parameters, Sterimol steric parameters developed by Verloop are a set of measures used to quantify the size and shape of ligands on metal centres within coordination complexes.^{137,138} These parameters involve evaluating the dimensions of specific substituents along defined directions in three-dimensional space,^{139,140} providing insights into the steric interactions within a catalyst's catalytic pocket, enabling the comparison of different ligands and their effects on catalytic behaviour.¹⁴¹ Knowing that understanding steric effects in asymmetric catalysts remains largely empirical, in 2012,



interactions between a drug molecule and its target protein. By analysing the steric maps of both the drug and the target protein, researchers can identify areas of the molecule that may cause unwanted interactions, such as clashes with other parts of the protein or other drugs. This information can be used to modify the drug molecule to minimize these interactions and increase its efficacy.

Another important application of steric maps is in the study of protein–protein interactions, where they can be used to visualize the complex interactions between different proteins in a cellular environment. By analysing the steric maps of multiple proteins, researchers can identify potential interactions between proteins and design experiments to test these interactions.

Steric maps have also been used in the study of enzyme–substrate interactions, where they can be used to predict the interactions between an enzyme and its substrate. By analysing the steric maps of both the enzyme and the substrate, researchers can identify potential interactions between the two molecules and design experiments to test these interactions.

5.5.3. How to understand a steric map. To construct the steric map, the entire complex was positioned within a Cartesian framework. This involved placing a designated point at the origin, a second point along the z -axis, and a third point in the xz plane. These points corresponded to either an atom or the midpoint of a specific group of atoms. For instance, in Fig. 1d, the complex was aligned so that the metal centre occupied the origin, the midpoint between the two P atoms extended in the negative z -axis direction, and one of the P atoms lay in the xz -plane. Following the alignment of the complex, any atoms not relevant to the steric map calculations were removed. Subsequently, an analysis was performed on the first coordination sphere surrounding the metal. In the case of Fig. 1d, the PdMe₂ fragment was eliminated from the complex, and the analysis focused on the MeDuPhos ligand.¹⁴

In more level of detail, behind the steric maps, upon orienting the complex and excluding specific atoms, a sphere with a radius of R , centred at the origin, is divided into sections using a regular three-dimensional cubic mesh of spacing s .¹² This mesh defines cubic voxels (v). To ascertain whether any ligand atoms fall within the van der Waals distance of a given voxel's centre, the distance between the centre of each voxel and all the ligand atoms is calculated. If no atom is within the van der Waals distance, the voxel is marked as “free”; otherwise, it is marked as “buried”. Once all voxels within the sphere are designated as free or buried, the program scans the sphere's top (from positive z -values) to identify the z -value of the first buried voxel. This process yields a surface denoted as $S(x,y) = z_B$. This surface outlines the ligand's exposed area facing the incoming reactants, essentially defining the shape of the reactive pocket. Positive z_B values indicate that the ligands extend into the $z > 0$ hemisphere, the region where the reacting groups are situated. A visual depiction of the interaction surface between the catalyst and the substrate can be found in Fig. 1d. Lastly, the maps are straightforward two-dimensional isocontour depictions of the interaction surface $S(x,y) = z_B$, illustrated in Fig. 1d. By default, a

sphere with a radius (R) of 3.5 Å surrounding the metal centre is utilized, and a mesh spacing of 0.1 Å is employed to survey the sphere for buried voxels. The van der Waals radii for atoms are sourced from the Cambridge Structural Database (CSD), with references provided in ref. 46. While hydrogen atoms weren't factored into the steric maps detailed in this study, the Web server database included the hydrogen radius from the Bondi set for alignment with CSD conventions.¹⁴⁶ Furthermore, radii not covered by these sources are standardized to 2.00 Å, followed by a scaling factor of 1.17, adhering to the CSD procedure. Currently, the steric maps presented herein are produced through the SambVca2.1 Web application. Since the graphical interface accommodates a maximum of 1000 atoms, the input files for generating the catalytic pocket of systems overcoming this number must be edited to exclude atoms. Take for instance, for the Amycalotopsis orientalis and the S221M/V223F/Y359A mutant of Streptomyces coelicolor, sourced from Protein Data Bank entries 2R5V and 3ZGJ respectively, were edited to exclude atoms situated more than 15 Å from the metal centre.¹² Following this adjustment, the steric maps were generated uploading the modified files.

Determining the appropriate size for the first coordination sphere prompts the question: “What Is the Optimal First Coordination Sphere Size?” In the analyses discussed earlier, a sphere encompassing the metal centre with a radius of 3.5 Å was consistently applied. However, this is particularly valid for transition metal catalysts. The chosen 3.5 Å radius holds significance as it was found optimal for correlating the DFT dimerization energy of 33 N-heterocyclic carbenes in the Wanzlick equilibrium.²² This value also demonstrated robustness in explaining a substantial amount of experimental data. Considering that typical bond distances between metals and ligands fall within the range of 2.0 to 2.5 Å, the utilization of a 3.5 Å radius to define the metal's first coordination sphere is expected to encompass the approximate van der Waals volume occupied by coordinated metal atoms. While this radius is the default choice, it is acknowledged that different studies might benefit from different values. For instance, surfaces, biological applications, bimetallic systems, where reactivity occurs between substrates connected to two distinct metal atoms, require a different protocol, and enlarge the size.

A similar outcome is observed when assessing the catalytic pocket of the metal organic framework (MOF) at radii of 3.5 Å and 6.0 Å, as shown in Fig. 4.¹⁴⁸ Despite these variations, this methodology has consistently succeeded in characterizing the overall asymmetry and contour of the catalytic pocket, regardless of the radius. However, it is important to note that for situations where bulkiness is not influenced by groups directly bound to the metal, a larger radius becomes essential to comprehensively capture the catalytic pocket's shape.

