# **Chemical Science**

# EDGE ARTICLE



Cite this: Chem. Sci., 2016, 7, 5032



www.rsc.org/chemicalscience

### 1. Introduction

In 1927–30, Thomas, Fermi, and Dirac proposed treating the energies of systems of many electrons by a statistical model of the electron density  $\rho$ .<sup>1-3</sup> However, Teller proved that no such model could predict the existence of stable molecules.<sup>4</sup> In 1951 Slater<sup>5</sup> proposed a set of self-consistent-field (SCF) equations involving effective local potentials defined in terms of  $\rho$  by the

free-electron gas model, and these equations do predict molecular binding. In 1964, Hohenberg and Kohn proved that the ground-state energy of a system can be written as a universal functional  $F$  of  $\rho$ ,<sup>6</sup> and in 1965 Kohn and Sham<sup>7</sup> represented the density as the one-electron density computed from the square of a Slater determinant and provided an SCF scheme for using the Hohenberg–Kohn theorem to calculate the energies of atoms, molecules, and solids. These equations resemble Slater's equations but the new derivation shows that SCF equations with local potentials are not just an approximation scheme but can be exact if Slater's approximation to exchange is replaced by a local potential derived from the so-called exact exchange– correlation functional (XCF), which is a functional of  $\rho$ . Unfortunately, this exact XCF is unknown and is essentially unknowable.<sup>8</sup> Progress in density functional theory is dominated by attempts to obtain better approximations to this functional; although somewhat unsystematic, this progress has revolutionized the practice of modern chemistry.

# MN15: A Kohn–Sham global-hybrid exchange– correlation density functional with broad accuracy for multi-reference and single-reference systems and noncovalent interactions†

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Kohn–Sham density functionals are widely used; however, no currently available exchange–correlation functional can predict all chemical properties with chemical accuracy. Here we report a new functional, called MN15, that has broader accuracy than any previously available one. The properties considered in the parameterization include bond energies, atomization energies, ionization potentials, electron affinities, proton affinities, reaction barrier heights, noncovalent interactions, hydrocarbon thermochemistry, isomerization energies, electronic excitation energies, absolute atomic energies, and molecular structures. When compared with 82 other density functionals that have been defined in the literature, MN15 gives the second smallest mean unsigned error (MUE) for 54 data on inherently multiconfigurational systems, the smallest MUE for 313 single-reference chemical data, and the smallest MUE on 87 noncovalent data, with MUEs for these three categories of 4.75, 1.85, and 0.25 kcal mol<sup>-1</sup>, respectively, as compared to the average MUEs of the other 82 functionals of 14.0, 4.63, and 1.98 kcal mol<sup>-1</sup>. The MUE for 17 absolute atomic energies is 7.4 kcal mol<sup>-1</sup> as compared to an average MUE of the other 82 functionals of 34.6 kcal mol<sup>-1</sup>. We further tested MN15 for 10 transition-metal coordination energies, the entire S66x8 database of noncovalent interactions, 21 transition-metal reaction barrier heights, 69 electronic excitation energies of organic molecules, 31 semiconductor band gaps, seven transition-metal dimer bond lengths, and 193 bond lengths of 47 organic molecules. The MN15 functional not only performs very well for our training set, which has 481 pieces of data, but also performs very well for our test set, which has 823 data that are not in our training set. The test set includes both ground-state properties and molecular excitation energies. For the latter MN15 achieves simultaneous accuracy for both valence and Rydberg electronic excitations when used with linear-response time-dependent density functional theory, with an MUE of less than 0.3 eV for both types of excitations. **EDGE ARTICLE**<br>
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<sup>†</sup> Electronic supplementary information (ESI) available: Mean unsigned errors of Database 2015B for 84 functionals and geometries of databases ABDE13, S6x6, SBG31, and EE69. See DOI: 10.1039/c6sc00705h

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We will discuss Kohn–Sham theory in the spin-unrestricted form in which the XCF is generalized to depend not just on total density but on the spin-up and spin-down densities  $\rho_{\alpha}$  and  $\rho_{\beta}$ (where  $\alpha$  and  $\beta$  denote spin-up and spin-down electrons).<sup>9,10</sup> The original approximations to the XCF involved separate approximations to exchange energy density and correlation energy density, at given points in space, that are explicit functionals of the local electron spin densities  $\rho_{\alpha}$  and  $\rho_{\beta}$ . These are called local-spindensity approximations (LSDAs).<sup>11-13</sup> Later work added dependence of exchange and correlation energy densities on the magnitudes  $s_{\alpha}$  and  $s_{\beta}$  of the reduced gradients of  $\rho_{\alpha}$  and  $\rho_{\beta}$ , leading to generalized gradient approximations (GGAs)<sup>14-16</sup> and on the local kinetic energy densities  $\tau_{\alpha}$  and  $\tau_{\beta}$  of spin-up and spin-down electrons,<sup>17</sup> leading to meta-GGAs. The partition of the electronic energy into exchange and correlation in exchange–correlation functionals is different from the partition usually used in wave function theory; and in Kohn–Sham theory one can also approximate exchange and correlation together, without separating them, leading to nonseparable gradient approximations (NGAs)<sup>18</sup> and meta-NGAs.<sup>19</sup> Note that the local kinetic energies are local functionals of the occupied Kohn–Sham orbitals.<sup>20</sup> It can easily be shown<sup>21</sup> though that the unknown exact functional is not a local functional of the orbitals or the densities; therefore the search for improved approximations must also consider nonlocal energy densities such as some percentage of the Hartree–Fock energy, leading to what are called hybrid functionals,<sup>22</sup> and/or nonlocal approximations to electron correlation, leading to so-called van der Waals functionals<sup>23</sup> or doubly hybrid functionals.<sup>24,25</sup> There are various kinds of hybrid functionals, but we especially distinguish global hybrids, where the percentage  $X$  of Hartree–Fock exchange is a constant, and range-separated (RS) hybrids, where it is a nonconstant function of interelectronic distance. Open Access Article. Published on 06 April 2016. Downloaded on 1/1/2025 4:32:51 AM. This article is licensed under a [Creative Commons Attribution 3.0 Unported Licence.](http://creativecommons.org/licenses/by/3.0/) **[View Article Online](https://doi.org/10.1039/c6sc00705h)**

Many research groups have proposed approximations to the XCF containing some or all of the elements reviewed in the previous paragraph, and we will give references to many of these below. Here we mention though that the new functional presented here builds on previous work in our group, especially the GAM<sup>26</sup> functional, which is an NGA, the M06-L<sup>27</sup> functional, which is a meta-GGA, the MN12-L<sup>28</sup> and MN15-L<sup>29</sup> functionals, which are meta-NGAs, the PW6B95,<sup>30</sup> M06,<sup>31</sup> M06-2X,<sup>31</sup> and M08-HX<sup>32</sup> functionals, which are global-hybrid meta-GGAs, and the MN12-SX<sup>33</sup> functional, which is a range-separated-hybrid meta-NGA. Even though these and many functionals from other groups (many of which are listed in Section 5 and discussed in Section 6) have had considerable success, there are still some important and challenging areas where exchange–correlation functionals can be improved, and we particularly single out the difficulty of finding a single functional that is accurate both for inherently multi-configurational systems, such as many systems containing transition metals, and for barrier heights. (We will follow the common convention of labeling inherently multi-configurational systems as "multireference" systems; such systems include those with near-degeneracy correlation, also called static correlation. Systems that are qualitatively well described by a single configuration state function will be called "single-reference" systems.)