The steric maps of Cavallo and coworkers introduced SambVca 2,¹² a freely accessible web application designed to generate topographic steric maps that describe the catalytic pocket encircling a metal centre in various types of transition metal complexes. SambVca 2 features a user-friendly graphical interface, and it not only computes the buried volume steric



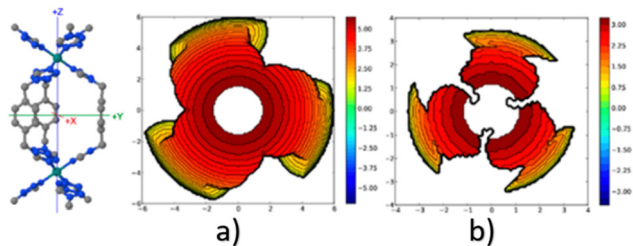


Fig. 4 Steric maps (plane xy) of the cavity of a MOF of iron, with a radius (a) 3.5 Å and (b) 6.0 Å. The centre of the MOF is at the origin, and the iron atoms are on the z axis. The isocontour curves of the steric maps are given in Å. Reproduced with permission.¹⁴⁸ Copyright © 2018, American Chemical Society.

descriptor but also serves as a full replacement for its predecessor, SambVca 1. These descriptors, the buried volume, and the topographic steric maps are designed to be adaptable across different systems, enabling the compilation of a comprehensive atlas spanning all ligand classes. They are also user friendly because the web server accepts different types of files, with format of X-ray,^{214,215} or any computed structure,^{33,216} even a designed structure with a building molecule package would lead to reasonable results.

In 2011, adhering to a similar approach, Houk, Montgomery, and Liu devised a method to visualize hindrance around a catalytic centre.²¹⁷ They utilized a graphical representation based on two-dimensional ligand steric contour maps, where contours indicate the distance between the ligand's van der Waals surface and the substrate. This approach was first employed to elucidate the reversal of regioselectivity in reductive couplings of alkynes and aldehydes catalysed by Ni-NHC complexes. More recently, Liu and Montgomery applied the same method to rationalize simultaneously the regio- and enantiocontrol in the same reaction when employing a novel class of highly hindered, enantiopure tailored NHC ligands.²¹⁸

Overall, steric maps are a powerful tool for predicting the steric effects on the reactivity and selectivity of chemical reactions, and they have found wide applications in various areas of chemistry, including organic synthesis, catalysis, and materials science. The method provides a quantitative measure of the steric hindrance in molecules and materials, allowing for a better understanding of their steric and electronic structure and reactivity. The applications of steric maps discussed in this article demonstrate its utility in the design of new catalysts and materials with specific steric and electronic properties, and its potential for advancing the fields of chemistry and materials science.

As computational power and theoretical methods continue to advance, the applications of steric maps are likely to expand even further, leading to new discoveries and advances in various fields. For example, in ionic liquids the steric maps have been able to understand the carbon dioxide,²¹⁹ fixation being the ionic liquids, not simple spectators, but the main protagonist. Moreover, the method may be applied to study the steric and electronic properties of new classes of materials, such as MOFs and porous organic polymers (POPs),¹⁴⁸ which

have attracted significant attention due to their potential applications in gas storage, separations, and catalysis.

5.6. Beyond the limits of the steric maps

Steric maps may be combined with other theoretical methods, such as density functional theory and molecular dynamics simulations, to gain a more comprehensive understanding of the properties and reactivity of molecules and materials. This will allow for the development of more accurate and predictive models for the design of new catalysts, materials, and devices.

To sum up, the applications of steric maps developed by Cavallo and coworkers have provided significant insights into the steric and electronic structure of molecules and materials,¹² allowing for the design of new catalysts and materials with specific properties. As the method continues to advance, it has the potential to revolutionize the fields of chemistry and materials science, leading to new discoveries and advances in various areas. Despite the clear qualitative character of steric maps,^{220,221} still currently to quantify, the $\%V_{\text{Bur}}$ index must be used to quantify,^{174,222–224} or in combination,^{31,219,225,226} preferentially because it covers both quantitative and qualitative aspects.

To demonstrate the versatility of steric maps of Cavallo and coworkers as a flexible instrument capable of offering a distinctive representation of nearly any ligand employed in homogeneous catalysis, they were utilized with a range of common ligands frequently found in organometallic chemistry. Embracing the principles akin to Tolman and Sterimol methodologies, the fundamental concept guiding the creation of topographic steric maps lies in their inherent user-friendliness, assuming a crystallographic or computationally determined structure is accessible.

In addition to providing a visual representation of the catalytic pocket, the steric maps of Cavallo and co-workers also serve as distinctive markers for transition metal complexes. This can be illustrated with a series of important chiral ligands in catalysis,²²⁷ since in the complexes where they participate they are recognized to induce enantioselectivity in several well-established reactions, such as Diels–Alder cycloadditions, Michael and Mukaiyama–Michael additions, Heck and asymmetric hydrogenations. Among them, bisoxazoline and binaphthyl ligands within C_2 symmetric complexes and a phosphinoxazoline ligand within C_1 symmetric complex. In detail, a straightforward visual examination of the ball-and-stick or CPK structures of those three complexes does not readily facilitate a comparison of these different systems. Conversely, the steric maps not only distinguished between the individual complexes but also provided an intuitive indication of how various ligands shape the catalytic pocket differently.¹² For instance, the slender bisoxazoline ligand occupies space around the equator of the catalytic pocket, with upward-facing *t*-butyl groups creating relatively gentle protrusions in the south-western and north-eastern quadrants. In contrast, the complex with a binaphthyl ligand forms a groove from the north-west to the south-east, akin to the former complex. The notable differences lie in the naphthyl groups of the binaphthyl ligand, which cover the bottom of the catalytic pocket in the north-western and



south-eastern quadrants, along with the enhanced hindrance caused by the phenyl groups in south-western and north-eastern quadrants. The catalytic pocket of the complex with the monodentate phosphinooxazoline ligand can be seen as a combination of both the two previous complexes, resembling the western hemisphere with the bisoxazoline and the eastern hemisphere of the binaphthyl. Consequently, the % V_{Bur} values indicate that the complex containing the binaphthyl has a significantly larger % V_{Bur} than with bisoxazoline, while the monodentate falls in between in terms of % V_{Bur} .

And the final goal of steric maps actually is the first aim that was thought for them. In detail, steric maps should also serve as valuable tools for catalyst design by facilitating the evaluation of the impact of structural modifications. In this sense, Bertrand and collaborators synthesized cyclic (alkyl)(amino)-carbenes containing a six-membered backbone (CAAC-6).²²⁸ In comparison to their counterparts with a five-membered structure (CAAC-5), these carbenes exhibit increased % V_{Bur} and enhanced donor and acceptor properties, as demonstrated by the observed $n \rightarrow \pi^*$ transition extending into the visible spectrum. Their pronounced ambiphilic nature enables the carbene to insert intramolecularly into an unactivated C(sp³)-H bond. As ligands, they demonstrate superior performance over the five-membered analogues in the palladium-mediated α -arylation of ketones using aryl chlorides. However, the use of % V_{Bur} and steric maps was preceded by former experimental work, involving CAAC-5 Pd-complexes revealed that featuring unhindered ethyl substituents, lacked the ability to catalyse the α -arylation reaction. In contrast, complexes incorporating a bulky substituent in place of ethyl groups, displayed high reactivity with the less sterically demanding chlorobenzene. However, the latter catalysts exhibited reduced reactivity when faced with bulkier aryl chlorides like *o*-chlorotoluene or 2-chloro-*m*-xylene.²²⁹ Thus, a combination of experiments and topographic maps enhanced the race for a new generation of efficient catalysts. And Szostak and coworkers increased even more the sterical hindrance of CAAC ligands, reaching the CAAC-IPr* ligand,²³⁰ catalytically proved for Cu-catalysed hydroboration of alkynes (Fig. 5). In detail, with the % V_{Bur} of 49.5% and 49.4% for [Cu(IPr*-CAAC^{Me})Cl] and [Cu(IPr*^{MeO}-CAAC^{Me})Cl], respectively, represented the bulkiest CAACMe ligands, for example, comparing the first with the homologous with the IPr ligand, the % V_{Bur} drops to 41.9%. Anyway, the values for the CAAC-IPr* ligands are still lower than the value of 52.1% for the catalyst [Cu(IPr*)Cl].

We can relate % V_{Bur} and steric maps to computational chemistry, but from the beginning it was created with the desire to go beyond the boundaries of computational chemistry, and reach all of chemistry, but this was not enough either, and it was reached physics and biology.

At the executive level, how would this be done? therefore facilitating its use and with any structure optimized with computational calculations these tools would be valid, but also with X-ray diffraction, and at an informative level it was intended not to have to make use of hydrogens because in this experimental technique they cannot be detected It must be said

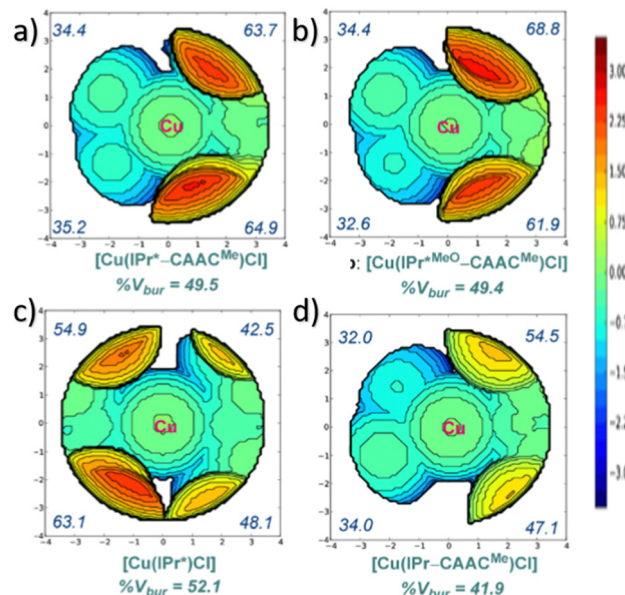


Fig. 5 Steric maps of (a) [Cu(IPr*-CAAC^{Me})Cl], (b) [Cu(IPr*^{MeO}-CAAC^{Me})Cl], (c) [Cu(IPr*)Cl], and (d) [Cu(IPr-CAAC^{Me})Cl] showing % V_{Bur} per quadrant. Adapted with permission.²³⁰ Copyright © 2022, Royal Chemical Society.

that the agreement with or not the inclusion of hydrogen atoms did not reduce precision in the correlations where they could be involved, sending the message that their weight is overestimated, and in fact, heteroatoms are the agents that mark the steric hindrance.

But if that was not enough, *i.e.* ignoring the hydrogen atoms, these tools would be almost as valid simply with structures created by hand using pre-parameterized bonds, from any molecule building software, protein building, or of materials.

6. Fields of practical utility of % V_{Bur} and steric maps

First of all, it must be said that among the applications only a minimal summary of these can be reached, since today there are already more than 2000 works that have made use of the % V_{Bur} concepts or steric maps, and that have cited the pioneering works by Cavallo and collaborators. And those studies that use them in combination with other indices or steric tools are particularly interesting, and validate them, or especially refute them, not only to know what the positive aspects are, but above all to visualize what their limitations are.