In the present article, building on the work summarized above, we present a global-hybrid meta-NGA, called MN15, that is able to achieve better simultaneous accuracy for these two properties (energies of multireference systems and barrier heights) than any previous functional and also has especially good performance for noncovalent interactions and excitation energies. Such a functional should be useful for a wide variety of application areas, including catalysis, the quest for new energy, nanotechnology, functional materials, synthesis, biochemistry, the drive toward a cleaner environment, and understanding chemical dynamics.

The article proper is kept brief for a general readership. Detailed results for specialists are presented in ESI.†

### 2. Databases

The largest database we consider in the present paper is called Database 2015B, which has 481 data and is a combination of databases AME471 and MS10. Database AME471 contains 471 atomic and molecular energies. It is obtained by extending a previous database, AME422,<sup>29</sup> by the addition of 49 new data, of which 36 are noncovalent interactions (subdatabase S6x6, which means six dimers at six intermonomer distances taken from the  $S66x8$  database<sup>34</sup>) and 13 alkyl bond dissociation energies from a recent paper of our group.<sup>35</sup> The distribution of data in AME471 across its various subdatabases is illustrated in Fig. 1, which shows that it is a very diverse database. Database 2015B also includes database MS10, which has ten molecular structure data, as used in a previous work.<sup>29</sup> All the data in Database 2015B are shown in Table 1 with their acronyms, brief descriptions, inverse weights for training (which are explained in Section 4), and references.

Database 2015B was used for both training and testing of the MN15 functional, and we calculated this full database for 82 other functionals for comparison. In Table 2, we list additional databases that we used only for testing the performance of the new MN15 functional and for more limited comparisons to other functionals. These data are not used for parameterization.

One of the databases in Table 2, namely EE69,<sup>36</sup> was used to test the performance of functionals on electronic excitation energies. This database contains the experimental vertical excitation energies of 30 valence and 39 Rydberg excitations of 11 organic molecules, namely acetaldehyde, acetone, ethylene, formaldehyde, isobutene, pyrazine, pyridazine, pyridine, pyrimidine, s-tetrazine, and s-trans-butadiene.

### 3. Computational details

All the calculations are performed by a locally modified version<sup>37</sup> of the Gaussian 09 program.<sup>38</sup> Grids, basis sets, and geometries are given in the ESI.†

Vertical electronic excitation energies of the molecules in the EE69 database were calculated by linear-response time-dependent density functional theory (TDDFT).<sup>39</sup> Twenty excitation energies were computed for each molecule. To compare the computed values with experiment, we made the assignments based on state symmetry as used and discussed in previous work.<sup>40</sup>–<sup>42</sup>



Fig. 1 The percentage of all atomic and molecular databases (AME471), the number after a name means the number of data in this database, for example, SR-MGM-BE9 mean 9 pieces of bonding energy data in this database. The explanation of the these names are shown in Table 1.

### 4. Design and parameterization of the MN15 functional

The LSDA exchange–correlation energy is a starting point for the MN15 functional. It can be written as:

$$
E_{\rm xc}^{\rm LSDA} = E_{\rm x}^{\rm LSDA} + E_{\rm c}^{\rm LSDA}
$$
  
= 
$$
\int dr \sum_{\sigma=\alpha}^{\beta} \rho_{\sigma} \varepsilon_{\rm xc}^{\rm UEG}(\rho_{\sigma}) + \int dr \rho \varepsilon_{\rm c}^{\rm UEG}(\rho_{\alpha}, \rho_{\beta})
$$
 (1)

where  $E_{\rm x}^{\rm LSDA}$  is the LSDA exchange energy,  $E_{\rm c}^{\rm LSDA}$  is the LSDA correlation energy,  $\varepsilon_{\rm xo}^{\rm UEG}$  is the exchange energy per electron with spin  $\sigma,$ and  $\varepsilon_{\rm c}^{\rm UEG}$  is the correlation energy per electron for which we use the parameterization of Perdew and Wang.<sup>13</sup> The MN15 functional is

$$
E_{\text{xc}}^{\text{MN15}} = \frac{X}{100} E_{\text{x}}^{\text{HF}} + E_{\text{nxc}} + E_{\text{c}}
$$
 (2)

where  $E_{\rm x}^{\rm HF}$  is the nonlocal Hartree–Fock exchange energy computed from the Kohn–Sham orbitals, X is the percentage of Hartree–Fock exchange,  $E_{\text{nxc}}$  is the nonseparable local exchange–correlation energy, and  $E_c$  is the additional correlation energy. The local terms are given by

$$
E_{\text{nxc}} = \int \mathrm{d}r \sum_{\sigma=\alpha}^{\beta} \rho_{\sigma} \left\{ \varepsilon_{\text{x}\sigma}^{\text{LSDA}}(\rho_{\sigma}) \sum_{i=0}^{3} \sum_{j=0}^{3-i} \right. \\ \times \sum_{k=0}^{5-i-j} a_{ijk} \left\{ v_{\text{x}\sigma}(\rho_{\sigma}) \right\}^{i} \left\{ u_{\text{x}\sigma}(s_{\sigma}) \right\}^{j} \left\{ w_{\sigma}(\rho_{\sigma}, \tau_{\sigma}) \right\}^{k} \right\} \tag{3}
$$

where the variables  $v_{xo}$ ,  $u_{xo}$ , and  $w_o$  are the same functions as we used in the MN15-L functional,<sup>29</sup> and

$$
E_{c} = \sum_{\sigma} \int dr \rho \varepsilon_{c}^{LSDA} (\rho_{\alpha}, \rho_{\beta}) \sum_{i=0}^{8} b_{i} \{w(\rho, \tau)\}^{i}
$$

$$
+ \int dr \rho H^{PBE} (\rho_{\alpha}, \rho_{\beta}, s) \sum_{i=0}^{8} c_{i} \{w(\rho, \tau)\}^{i}
$$
(4)

where w is a function defined in a previous paper,<sup>32</sup> and  $H^{\text{PBE}}$  is the PBE gradient correction<sup>43</sup> for the correlation energy. Eqn  $(3)$ is the same form as used in the MN12-L functional,<sup>28</sup> and eqn  $(4)$  is the same as introduced<sup>32</sup> for M08-HX except that the upper limits of sums have been reduced to 8. We optimized the value of X along with the parameters  $a_{ijk}$ ,  $b_i$ , and  $c_i$  in eqn (3) and (4) by minimizing

$$
U = \sum_{n=1}^{32} R_n / I_n + \lambda (a + b + c)
$$
 (5)

where  $R_n$  and  $I_n$  are respectively the root mean squared error and inverse weight of database or subset  $n$  in Table 1. Each term in the sum of eqn (5) corresponds to one of the 32 rows of the table that has an entry in the inverse weight column. The product of  $\lambda$  and  $(a + b + c)$  is a smoothness restraint, which has the same form as used previously.<sup>26,29</sup> In particular, we have



Table 1 Databases included in Database 2015B<sup>ª</sup>  $\blacksquare$  Table 1 Databases included in Database 2015 $\mathsf{B}^a$ 

 $\sim$ 







$$
a = \sum_{i=0}^{3} \sum_{j=0}^{3-i} \sum_{k=0}^{4-i-j} (a_{i,j,k} - a_{i,j,k+1})^2
$$
(6)

$$
b = \sum_{i=0}^{7} (b_i - b_{i+1})^2
$$
 (7)

$$
c = \sum_{i=0}^{7} (c_i - c_{i+1})^2
$$
 (8)

We choose the same value, namely 0.01, of  $\lambda$  as we used in the MN15-L parametrization.<sup>29</sup> All the optimized parameters are shown in Table 3.

The final functional is not overly sensitive to the relative number of data in each subdatabase because the inverse weights are changed manually (by trial and error) to try to obtain balanced performance across the full collection of data.

### 5. Density functionals for comparison

In order to evaluate the performance of the new MN15 functional, we compare the results obtained with the MN15 functional to those obtained with 82 previously developed density functional approximations for the entire Database 2015B. The results for 48 of the previous functionals are compared to the results for the new MN15 functional in the article proper, and the results for the remaining 34 functionals are in ESI.† The 49 functionals considered in the article proper are sorted into types as follows:

 $\bullet$  LSDA: GKSVWN5.7,44,45

 $\bullet$  GGA: SOGGA,<sup>46</sup> PBEsol,<sup>47</sup> SOGGA11,<sup>48</sup> BP86,<sup>49,50</sup> BLYP,<sup>49,51</sup> PW91,<sup>52</sup> BPW91,<sup>49,52</sup> PBE,<sup>43</sup> mPWPW,<sup>53</sup> revPBE,<sup>54</sup> RPBE,<sup>55</sup> HCTH407,<sup>56</sup> OLYP,<sup>51,57</sup> and OreLYP.<sup>51,57,58</sup>

 $\bullet$  NGA: N12<sup>18</sup> and GAM.<sup>26</sup>

 $\bullet$  meta-GGA: VSXC,<sup>59</sup>  $\tau$ -HCTC,<sup>60</sup> TPSS,<sup>61</sup> M06-L,<sup>27</sup> revTPSS,<sup>62</sup> M11-L,<sup>63</sup> and MGGA MS2.<sup>64</sup>

• meta-NGA: MN12-L<sup>19</sup> and MN15-L.<sup>29</sup>

 $\bullet$  Global-hybrid GGA: B3LYP,  $49,51,65$  PBE0,  $66$  B98,  $67$  B97-1,  $68$ O3LYP,<sup>69</sup> B97-3,<sup>70</sup> and SOGGA11-X.<sup>71</sup>

• Range-separated hybrid GGA: CAM-B3LYP,<sup>72</sup> LC- $\omega$ PBE,<sup>73-76</sup> HSE06,<sup>77,78</sup> ωB97,<sup>79</sup> and ωB97X.<sup>79</sup>

 Range-separated hybrid GGA plus molecular mechanics (also called empirical dispersion correction): ωB97X-D.<sup>80</sup>

 $\bullet$  Global-hybrid meta-GGA: TPSSh,  $s_1$   $\tau$ -HCTHhyb,  $s_2$ BB1K,<sup>14,17,83</sup> BMK,<sup>84</sup> PW6B95,<sup>30</sup> M06,<sup>31</sup> M06-2X,<sup>31</sup> and M08-HX.<sup>32</sup>

Global-hybrid meta-NGA: MN15.

• Range-separated hybrid meta-GGA: M11.85

For each of the functionals in the above list, Table 4 shows the percentage of nonlocal Hartree–Fock exchange, the year in which the functional was published, and the original reference or references.

We do not consider the more expensive doubly hybrid functionals (functionals with nonlocal correlation) in this article, except briefly in Sections 6.2 and 6.4.

#### Table 2 Databases for testing only $\alpha$



Table 3 Optimized parameters of the MN15 functional

	<b>Edge Article</b>							<b>Chemical Science</b>	
	Table 2 Databases for testing only $^a$								
n	Database		Description					Reference(s)	
$\mathbf{1}$	$SBG31^b$				Semiconductor band gaps (31 data)			93	
2	WCCR10			Ligand dissociation energies of large cationic TM complexes (10 data)					
	$S492^c$	Interaction energies relevant to bimolecular structures (492 data)						92 34	
$\overline{4}$	TMBH21		TM reaction barrier heights (21 data)					89-91	
	EE69		Excitation energies of 30 valence and 39 Rydberg states of 11 organic molecules (69 data)					40	
5 6	TMDBL7		Bond lengths of homonuclear TM dimers (7 data)				94		
$\overline{7}$	<b>SE47</b>						Semi-experimental structures of 47 organic molecules (193 data) $\epsilon$	95	
			Table 3 Optimized parameters of the MN15 functional				· IsoE14: 14 isomerization energies, in particular IsoL6		
Exchange				$X^a$ Correlation					
							4pIsoE4, and 2pIsoE4. · HCTC20: 20 hydrocarbon thermochemical data, in partic		
	0.073852235	$a_{102}$	6.89391326	$b_0$	1.093250748	44	ular $\pi$ TC13 and HC7.		
	$-0.839976156$	$a_{103}$	2.489813993	b <sub>1</sub>	$-0.269735037$		• MS10: 10 molecular structures, in particular DGL6 and		
	$-3.082660125$	$a_{104}$	1.454724691	b <sub>2</sub>	6.368997613		DGH4.		
	$-1.02881285$	$a_{110}$	$-5.054324071$	$b_3$	$-0.245337101$				
	$-0.811697255$	$a_{111}$	2.35273334	$b_4$	$-1.587103441$		• AME454xAE: 454 atomic and molecular energies, ir		
	$-0.063404387$	$a_{112}$	1.299104132	$b_5$	0.124698862		particular AME471 without AE17 (which denotes 17 absolute		
	2.54805518	$a_{113}$	1.203168217	b <sub>6</sub>	1.605819855		energies from H to Cl).		
	$-5.031578906$	$a_{120}$	0.121595877	b <sub>7</sub>	0.466206031		· NC87: 87 noncovalent interactions, in particular the		
	0.31702159	$a_{121}$	8.048348238	$b_8$	3.484978654		combination of NCCE30, NGDWI21, and S6x6.		
	2.981868205	$a_{122}$	21.91203659						
	$-0.749503735$	$a_{200}$	$-1.852335832$	$c_0$	1.427424993				
	0.231825661	$a_{201}$	$-3.4722735$	$c_{1}$	$-3.57883682$		Another way to classify some of the data is into single		
	1.261961411	$a_{202}$	$-1.564591493$	$c_{2}$	7.398727547		reference and multi-reference systems (these terms are		
	1.665920815	$a_{203}$	$-2.29578769$	$c_3$	3.927810559		explained in the introduction). Database 2015B has 481 data: 10		
	7.483304941	$a_{210}$	3.666482991	$c_4$	2.789804639		structural data (MS10) and 471 energetic data (AME471). If we		
	$-2.544245723$	$a_{211}$	10.87074639	$c_{5}$	4.988320462		exclude 87 noncovalent interactions (NC87) and 17 absolute		
	1.384720031	$a_{212}$	9.696691388	$c_{6}$	3.079464318		energies (AE17) from AME471, we have 367 energetic data that		
	6.902569885	$a_{300}$	0.630701064	$c_7$	3.521636859				
$a_{000}$ $a_{001}$ $a_{002}$ $a_{003}$ $a_{004}$ $a_{005}$ $a_{010}$ $a_{011}$ $a_{012}$ $a_{013}$ $a_{014}$ $a_{020}$ $a_{021}$ $a_{022}$ $a_{023}$ $a_{030}$ $a_{031}$ $a_{032}$ $a_{100}$ $a_{101}$	1.657399451 2.98526709	$a_{301}$ $a_{302}$	$-0.505825216$ $-3.562354535$	$c_{8}$	4.769671992		can be classified as either single-reference data or multi-refer ence data. We classified these based on the generalized B1		

 $a<sup>a</sup>$  X denotes the percentage of Hartree–Fock exchange.

### 6. Performance of the MN15 functional

### 6.1. Performance for database 2015B

In order to compare the functionals for different categories of data, we used the following combinations of the subdatabases that are explained more fully in Table 1:

 MGBE150: 150 main-group bond energies, in particular SR-MGM-BE9, SR-MGN-BE107, MR-MGM-BE4, MR-MGN-BE17, and ABDE13.