The concepts of steric maps and % V_{Bur} have uses in all the range of computational chemistry. Thus, knowing about computational chemistry is an interdisciplinary field that combines chemistry, physics, mathematics and computer science to develop and apply computational methods to study chemical systems, to distinguish what are the hot topics in computational chemistry is intriguing, and in all there is potential participation, thus let's summarize them in the following list:



6.1. Sterical hindrance of ligands

The main use of $%V_{\text{Bur}}$ and steric maps is simple quantification of the steric hindrance that a particular ligand causes at the metal centre, and especially if there is asymmetry the steric maps and $%V_{\text{Bur}}$ by quadrants gain weight. In summary, it is about describing how the size of a ligand affects, that is, how it is arranged in space. Most of the papers are related to NHC ligands, and in particular ligated to ruthenium,^{206,231} copper,^{232,233} gold,²³⁴ and palladium,^{178,235} but actually, the most important actor is the ligand itself,^{24,236} and the metal simply can tune a bit its flexibility. There is a tendency to think that the rule that the more sterically a ligand is, the worse is the resulting catalysis. But we could give many counterexamples that this is not the case. There are different reasons, either because it blocks a coordination of an additional ligand, because it leaves the perfect cavity in the first coordination sphere around the metal to interact with the substrates, and actually, it is common then that the most sterically protected metal centre is the most effective catalytically.^{237,238}

Take for instance, the IPr* (1,3-bis(2,6-diisopropylphenyl)imidazol-2-ylidene) ligand represents a sterically hindered ligand compared to the common IPr (1,3-bis(2,6-diisopropylphenyl)imidazol-2-ylidene), however Nolan, Szostak and coworkers unveiled that in most of the cases the catalytic performance is better for the first NHC ligand in cross coupling reactions,^{239,240} despite the origin of IPr* in 2012 by Nolan was devoted basically only to selectivity, and steric maps unveiled that the increase of $%V_{\text{Bur}}$ was not fundamental,²⁴¹ but also the asymmetry of each quadrant unveiled by the steric maps (see Fig. 6).²⁴² Through the utilization of density functional theory

some scenarios were delineated in which the NHC ligand assumes an active role and instances in which it merely acts as an observer. Alternatively, its modification does not yield substantial alterations to its catalytic function or performance. Actually, the differentiation of the maps, as well as the $%V_{\text{Bur}}$ is much greater for the free ligands than when they are in the complex. In fact, in the same study, maps are also provided without the ligand *trans* to NHC, which is dissociated by olefin metathesis catalysis. But the conclusion of this and other studies is that it is better to do the steric study of a ligand with the conformation of the complex,^{14,242} since the flexibility exists.

The next IPr# by Szostak and coworkers even increased more the steric hindrance,²⁴³ up to the limit in terms of NHC ligands. And in 2016, Michalak, Poater and coworkers in 2016 an efficient and readily scalable process involving NHC-copper(i) halide catalysis for the addition of terminal alkynes to 1,1,1-trifluoromethyl ketones,²³² in an aqueous environment for the first time. Utilizing DFT calculations, the relationship between the catalytic alkynylation yield and the sterics of N-heterocyclic carbenes (NHCs), represented by buried volume ($%V_{\text{Bur}}$), is elucidated. The findings underscore the predominant influence of steric effects on the reaction yield, the kinetic cost being lower the greater the steric hindrance of the NHC ligand. The reason is not a reduction in the relative energy of the rds but of the rate determining intermediate (rdi).⁹⁷

But after these last 2 decades with supremacy of NHC ligands in organometallic catalysis, particularly due to the strong donation of σ in combination with an adjustable steric environment, it has not meant stagnation, since recently Szostak and coworkers for example have generated thiazol-2-ylidenes,²⁴⁴ which are competitive to NHCs. The thiazole heterocycle and enhanced π -electrophilicity are shown to result in a class of highly active carbene ligands for electrophilic cyclization reactions to form valuable oxazoline heterocycles, and this transformative process has been partly mapped sterics, which have helped to locate the structural differences between both types of ligand.

Moreover, switching to pincer based catalysis, and in particular, to the Overman rearrangement reaction with a gold catalyst, achieving precise spatial control in gold catalysis presents inherent challenges due to the linear coordination nature of Au(i) and the generally planar ligands adopted by Au(III) complexes. Klein and coworkers introduced a novel $(\text{NNN})_{\text{diiPr}}\text{Au-OH}$ complex characterized by sterically hindered properties, suggesting the potential of the (NNN) ligand framework to interact steric interactions with substrates through its strategically positioned aryl groups.²⁴⁵ To explore these steric aspects in greater depth, variations in the *ortho*-substituent of the aryl group in $(\text{NNN})_{\text{x}}\text{Au-Cl}$ complexes allowed to manipulate the buried volume surrounding the Cl atom, and this variation was subsequently correlated with yields achieved in an Au-catalysed Overman rearrangement. Thus, the (NNN) pincer ligand was announced as an attractive platform for efficiently designing Au(III) complexes with the potential for stereoselective catalysis. On the other hand, the NNN pincer

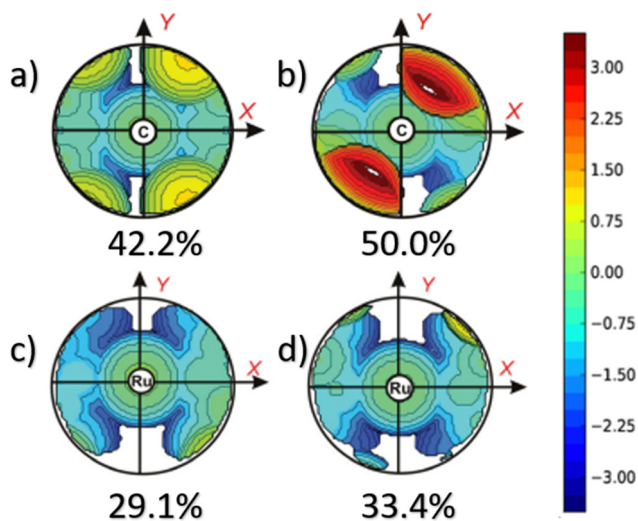


Fig. 6 Steric maps and $%V_{\text{Bur}}$ of the free (a) IPr and (b) IPr* NHC ligands, and in the ruthenium dichloride indenylidene complexes for (c) IPr and (d) IPr* NHC ligands. The isocontour curves of the steric maps are in Å. The *xz* plane is the mean plane of the NHC ring, whereas the *yz* plane is the plane orthogonal to the mean plane of the NHC ring, and passing through the carbene C atom of the NHC ring. The carbene C atom of the NHC ring is at the origin. Reproduced with permission.²⁴² Copyright © 2013, Royal Chemical Society.