 TMBE33: 33 transition-metal bond energies, in particular SR-TM-BE17, MR-TM-BE13, and MR-TMD-BE3.

 BH76: 76 reaction barrier heights, in particular HTBH38 and NHTBH38.

 EE18: 18 excitation energies, in particular 3dEE8, 4dAEE5, and pEE5.

Another way to classify some of the data is into singlereference and multi-reference systems (these terms are explained in the introduction). Database 2015B has 481 data: 10 structural data (MS10) and 471 energetic data (AME471). If we exclude 87 noncovalent interactions (NC87) and 17 absolute energies (AE17) from AME471, we have 367 energetic data that can be classified as either single-reference data or multi-reference data. We classified these based on the generalized B1 diagnostic,<sup>86–88</sup> which separates these data into 54 multi-reference data (subdatabase MR54) and 313 single-reference data (subdatabase SR313). The numbers of multi-reference systems and single-reference systems in each database are shown in Table 5. Table 5 shows that the MR54 subset contains 13 transition-metal data from MR-TM-BE13 but also a significant amount of other data, including main-group bond energies, reaction barrier heights (which do not contain transitionmetals), hydrocarbon systems, etc. Therefore, the MR54 subset includes a variety of kinds of systems.

The performance of the new MN15 functional and the 48 other functionals in Table 4 has been evaluated for the entire Database 2015B. The results of these tests are given in Tables 6– 9 for the energetic part of Database 2015B and in Table 10 for the structural portion of Database 2015B. In addition we compared MN15 against results in the literature for the following additional test sets: the noncovalent interactions of the S66 and S66x8 databases, $34$  the excitation energies of selected organic molecules in the EE69 database,<sup>40</sup> transitionmetal reaction barrier heights in the TMBH21 database,<sup>89-91</sup> transition-metal coordination energies in the WCCR10 database,<sup>92</sup> semiconductor band gaps in the SBG31 database,<sup>93</sup>



 $^a$  X is the percentage of nonlocal Hartree–Fock exchange. When a range is given, the first value is for small interelectronic distances, and the second value is for large interelectronic distances. Details of the functional form that joins these regions of interelectronic separation are given in the references. <sup>b</sup> GVWN5 denotes the Gáspár approximation for exchange and the VWN5 fit to the correlation energy; this is an example of the local spin density approximation (LSDA), and it has the keyword SVWN5 in the Gaussian 09 program. Note that Kohn-Sham exchange is the same as Gáspár exchange, but Slater exchange (not tested here) is greater by a factor of 1.5. <sup>c</sup> PW91 formally satisfies the gradient expansion for exchange to second order but only at such small values of the gradient that for practical purposes it should be grouped with functionals that do not satisfy the gradient expansion to second order. <sup>d</sup> MM denotes molecular mechanics (also called empirical dispersion correction), which in this case corresponds to atom–atom pairwise damped dispersion terms added post-SCF to the calculated energy.

transition-metal dimer bond lengths in the TMBDL7 database,<sup>94</sup> and geometrical parameters of 47 selected organic molecules.<sup>95</sup> These evaluations and comparisons are discussed in the following subsections.

All energetic quantities in this paper are calculated from Born–Oppenheimer energies without including vibrational energies. Therefore when the reference data are from experiment, we removed zero-point vibrational energy and thermal





 $a$  The 26 energetic databases from Table 1 are shown here with the number of multi-reference systems and single-reference systems defined by generalized B1 diagnostics.<sup>86-88</sup>

vibration-rotational energies, if present in the original data. For example, bond energies are equilibrium dissociation energies  $(D_e)$ , not ground-state dissociation energies  $(D_0)$ . Most of the bond energies in the databases are average bond energies calculated as atomization energy per bond. Even when all bonds are not atomized, we calculate bond energies on a per bond basis. For example one of the bond-breaking processes in subdatabase PdBE2 is the dissociation of  $Pd(PH<sub>3</sub>)<sub>2</sub>C<sub>6</sub>H<sub>8</sub>$  into  $Pd(PH<sub>3</sub>)<sub>2</sub>$  and  $C<sub>6</sub>H<sub>8</sub>$ . As shown in Fig. 2, this involves breaking two bonds, so the energy of dissociation is divided by 2. Further

Table 6  $\,$  MUE (kcal mol $^{-1}$ ) for the AME471 database and its subdatabases: LSDA and other gradient approximations



 $^a$  The MGBE150 database consists of SR-MGM-BE9, SR-MGN-BE107, MR-MGM-BE4, MR-MGN-BE17, and ABDE13.  $^b$  The TMBE33 database consists of SR-TM-BE17, MR-TM-BE13, and MR-TMD-BE3.  $\epsilon$  The BH76 database consists of HTBH38/08 and NHTBH38/08.  $^d$  The NC87 database consists of NGDWI21, S6x6, NCCE23 and CT7. <sup>e</sup> The EE18 database consists of 3dEE8, 4dAEE5, and pEE5. The IsoE14 consists of 2pIsoE4, 4pIsoE4, and IsoL6/ 11. The HCTC20 subdatabase consists of HC7/11 and  $\pi$ TC13.  $^h$  The AME471 database consists all the 27 subdatabases (from 1 to 27 in Table 1) and the AME454xAE consists all the subdatabases except AE17.

<code>Table 7 MUE</code> (kcal mol $^{-1}$ ) for the AME471 database and its subdatabases: meta-GGAs, MN12-L, MN15-L, and global-hybrid GGAs



 $^a$  The MGBE150 database consists of SR-MGM-BE9, SR-MGN-BE107, MR-MGM-BE4, MR-MGN-BE17, and ABDE13.  $^b$  The TMBE33 database consists of SR-TM-BE17, MR-TM-BE13, and MR-TMD-BE3. <sup>c</sup> The BH76 database consists of HTBH38/08 and NHTBH38/08. <sup>d</sup> The NC87 database consists of NGDWI21, S6x6, NCCE23 and CT7.<sup>e</sup> The EE18 database consists of 3dEE8, 4dAEE5, and pEE5.<sup>f</sup> The IsoE14 consists of 2pIsoE4, 4pIsoE4, and IsoL6/ 11. The HCTC20 subdatabase consists of HC7/11 and  $\pi$ TC13.  $^h$  The AME471 database consists all the 27 subdatabases (from 1 to 27 in Table 1) and the AME454xAE consists all the subdatabases except AE17.

details of the molecules in the databases, the bonds broken, and the removal of vibrational energy from experiment are given in ref. 26 and in the references in Tables 1 and 3.

In evaluating performance, we always use the unweighted mean unsigned error (MUE) over the data in a database, subdatabase, or specified collection of subdatabases. Thus, even though the MN15 functional was optimized using root-meansquare errors and weights, performance is consistently measured with unweighted MUEs.

We show the performance of the MN15 functional for AME471 and the 11 combined energetic subdatabases delineated above in Tables 6–8. Table 6 has the performance of

a local spin density approximation (LSDA) and 16 gradient approximations (GAs); Table 7 has the performance of nine meta gradient approximations (meta-GAs) and seven globalhybrid gradient approximations (hybrid GAs); and Table 8 has the performance of six range-separated hybrid gradient approximations (RS-hybrid GAs), nine global-hybrid meta gradient approximations (hybrid meta-GAs), and one rangeseparated-hybrid meta gradient approximation (RS-hybrid meta-GA).

Tables 6–8 show that the MN15 functional gives the best result for main-group bond energies (MGBE150) with an MUE of 1.36 kcal mol<sup>-1</sup>, followed by MN15-L, M06-2X, and M06 with

<code>Table 8 MUE</code> (kcal mol $^{-1}$ ) for the AME471 database and its subdatabases: range-separated-hybrid GGAs, global-hybrid meta-GGAs, MN15 Type RSH GGA RSH GGA + MM GH mGGA RSH mGGA mNGA GH Functional CAM-B3LYP ωPBE HSE06 ωB97 ωB97X ωB97X-D TPSSh τ-HCTHhyb BB1K BMK PW6B95 M06 M06-2X M08-HX M11 **- MN1**5 LC- $MGBE150^a$  3.09 3.47 2.97 2.46 2.42 2.23 3.76 2.37 3.63 2.06 2.65 1.90 1.87 2.74 2.36 1.36  $\text{TMBE3}^{b}$  10.75 12.43 9.92 10.31 10.57 8.70 6.78 5.62 16.20 13.02 9.50 7.10 19.11 17.59 14.36 5.54 BH76<sup>c</sup> 2.90 1.77 3.98 2.15 2.45 3.05 6.39 4.88 1.30 1.21 2.98 2.16 1.18 0.97 1.29 1.36  $NCS7^d$  1.35 1.56 1.31 0.55 0.64 0.30 1.92 1.61 1.56 1.68 1.03 0.67 0.35 0.37 0.39 0.25 EE18<sup>e</sup> 6.20 8.46 8.09 11.51 8.94 8.30 6.46 9.00 6.58 6.74 5.34 8.45 7.86 5.78 8.99 6.28  $\mathrm{IsoE14}^{f}$  2.89 1.55 1.99 1.42 1.79 1.75 3.02 2.50 1.68 1.54 2.01 1.66 1.94 1.26 1.74 1.41  $HCTC20<sup>g</sup>$  4.57 8.96 6.60 6.58 5.21 5.68 7.65 7.25 7.23 5.09 5.24 3.83 1.72 2.93 2.77 3.59 AME454xAE<sup>h</sup> 3.59 4.02 3.85 3.39 3.20 2.99 4.52 3.57 4.08 3.07 3.21 2.59 3.11 3.25 3.20 1.88 AME471<sup>h</sup> 3.85 4.79 4.89 3.49 3.29 3.09 4.91 3.66 4.49 3.57 6.65 2.66 3.08 3.28 3.41 2.08 MR54 9.64 11.81 10.18 10.26 9.72 8.89 9.41 6.38 15.20 10.03 8.95 5.43 13.33 13.62 10.35 4.75 SR313 3.20 3.39 3.54 3.05 2.84 2.77 4.43 3.68 2.85 2.30 2.86 2.71 2.13 2.29 2.80 1.85

 $a$  The MGBE150 database consists of SR-MGM-BE9, SR-MGN-BE107, MR-MGM-BE4, MR-MGN-BE17, and ABDE13.  $b$  The TMBE33 database consists of SR-TM-BE17, MR-TM-BE13, and MR-TMD-BE3. <sup>c</sup> The BH76 database consists of HTBH38/08 and NHTBH38/08. <sup>d</sup> The NC87 database consists of NGDWI21, S6x6, NCCE23 and CT7. <sup>e</sup> The EE18 database consists of 3dEE8, 4dAEE5, and pEE5. <sup>f</sup> The IsoE14 consists of 2pIsoE4, 4pIsoE4, and IsoL6 11.  $\frac{1}{2}$  The HCTC20 subdatabase consists of HC7/11 and  $\pi$ TC13.  $^h$  The AME471 database consists all the 27 subdatabases (from 1 to 27 in Table 1) and the AME454xAE consists all the subdatabases except AE17.





<sup>a</sup> The seven intermolecular charge transfer systems included in the database are  $C_2H_4\cdots F_2$ ,  $N_{13}\cdots C_2$ ,  $H_2\cdots C_2$ ,  $H_3\cdots C_2$ ,  $H_2O\cdots C_2$ ,  $H_2O\cdots$ and  $NH_3\cdots$ ClF.





 $a$ <sup>a</sup> The MS10 database consists of DGL6 and DGH4 subdatabases. The functionals are listed in the order of increasing value in the last column.



Fig. 2 The structure of Pd(PH<sub>3</sub>)<sub>2</sub>C<sub>6</sub>H<sub>8</sub> in subdatabase PdBE2.

MUEs of 1.84, 1.87, and 1.90 kcal mol<sup>-1</sup> respectively. The MN15 functional gives the fourth best result for transition-metal bond energies (TMBE33) with an MUE of 5.54 kcal mol $^{-1}$ , with B97-1, MN15-L, and M06-L being the top three with MUEs of 5.04, 5.16, and 5.29 kcal  $mol^{-1}$ , respectively. The other functional that gives an MUE smaller than 6.0 kcal mol $^{-1}$  for this subset of the data is  $\tau$ -HCTHhyb, with an MUE of 5.62 kcal mol<sup>-1</sup>.

We found that it is hard to get simultaneously good accuracy for certain pairs of databases, even when both are present in the training set. For example, a functional that gives good results for transition-metal bond energies does not usually work very well for reaction barrier heights, and vice versa. Incorporating Hartree–Fock exchange usually improves the performance on hydrocarbons, weak interactions, and barrier heights, but it usually worsens the results for multi-reference systems. Therefore, the mathematical form of the functional and a carefully selected set of diverse databases are very important for the design of a universal and highly transferable functional such as MN15. For the reaction barrier heights database (BH76), the top six best functionals are M08-HX, M06-2X, BMK, BB1K, M11, and MN15, which all have an MUE smaller than 1.50 kcal mol $^{-1}$ .

The MN15 functional gives the second best results for the isomerization energies database (IsoE14), with an MUE of 1.41 kcal mol<sup>-1</sup>, following only M08-HX with an MUE of 1.26 kcal  $mol<sup>-1</sup>$ . The MN15 functional gives the fourth best results for the hydrocarbon thermochemistry database (HCTC20), with an MUE of 3.59 kcal mol<sup>-1</sup>. The top three functionals for HCTC20 are M06-2X with an MUE of 1.72 kcal mol<sup>-1</sup>, M11 with an MUE of 2.77 kcal mol<sup>-1</sup>, and M08-HX with an MUE of 2.93 kcal mol<sup>-1</sup>.

The MN15 functional gives the best results for the noncovalent interactions (NC87), with an MUE of 0.25 kcal mol<sup>-1</sup>, followed by uB97X-D, M06-2X, M08-HX, M11, uB97, M06-L,

uB97X, and M06, which have MUEs of 0.30, 0.35, 0.37, 0.39, 0.55, 0.57, 0.64, and 0.67 kcal  $\mathrm{mol}^{-1}$ , respectively.

For the excitation energies database (EE18), the top eight functionals are MN15-L, SOGGA11-X, PW6B95, M08-HX, O3LYP, B97-3, CAM-B3LYP and MN15, which are the only functionals that have MUEs for EE18 smaller than 6.30 kcal  $\mathrm{mol}^{-1}.$  It is especially noteworthy that MN15-L gives an MUE of only 3.71 kcal mol<sup>-1</sup> and the other five give MUEs in the range 5.00–6.30 kcal mol<sup>-1</sup>.

The MN15 functional is the only one that gives MUEs no larger than 2.10 kcal mol<sup>-1</sup> for both AME471 with 471 energetic data and AME454xAE with 454 energetic data excluding absolute energies (AE17). For the SR313 database of 313 singlereference data, the top five functionals are MN15, M06-2X, MN15-L, M08-HX, and BMK with MUEs of 1.85 kcal  $\mathrm{mol}^{-1}$ , 2.13 kcal mol $^{-1}$ , 2.15 kcal mol $^{-1}$ , 2.29 kcal mol $^{-1}$ , and 2.30 kcal  $mol^{-1}$  respectively. For the MR54 database of 54 multi-reference data, MN15-L and MN15 are the only two functionals that give an MUE smaller than 5.00 kcal mol $^{-1},$  and M06 and M06-L are the only two functionals with MUEs in the range of 5.00 kcal  $\mathrm{mol}^{-1}$  and 6.00 kcal  $\mathrm{mol}^{-1}$ .