ligand was also ligated to ruthenium by Milstein and coworkers to carry out the reforming of aqueous ethylene glycol to glycolic acid and molecular hydrogen as a green subproduct.²⁴⁶ Indeed, the planarity of pincer ligands makes them excellent candidate ligands for using steric maps and $\%V_{\text{Bur}}$,²⁴⁷ to elucidate how hindered the reactivity of their metal centre will be.

Cavallo, Rueping and coworkers in 2020 introduced a regio-divergent protocol for the cleavage of C–O bonds using magnesium catalysis.²⁴⁸ Employing readily accessible magnesium catalysts, this method achieves the targeted hydroboration of diverse epoxides and oxetanes (see Fig. 7a), leading to the formation of secondary and tertiary alcohols with exceptional yields and regioselectivities. The selectivity towards 2,2-disubstituted epoxides was achieved for 13 examples, whereas the other 42 led to excellent enantiospecificity. Detailed mechanistic investigations along with DFT calculations shed light on the unanticipated regiodivergence observed in the reaction. Essentially, the steric maps of the coordinated epoxide intermediate, *i.e.* the rdi depicted in Fig. 7b, reveal a predictable observation: the western quadrants containing the unaltered carbon atom of the epoxide exhibit lower hindrance compared to the eastern quadrants housing the substituted carbon atom (Fig. 7c).

6.2. Selectivity in catalysis

Steric maps serve to highlight the catalyst's exposed surface facing the substrate, which can then undergo a more comprehensive analysis to uncover the physicochemical attributes of the catalytic pocket.^{249–251} For instance, scrutinizing the catalytic pocket solely through its steric attributes falls short in elucidating the enantioselectivity in the asymmetric 1,4-addition of phenylboronic acid to 2-cyclohexenone, resulting in chiral 3-phenylcyclohexanone, driven by Rh-catalysts led by Dorta and coworkers.¹⁸⁷ With two catalysts exhibiting remarkable enantioselectivity, with enantiomeric excesses of 90% and 99%, the great performance of the first was anticipated, however the even higher enantioselectivity of the second was unexpected, and posed a conundrum, considering that the enantioselective induction is expected to be influenced by the small, upward-oriented S=O moieties. Investigating the catalytic pocket through its steric characteristics confirmed the

marked asymmetry of the catalytic pocket of the first, exhibiting a groove aligned from northwest to southeast shaped by the upward-oriented *p*-tolyl groups. Computational analysis validated that the preferred transition state accommodates the reacting groups within this groove. Conversely, the steric map of the second portrays a notably flatter catalytic pocket, implying that the heightened enantioselectivity exhibited is not solely driven by steric effects. Here, expanding the assessment to the surfaces of the catalytic pockets, based on the steric maps using electrostatic potentials derived from DFT calculations, yields an alternative perspective on the two catalysts. Remarkably, the electrostatic potential map of the catalytic pocket of the best displays marked asymmetry, featuring areas of highly negative electrostatic potential near the S=O moieties. Conversely, the electrostatic potential map of the less good is more uniform. Consequently, the electrostatic map of both catalysts shares a resemblance in shape.¹⁸⁷ The favoured transition state once again positions the reacting groups within the northwest to southeast groove, which, in this instance, is influenced by the electrostatic potential on the catalytic pocket's surface. In summary, employing steric maps to visualize catalytic pockets and characterizing their surfaces with a broad property such as electrostatic potential enables the elucidation of the origins of the stereoselectivities exhibited by this pair of catalysts.

Aiming to predict and understand how different structures influence reactivity, Doyle and coworkers introduced a versatile and quantitative method for classifying reactivity differences,²⁵² known as 'reactivity cliffs,' in 11 sets of chemical reactions involving Ni and Pd catalysts. These reactions utilize monodentate phosphine ligands. The study revealed a distinct molecular descriptor, termed 'minimum percent buried volume [$\%V_{\text{Bur}}(\text{min})$]', which effectively separates the datasets into regions of high and low reactivity using a consistent threshold in cross-coupling catalysis. Experimental investigations on organometallic systems show that this threshold aligns with whether the metal is doubly or singly bound to ligands, and that $\%V_{\text{Bur}}(\text{min})$ provides a valuable predictive insight into ligand behaviour in catalytic processes. Thus, $\%V_{\text{Bur}}$ can put the threshold in any catalytic reaction to predict if it will be or not feasible, and it is also a tool to discriminate between families of substrates to interact with the metal centre, either by the total number or by quadrants.²⁵³

Stereoselectivity plays an important role in some chemical reactions, particularly in polymerizations involving a semi-chiral monomer.²⁵⁴ Under these circumstances, the steric index around the active catalytic centre, where the living polymer chain is growing, dictates the stereoregularity of the final polymer.²⁵⁵ It helps to design new efficient catalysts that subsequently lead to major advances in the development of either homogeneous or heterogeneous catalysts for semi-chiral derivative polymers, such as polyolefins.