In order to show the performance for intermolecular charge transfer, we found the ten functionals (out of 83 tested) that perform best for the CT7 charge transfer database, and we list their MUEs for this database in Table 9. The MN15-L and MN15 functionals are essentially tied for third best with an MUE of  $0.25$  kcal mol<sup>-1</sup>. The MN15-L and MGGA\_MS2 functionals are the only two local functionals that gives an MUE smaller than 0.30 kcal mol<sup>-1</sup> for this database. The MUEs of the other 73 functionals against CT7 database are shown in the ESI.†

Table 10 shows the performance of 49 functionals including MN15 for the molecular structure database MS10 and its subdatabases DGL6 and DGH4. The MN15 functional is the tenth best for DGL6 with an MUE of 0.005  $\AA$  and the third best for DGH4 with an MUE of  $0.008$  Å. Because it is rare for a functional to do well on both databases, MN15 is the best overall when these databases are combined into MS10. We conclude that the MN15 functional gives good results not only for energies but also for geometries.

Some functionals that are good for one property (such as main-group bond energies, transition-metal bond energies, noncovalent interactions, etc.), as seen for certain databases in Tables 6–10, fail badly for other properties. Parameterizing a functional that is good for several properties is the challenge of designing and parameterizing a universal functional. Average errors over diverse databases such as ME471 are one way to gauge the universality of a functional, but such mean errors are sensitive to the number of data of each type and can underweight performance in categories with intrinsically smaller errors (such as noncovalent interactions) unless those categories have a compensatingly large number of data, but the amount of compensation is arbitrary.

A better measure, at least for relative success compared to other functionals, is to look at the ranking of the mean unsigned error in each category. To illustrate the high universality of the MN15 functional, we consider Table 11, which lists the rankings for 28 subdatabases of 12 selected functionals

among the set of 83 functionals that were tested for the entire Database 2015B. The table also shows the average ranking, which is a way to summarize combined performance on diverse databases that does not suffer from the two deficiencies mentioned at the start of this paragraph, and it also shows the lowest ranking. The MN15 functional gives the best average ranking, which is 9, with MN15-L being second with an average ranking of 17. The average ranking of the other functionals is in the range of 22–52. Furthermore, every other functional in the list has at least one ranking of 62 or lower, whereas MN15 ranks lower than 29th in none of the 28 categories. MN15-L ranks lower than 29th in 4 categories, M06, M06-2X, and uB97X-D rank lower than 29th in 9–11 categories, M06-L ranks lower than 29th in 14 categories, and the other functionals in the table rank lower than 29th in 19–23 categories.

#### 6.2. Performance for S66 and S66x8 databases

The S66x8 and S66 databases<sup>34</sup> were developed primarily for the validation of new quantum chemical methods. The S66 database contains accurate interaction energies for 66 noncovalently bound complexes at the equilibrium van der Waals geometry of the complex, and it is divided into three sub-databases: complexes that are damped-dispersion-dominated (DD23), hydrogen-bonding complexes (HB23), and complexes dominated by a mix of damped dispersion and electrostatics (Mix20). DD23 contains stacked and unstacked aromatic complexes and complexes containing aliphatic molecules; HB23 contains both singly H-bonded and doubly H-bonded complexes. Examples of mixed systems are benzene–water and pentane–acetic acid. The S66x8 database contains not only S66 as a subset but also interaction energies of the 66 complexes at seven other intersubsystem distances, ranging from 0.9 times the van der Waals distance to 2.0 times the van der Waals distance. The average interaction energy in S66 is 5.46 kcal  $mol^{-1}$ , and the average interaction energy in S66x8 is 4.15 kcal  $mol^{-1}$ . Openical Science<br>
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> Goerigk et al.<sup>96</sup> choose to benchmark several methods against the S66 and S66x8 databases, choosing methods based mainly on their good performance relative to other functionals of the same class in previous tests, but also including three functionals based on their widespread use or general interest. In Tables 12 and 13 we compare the performance of MN15 to the performance of the functionals they chose and also to uB97X-D, which is not in their paper but is added to the comparison here. The functionals they chose include functionals with nonlocal correlation terms, which are more expensive but often favored for treating noncovalent complexes because dispersion interactions in the long-range region where the subsystems have no overlap can only be treated by nonlocal correlation.

> S66x8 has 528 data and S66 has 66 data; of these we used only 36 (which is 7%) of the data of S66x8 for training MN15, and we used only six of the data in S66 for training. These subsets are called S6x6 (six systems at 0.90, 0.95, 1.00, 1.05, 1.10, and 1.25 times the equilibrium distances) and S6 (six systems at the equilibrium distances) respectively. The remaining data





Table 12 The performance (kcal mol $^{-1}$ ) of selected density functionals for the S60 and S492 databases and subdatabases<sup>a</sup>



 $^a$  Results for the full databases (DA23, HB23, Mix20, S66, and S66x8) are in the next table. This table includes only functionals without nonlocal correlation and without empirical molecular mechanics correction terms.  $\overset{b}{\sim}$  DD21 is the dispersion-dominated subdatabase.  $\overset{c}{\sim}$  HB20 is the hydrogen bonding subdatabase.  $\frac{d}{dx}$  Mix19 is the mixed subdatabase.

after excluding the 36 data from the 528 data in S66x8 form a test that was called S492 in Table 2. The remaining data after excluding the 6 data from the 66 data in S66 will be called test set S60. Similarly the data remaining after excluding training

data from DD23, HB23, and Mix20 constitute test sets DD21, HB20, and Mix19. Table 12 shows comparisons for the test-only sets and Table 13 shows comparison for the full data sets.

Before considering Tables 12 and 13 in more detail, we provide background contextual comments on dispersion. Dispersion interactions were originally modeled by London, and his formulas apply only at long range where the subsystems have no charge cloud penetration.<sup>97</sup> In the strict meaning of the term, "dispersion force" applies only in such regions. As subsystems begin to overlap there is no unique way to partition the interaction energy into dispersion and the remainder, although there are reasonable ways to do so in an approximate sense.<sup>98</sup> However, when doing this, the dispersion component of the interaction energy no longer satisfies the inverse power laws of long-range dispersion; rather it is damped by overlap. This overlap also leads to Pauli repulsion, and when the intersubsystem distance is reduced to the van der Waals distance, since that represents equilibrium geometry, the repulsive force exactly cancels the attractive force (and therefore is not negligible). Functionals with local correlation, such as MN15 and the other functionals to which we compare in the other sections of this article, cannot treat the inverse power dependence of longrange dispersion, but in principle they can treat attractive noncovalent interactions at the van der Waals geometry, as tested by S66 and its subdatabases, and in principle they can even provide reasonable results at larger distances in the range

Table 13 The performance (kcal mol $^{-1}$ ) of selected density functionals for the S66 and S66x8 databases and subdatabases

DFT	$\mathrm{DD23}^a$	$H B23^b$	Mix20 <sup>c</sup>	S66	S66x8	though none of this data was used for training and even though the other functionals in the table were selected <sup>96</sup> mainly for
Without nonlocal correlation or molecular mechanics (also called empirical dispersion correction)						their good performance on this kind of problem. The MN15 functional gives the second best results for the dispersion
<b>MN15</b>	0.57	0.36	0.32	0.42	0.32	dominated (DD23), mixed (Mix19) and S60 databases, with
$M06-2X$	0.35	0.24	0.25	0.28	0.34	
$M05-2X$	1.03	0.27	0.43	0.58	0.49	M06-2X being the best. The MN15 functional gives the fourth
$M06-L$	0.69	0.39	0.73	0.60	0.51	best results for hydrogen bond (HB20) database with M06-2X
<b>PW6B95</b>	2.39	0.99	1.40	1.60	1.13	M05-2X and M06-L being the first, second, and third. The MN15
MPW1B95	2.76	1.13	1.62	1.85	1.31	functional performs nearly the same in both Tables 12 and 13
PBE	3.63	0.74	1.94	2.11	1.51	It is especially remarkable that MN15 does better than all
$LC$ - $\omega$ PBE	3.85	1.35	2.14	2.46	1.78	three functionals with nonlocal correlation but without molec
<b>TPSS</b>	4.83	1.38	2.77	3.00	2.11	
B3LYP	5.17	1.47	3.00	3.22	2.37	ular mechanics dispersion in Table 13. Functionals with
oTPSS	5.99	2.17	3.55	3.92	2.72	empirical molecular mechanics terms are the best for non-
<b>BLYP</b>	6.15	2.22	3.77	4.06	2.97	covalent interactions, but they are sometimes poorly balanced
revPBE	6.35	3.03	3.91	4.45	3.13	when one consider properties more general than noncovalent
						interactions. For example, Table 11 shows that the wB97X-L
With nonlocal correlation						functional is below average for multi-reference main-group
B2GP-PLYP	1.88	0.32	1.01	1.07	0.81	metal bond energies, for multireference main-group non-metal
PWPB95	1.81	0.95	1.09	1.29	0.92	
B2-PLYP	2.66	0.61	1.51	1.60	1.19	bond energies, for proton affinities, for thermochemistry of $\pi$
With molecular mechanics (also called empirical dispersion correction) added						systems, for 4d atomic excitation energies, for p-block excitation energies, and for multireference transition-metal-dimer bond
PW6B95-D3(BJ)	0.16	0.23	0.15	0.18	0.21	energies.
$\omega$ B97X-D	0.47	0.28	0.22	0.32	0.23	
With nonlocal correlation and molecular mechanics (also called empirical dispersion correction)						6.3. Performance for excitation energy database
DSD-BLYP-D3	0.23	0.36	0.14	0.21	0.16	Because of its modest computational cost and often-useful
$a$ DD23 is the dispersion-dominated subdatabase. $b$ HB23 is the hydrogen bonding subdatabase. <sup>c</sup> Mix20 is the mixed subdatabase.						accuracy, TDDFT is now the workhorse for calculating energies and properties of electronically excited states of large mole cules. However, one of the most vexing problems of TDDFT with functionals in the literature is the difficulty of achieving high

tested by S66x8. Table 12 tests how well this is done by various functionals.

In Table 12 we show the performance of density functionals without nonlocal correlation or molecular mechanics (also called empirical dispersion correction) against data that are only used for testing. In Table 13 we show the results for the whole S66x8 database including the six molecular systems we included in training. We have classified functionals in Table 13 based on whether or not they contain nonlocal correlation or damped-dispersion molecular mechanics terms or both. Note that Table 13 includes results for the oTPSS<sup>99</sup> functional which is not among 83 functionals in Table 4 and in the ESI†. Although the table includes all the functionals without molecular mechanics that were selected by the authors of ref. 96, it includes only two of their examples including molecular mechanics (the best for S66 with local correlation and one with nonlocal correlation that was specifically designed for use with molecular mechanics terms), since molecular mechanics corrections are not the focus of our article. For greater completeness we also include our results for  $\omega$ B97X-D. There are four doubly hybrid functionals included in Table 13, in particular B2-LYP,<sup>100</sup> B2GP-PLYP,<sup>101</sup> DSD-BLYP-D3,<sup>96,102</sup> and PWPB95.<sup>103</sup>

#### 6.3. Performance for excitation energy database

Because of its modest computational cost and often-useful accuracy, TDDFT is now the workhorse for calculating energies and properties of electronically excited states of large molecules. However, one of the most vexing problems of TDDFT with functionals in the literature is the difficulty of achieving high accuracy for valence and Rydberg excitations with the same functional.<sup>40</sup> Local functionals are well-known to underestimate Rydberg excitations severely,<sup>104</sup> causing serious problems in applications to spectroscopy where Rydberg states are intermixed with valence states and in applications to photochemistry where states with Rydberg character play an important role in many photodynamical processes; for example, the repulsive  $n\sigma^*$  state important for the photodissociation of thioanisole has Rydberg character.<sup>105</sup> Even when Rydberg states are high in energy and unimportant, underestimation of their energy by TDDFT brings them down to the energy range of valence states and damages the calculated spectrum. The problem is fundamental and has to do with the self-interaction error and the consequent incorrect asymptotic behavior of the exchange– correlation potential.41,42,104,106,107

Methods have been proposed to solve the problem within the conventional Kohn–Sham framework. Hybrid functionals alleviate this problem by incorporating a fraction of the Hartree– Fock exchange, whose derived exchange potential has the correct asymptotic behavior. Although a hybrid functional generally still does not have the correct asymptotic exchange potential unless it has 100% Hartree–Fock exchange, it improves the potential in the middle range of interelectronic separation, which is most important for Rydberg excitations.<sup>42</sup> However, a high percentage of Hartree–Fock exchange, while Table 14 The mean unsigned errors (MUE, in eV) of 60 selected methods for the vertical excitation energies of 30 valence, 39 Rydberg, and all 69 transitions



 $^{\,a}$  WFT indicates wave function theory; other rows are density functional theory, and X is the percentage of Hartree–Fock exchange. When a range of X is indicated, the first value corresponds to small interelectronic separations, and the second to large interelectronic separations.

helpful for Rydberg excitations, often compromises the accuracy for valence states. A good overall design and balance in the functional, instead of simply raising the fraction of Hartree– Fock exchange, is therefore required to obtain simultaneous accuracy for valence and Rydberg excitations. Range-separated hybrid functionals<sup>73</sup> are another popular way to achieve a better balance of accuracy for valence and Rydberg excitations. A local exchange enhancement scheme has also been proposed for this purpose.<sup>42</sup>

The general trends discussed in the preceding paragraphs can be seen in Table 14, where we compare the performance of MN15 with 60 other methods reported in ref. 36 and 40 for the calculation of valence and Rydberg excitation energies of selected organic molecules (EE69). Many of the functionals in Table 14 are already defined and referenced in Table 5; references<sup>108-121</sup> for the others are given in the table. The table shows that MN15 gives the lowest mean unsigned error averaged over all states and is the only functional that gives an error smaller than 0.3 eV for both valence and Rydberg states. The functionals that do nearly as well, with an error smaller than 0.4 eV for both types of excitations, include M06-2X, BMK, and M05-2X, which are hybrid functionals with a high percentage (42–56%) of Hartree-Fock exchange, and  $\omega$ B97X-D and CAM-B3LYP, which are range-separated hybrid functionals with respectively 100 and 65% Hartree–Fock exchange at large interelectronic separation.

We emphasize that the accuracy for electronic excitations is not an isolated issue but an integral part of the overall performance of a functional. Consistently good performance in both the ground state and excited states is crucial for the application of density functional methods to dynamical problems such as photochemistry. We conclude that MN15 is well suited for photochemical studies.

We also emphasize that in our training set, we only have excitation energies for atoms and for the first excitation energy of the  $Fe<sub>2</sub>$  molecule. No TDDFT calculations were used in training (the few excitation energies in the training set were calculated as the difference in SCF energies, which is possible when the excited state has a different symmetry than the ground state).