In 2023, Osuna, Bietti, Costas and coworkers used the steric maps to enhance the enantioselective hydroxylation at unactivated tertiary C–H Bonds.²⁵⁶ Here, the challenge of achieving those enantioselective C–H oxidations has been endeavoured. In detail, the hydroxylation of tertiary C–H bonds within cyclohexane structures using H_2O_2 as the oxidant combined

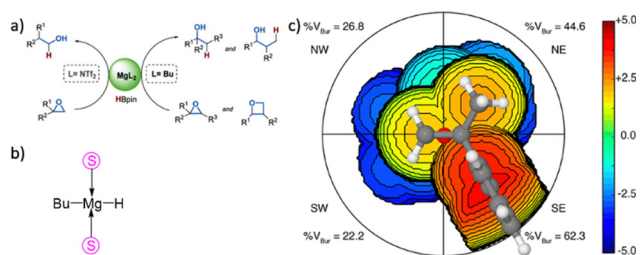


Fig. 7 (a) Regiodivergent ring opening of epoxides. (b) Rate determining intermediate and (c) its steric map including the $\%V_{\text{Bur}}$ of the quadrants (The origin is set at the midpoint of the C–C bond within the epoxide, and the O atom of the epoxide is positioned along the Z-axis. The scale of the steric contours is included in Å). Adapted with permission.²⁴⁸ Copyright © 2020, American Chemical Society.



with a manganese catalyst, exhibited a structural congruence with the substrate, much like the lock-and-key recognition mechanism observed in enzymatic active sites. Theoretical calculations in this project by molecular dynamics simulations is the right tool to understand the behaviour of those catalysts in solution and at interfaces. They revealed that the enantioselectivity observed is a consequence of the precise fitting of the substrate's scaffold into the catalytic site. This fitting was accomplished through a network of weak non-covalent interactions that are complementary in nature. This enantioselective process leads to stereoretentive C(sp³)-H hydroxylation, enabling the generation of multiple stereogenic centres within the molecule in a single step—up to four in number. These centres can be individually controlled and manipulated using conventional methods, offering a streamlined approach to access a range of chiral frameworks from a single precursor. The importance of the binding pockets is thus described by the steric maps leading to simple images that help to understand how the substrate interact, and even can be expanded to chirality.²⁵⁷

6.3. Reactivity in catalysis

Since copper-catalysed atom transfer radical polymerization (Cu-ATRP) is a widely utilized technique for controlled radical polymerization,²⁵⁸ extensive research into its mechanisms had been performed. And in 2019, the crucial transition states governing the activation and deactivation of the polymer chain during Cu-ATRP were computationally explored by Coote, Matyjaszewski, Peng and coworkers.²⁵⁹ They covered the gap to understand how ligands and initiators influence these processes. This study introduced the first computational investigation into the activation transition states of Cu-ATRP, shedding light on factors that impact activation and deactivation rates. Unusually, the bromine atom transfer between the polymer chain and the Cu catalyst occurs through a bent geometry, involving significant interactions between the polymer chain end and the catalyst's ancillary ligand. This interaction complexity dictates activation and deactivation rates, controlled by both the Cu catalyst's electronic properties and steric repulsions between the ligand and initiator. Furthermore, the flexibility of the ligand backbone was found to be pivotal in activation. Through theoretical analyses, three important descriptors were identified: catalyst's HOMO energy (E_{HOMO}), percent buried volume (% V_{Bur}), and catalyst distortion energy (ΔE_{dist}), capturing electronic, steric, and flexibility influences on reactivity, respectively. These descriptors were then correlated with experimental activation rate constants using multivariate linear regression, yielding a reliable predictive model for ligand effects on reactivity. Validation using various ligands demonstrates an average error of under ± 2 kcal mol⁻¹ compared to experimentally derived activation energies. The same methodology was extended to predict the reactivity of different alkyl halide initiators, using descriptors of bond dissociation energy (BDE) and Cu-X halogenophilicity. Thus, the digital rendition of the steric map, represented as an array of points defining the surface in Cartesian space, holds potential as a

digital steric descriptor for applications like multilinear regression analysis.^{260–263} And thus, it could also be incorporated into workflows designed for high-throughput screening of novel catalysts using ML methodologies.²⁶⁴ Even though in most of cases the % V_{Bur} is a perfect structural tool/parameter to correlate with a series of data, experimental or computational,^{265–267} it is also necessary to be honest in recognizing that it does not seem that the % V_{Bur} is one of the variables with the best behaviour in terms of the reproduction of the structural and/or steric part,^{61,268,269} with preference for the Tolman electronic parameter (TEP),^{270,271} and sometimes the bond distances²⁶⁸ and angles can be a better choice.^{61,269} And particularly, in the activation of the C–O bond of the allylic alcohol, catalysed by palladium, the multiple linear regression analysis does not use as a variable either the % V_{Bur} or the C–O bond distance but the angle with the metal.²⁶⁹ Actually, most of the studies base the use of steric maps, independently of the catalytic reaction under study, and based on the cleft on the metal centre, where the substrate should interact, as Gramage-Doria and coworkers have developed with non-biological catalysts that mirror enzyme principles.²⁷² These catalysts typically possess adaptable molecular recognition sites for substrate binding, often following conformational selection pathways. The catalyst comprised an inflexible substrate-recognition site stemming from a zinc-porphyrin in the second coordination sphere. Thus, the catalytic pocket is the aim of steric maps,^{273,274} and the lability of the ligands is also fundamental, because if it is easy to dissociate a ligand, take for instance a chloride, the sterical hindrance on the metal centre drops automatically.²⁷⁵ In 2020, Occhipinti, Jensen and coworkers unveiled the Z-selectivity in olefin metathesis of monothiolate ruthenium indenylidene catalysts.²⁷⁶ The study is very simple, it was the replacement of a pyridine ligand by a thiolate ligand (see Fig. 8a). To further investigate the steric influences of the thiolate, a comparison was made between the % V_{Bur} of 95.0% for its active species, after removing a pyridine ligand (see Fig. 8b), and its corresponding dichloride (without pyridine, as well). The results indicate a substantial contribution of the thiolate to the overall steric congestion, with a % V_{Bur} value of 86.4% for the dichloride catalyst. Interestingly, a minor portion of this difference arises from the sulphur atom itself, signifying the effect of replacing chlorine with sulphur. Upon solely considering ruthenium and the atoms directly bonded to it, the % V_{Bur} values were slightly larger again for the thiolate (67.7%) in comparison to the dichloride (67.0%).

6.4. Macromolecular chemistry

One instance of catalyst design through the comparative analysis of catalytic pockets within related complexes can be observed in the advancement of chelating ligands tailored for the Pd-catalysed copolymerization of ethylene with olefins possessing a polar functional group. Presently available catalysts encounter a limitation in selectively producing high-molecular-weight linear polymers, a drawback that prompts the need for improvement. Additionally, the conventional approach of augmenting molecular weight in olefin polymerization—by



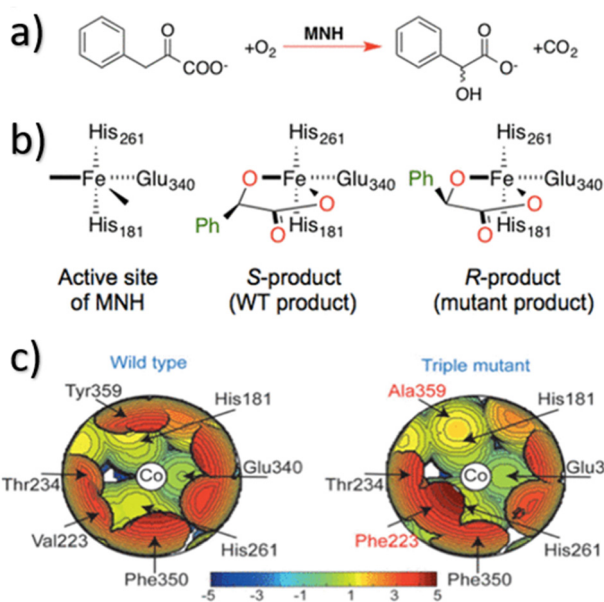


Fig. 10 (a) Scope of the reaction. (b) Catalytic active site. (c) Steric map of the catalytic active site from the crystal of the wild-type *A. orientalis* (left) and of S221M/V223F/Y356A mutant of *S. coelicolor*. Reproduced with permission.¹² Copyright © 2016, American Chemical Society.

the steric maps of the catalytic pockets reveals that mutations such as V223F and Y359A have distinct effects: V223F impedes the open space where (*S*)-mandelate's aromatic ring would be situated, while Y359A opens up space in the north-western quadrant, facilitating accommodation of (*R*)-mandelate's aromatic ring. Thus, by employing steric maps, the study gains insight into the structural basis of enantioselectivity alterations in metalloenzymes, underscoring the broader applicability of these maps in elucidating catalytic mechanisms in diverse contexts.

In 2020, in one of the pioneering research works about pandemic COVID, Poater showed the effectiveness of inhibitors in blocking the COVID protease (SARS-CoV-2 M^{Pro}), and α -ketoamide derivative substrates experimentally developed by Hilgenfeld and coworkers.²⁹⁹ Following biological characterization, DFT calculations were employed to not only elucidate the thermodynamically favourable interaction mechanism of the active inhibitor with respect to the non-active ones, but also to shed light on the kinetic aspects of this interaction.³⁰⁰ Aromaticity, a concept that has been debated and unverifiable,³⁰¹ plays a pivotal role in explaining the role of histidine (His41 of M^{Pro}, see Fig. 11a). His41 establishes a hydrogen bond with the hydroxyl group and facilitates the proton transfer from the thiol of Cys145 at negligible energy expense, thus favouring the interaction with the inhibitor acting as a Michael acceptor. The steric hindrance around His41 is minimal, as evidenced by the steric maps (Fig. 11b), confirming it as the region with the least hindrance within the first coordination sphere of the thiol group where reactivity occurs. These maps also provide insight into the significantly poorer activity of certain inhibitors. In these cases, the presence of a cyclohexyl residue (as opposed

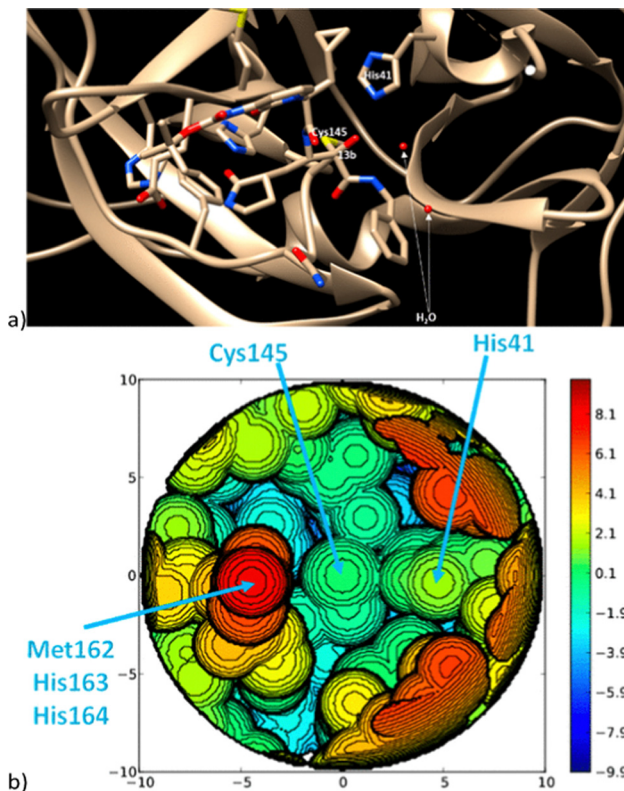


Fig. 11 (a) Region around the sulphur of Cys145 of M^{Pro} in the X-ray structure in space group C₂. (b) Steric map around the sulphur of Cys145 of M^{Pro} from the crystallographic structure (space group C₂) of the protease M^{Pro}. On the z axis, there is the sulphur atom of Cys145 and the carbon atom of the carbonyl of the substrate is at the origin, while its oxygen atom is on the xz plane (in Å). Reproduced with permission.³⁰⁰ Copyright © 2020, American Chemical Society.

to the cyclopropyl residue) complicates the necessary free rotation of His41. This rotation is crucial for sequential proton donation and acceptance. The expected steric clash between the pyridine ring and Gln189 is replaced by this hindered rotation. Thus, apart from the aromaticity to explain the reactivity, to unveil why the substrate with a cyclohexyl substituent does not work, steric maps were basic. Continuing with hydrogen bonds, steric maps allowed Posada-Pérez *et al.* also explain how the series of halides interacted with nickel but had a totally different stability due to hydrogen bonds with 3 amino groups that are located,³⁰² if possible in the same plane as the halide symmetrically. Thus, in this case, steric maps achieve an understanding not only of the first coordination sphere, but of the second, and doing so not in a biological system, but a simple inorganic complex. This is another reference of the versatility of this type of topographic steric maps.

In homology with the characterization of protein surfaces based on the attributes of exposed amino acids, a similar approach could be adopted for characterizing the catalytic pocket of catalysts.³⁰³ This involves assessing attributes like hydrophobic/hydrophilic nature, polar/apolar regions, aromatic patches, hydrogen bond donors and acceptors, as well as Lewis base and acid groups. Simplistically Vidal-López *et al.* with an



properties, catalyst catalytic pockets could be assessed for hydrophobic/hydrophilic, polar/apolar, and aromatic regions, hydrogen bond donors and acceptors, or Lewis base and acid groups.

Another avenue for development could entail integrating the steric map source code with computational methods aimed at real-time geometry optimization of molecular structures. This integration could pave the way for a web-based 3D computer-aided design system that enables interactive modification of initial catalyst skeletons. Actually, new webs or servers that collect several of those indices are ongoing. Take for instance, in 2023, GitHub was presented by dos Passos Gomes, Friedrich and Gensch,³¹³ gathering the calculation of the bite angle,^{314,315} buried volume,²² conformer tools, dispersion descriptor, exact ligand CA,²⁰ ligand solid angle,^{316–318} local force constant, pyramidalization, solvent accessible surface area, Sterimol parameters^{137,138} and Extended tight-binding (XTB) electronic descriptors.^{319,320}

Lastly, the digital steric map, represented by an array of Cartesian points defining the surface, might be utilized as a steric descriptor within multilinear regression analysis or embedded within a workflow for high-throughput catalyst screening through machine learning techniques.

8. Conclusions and future perspectives

Overall, predictive catalysis by DFT calculations is a powerful tool for the study of catalytic reactions and provides a comprehensive understanding of the reaction mechanism, energy profile, and reactivity of catalytic systems. These concepts of % V_{Bur} and steric maps are tools commonly used in the design and optimization of drugs, catalysts, and materials, as well as in the prediction of the properties and behaviour of molecules in different environments.

Knowing that steric indices are a type of parameter that describes the spatial arrangement of atoms in a molecule, they can be used to quantify the degree of steric hindrance or sterically-induced strain in a molecule, which can in turn affect its reactivity and stability.

In conclusion, the Steric index, Tolman cone angle, steric maps, and % V_{Bur} of Nolan and Cavallo are all important parameters used in the study of molecular interactions and in the design of chemical reactions. These tools provide valuable information about the molecular size and shape, spatial orientation, and stability of molecular interactions, and they are widely used in the fields of drug design, chemical engineering, and materials science.

In conclusion, the applications of steric maps in drug design, structure prediction, reactivity prediction, synthesis design, and molecular docking make them an important tool in the field of molecular science. With their ability to visualize the distribution of electron density in a molecule and to predict the interactions between atoms and molecules, they provide

valuable information for understanding and optimizing the properties of chemical compounds.

Overall, the relationship between steric index or % V_{Bur} and yield is complex and can depend on multiple factors. However, these parameters can be useful in predicting the reactivity and selectivity of chemical reactions and in designing more efficient catalysts.

The acceleration of catalyst engineering through computer-aided design is achievable by condensing molecular catalyst features into numerical descriptors. These descriptors establish connections between reactivity and structure. Within this framework, the introduction of topographic steric maps has been pivotal. Apart from being user friendly to obtain, these maps offer a visual representation of the catalytic pocket, the area in a catalyst where substrates bind and undergo reactions. This visualization allows for alterations by adjusting various parameters. When combined with DFT calculations, these steric maps facilitate swift exploration of catalyst structural modifications, enabling the online creation by a web server of new catalysts for the broader chemical community. Thus, this review focuses on the utilization of topographic steric maps to either rationalize the behaviour of existing catalysts, ranging from synthetic molecular entities to metalloenzymes, or to design enhanced catalysts.

Last but not least, future research in the field of % V_{Bur} and steric maps is likely to focus on the development of new computational tools and techniques to enhance its accuracy and predictive power. As the field of chemistry continues to evolve, those tools will undoubtedly play an important role in enabling researchers to understand and control the electronic properties and reactivity of molecules and materials, opening up new avenues for the design and synthesis of novel compounds and materials with important applications in various fields.

Author contributions

All the authors wrote and revised the manuscript.

Abbreviations

% V_{Bur}	Percentage of buried volume
AI	Artificial intelligence
CA	Cone angle
DFT	Density functional theory
ETS-NOCV	Extended Transition State-Natural Orbital for Chemical Valence
HOMO	Highest occupied molecular orbital
LUMO	Lowest unoccupied molecular orbital
ML	Machine learning
MOF	Metal organic framework
NHC	N-heterocyclic carbene
NHO	N-heterocyclic olefin
rdi	Rate determining intermediate
rds	Rate determining step
SHI	Steric hindrance index



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