#### 6.4. Performance for transition-metal barrier height database

Table 15 tests the performance of the MN15 functional for the transition-metal reaction barrier heights published by Chen et  $al.^{89-91}$  The transition metals involved are Mo, W, Zr, and Re. This is a good test for MN15 because the training set involved no W or Re data, only one Zr datum (the  $Zr_2$  bond energy in SR-TM-BE17), and only three data for Mo (all for monatomic species: Mo and  $Mo<sup>+</sup>$  in IP23 and the quartet-to-sextet transition energy of Mo in 4dAEE5). The results for MN15, MN15-L, and GAM were calculated by the present authors, and the other results in the table are the calculations of Chen et al. The functionals they selected for testing include two doubly hybrid functionals (both of which were also included in Section 6.2), B2GP-PLYP, which gives the best results, and B2-PLYP, which gives the fourth best results. Among the 20 functionals without nonlocal correlation





 $a$  There are respectively 6, 6, 4, and 5 reactions in the database for Mo, W, Zr, and Re reaction barrier heights.  $<sup>b</sup>$  The average mean unsigned</sup> error for all 21 barrier heights.

in Table 15, the MN15 functional gives the second best results, behind M06. In Fig. 3, we show one typical example from this test set, which is a carbon–carbon formation process catalyzed by rhenium. The MN15 functional is the only one that gives good results for both barrier height 1 and barrier height 2. We conclude that the MN15 functional provides good results for transition-metal reaction barrier heights, even though there were no transition-metal barrier heights in the training set.

#### 6.5. Performance for transition-metal coordination database

Table 16 shows the performance of 11 density functionals for the transition-metal coordination database (WCCR10).<sup>92</sup> The



Fig. 3 Reaction barrier heights of rhenium catalyzed reaction.

Table 16 Mean unsigned errors (kcal mol $^{-1}$ ) for the WCCR10 database

Functional <sup>a</sup>	Type	WCCR10
<b>MN15</b>	Hybrid meta-NGA	5.04
$MN15-L$	meta-NGA	5.46
PBE <sub>0</sub>	Hybrid GGA	6.40
<b>GAM</b>	<b>NGA</b>	6.60
<b>PBE</b>	GGA	7.58
TPSSh	Hybrid meta-GGA	7.62
<b>TPSS</b>	GGA	7.84
$B97-D-D2$	$GGA + MM^b$	8.59
B3LYP	Hybrid GGA	9.30
<b>BP86</b>	GGA	9.42
BP86-D3	$GGA + MM^b$	10.62

 $a$  The MN15, MN15-L, and GAM results are from the present calculations, but all other results in this table are from ref. 92.  $<sup>b</sup>$  MM</sup> denotes molecular (also called empirical dispersion correction), which in this case corresponds to atom–atom pairwise damped dispersion terms added post-SCF to the calculated energy.

results for MN15, MN15-L, and GAM were calculated by the present authors, and the other results in the table are calculations from ref. 92. The molecules used for training are relatively small compared to all the molecules in the WCCR10 database. We find that the MN15 functional gives the best results for the WCCR10 database with an MUE of 5.04 kcal  $\mathrm{mol}^{-1}.$  The MN15 functional gives the fourth best results for TMBE33 in Tables 6– 8, with an MUE of 5.54 kcal mol<sup>-1</sup>, which is close to the MUE that MN15 gives for the WCCR10 database. Therefore, the good performance of the MN15 functional for transition-metal bond energies in TMBE33 does transfer well to other databases.

#### 6.6. Performance for semiconductor band gap database

Table 17 shows the performance of 28 density functionals for the semiconductor band gap database  $(SBG31).^{93}$  The functionals included in the table are the same as those in Table S17† of our previous paper<sup>29</sup> plus MN15. The band gaps in this table are calculated, as is widely done,<sup>122</sup> as the orbital energy of the lowest unoccupied crystal orbital minus the orbital energy of the highest occupied crystal orbital. When calculated this way, the MN15 functional gives the eleventh best results for the SBG31 database, with an MUE of 0.92 eV.

This kind of a test of an XCF suffers from the fact that the orbital energies are not physically observable quantities such as the quasiparticle energies appropriate for interpreting optical excitation, photoemission, and inverse photoemission experiments. A better way to calculate the observable band gap for a single-reference system is to use the Green's function– screened potential (GW) method to calculate true quasiparticle energies from the Kohn–Sham single-particle orbitals and orbital energies.<sup>123</sup> It would be interesting to carry out such a calculation with the MN15 potential, but it is beyond the scope of the present paper.

#### 6.7. Performance for transition-metal bond length database

Table 18 shows the performance of 14 density functionals for the homonuclear transition-metal dimer bond length database Table 17 Mean unsigned errors for the SBG31 database in eV



(TMBDL7).<sup>94</sup> The functionals included in the table are the same as those in Table S17† of our previous paper<sup>29</sup> plus MN15. The MN15 functional gives the seventh best results for the TMBDL7 database with an MUE of  $0.045$  A. Interestingly, local functionals often give more accurate bond lengths and vibrational frequencies than nonlocal ones, and this trend is seen for the bond lengths in Table 18. Of the six nonlocal functionals in Table 15 (HSE06, MN15, B3PW91, B3LYP, B97-1, and  $\omega$ B97X-D), MN15 gives the second best MUE, and its MUE of 0.045  $\AA$  may be compared to an average MUE for the five other nonlocal functionals of 0.063  $\AA$ .

#### 6.8. Performance for organic molecule geometry database

Table 19 shows the performance of 13 density functionals for a recently published structural database called SE47, which denotes semi-experimental structures of 47 organic molecules.<sup>95</sup> The functionals included in this table are those selected as the best ranking functionals in Table 10. The SE47 database consists of 193 bond lengths of the 47 organic molecules. In this case MN15 gives the best bond lengths with an MUE of only 0.0033 Å. In Table 19 the nonlocal functionals (hybrid functionals) do better, on average, than the local ones. Since we only have ten molecular structure data (MS10) in our training set, the good performance of the MN15 functional against these 193 bond length shows that our functional is highly transferable to quantities not in our training set.

Table 18 Homonuclear transition-metal dimers: equilibrium bond lengths  $(A)$  and mean unsigned errors as compared to experiment



<sup>a</sup> The experimental values are from ref. 94.  $\rm^b$  The bond lengths in the table and MUE(1) are calculated with the LANL2DZ basis set for comparison with previous work, and MUE(2) is averaged over this basis set and also over the higher-quality def2-TZVP basis set.

Table 19 Mean unsigned errors for the SE47 database in Å



 $a$  The SE47 is a new geometry database published by M. Piccardo et al.<sup>95</sup> There are 193 bond length from 47 organic molecules being calculated by 13 functionals above. The original database SE47 includes both bond lengths and bond angles, however, in the present paper we only compare the bond lengths.  $\frac{b}{b}$  MM denotes molecular mechanics (also called empirical dispersion correction), which in this case corresponds to atom–atom pairwise damped dispersion terms added post-SCF to the calculated energy.

### 7. Concluding remarks

The new functional presented in this article is the culmination of a large body of work. Our group tested the performance of a variety of functionals on various kinds of properties one at a time and used the knowledge so gained to improve our databases. Some of our small databases were built by selecting representative data from larger databases by statistical processes. Our strategy for developing a better density functional has two components; one is the development of a good database for training, and the other is the choice of functional form. In the current round of functional optimization, we first

worked out the best gradient approximation, which is the GAM functional.<sup>26</sup> Next we made the MN15-L functional<sup>29</sup> by adding kinetic energy density to the ingredients in the GAM functional. As we can see from Table 7, averaged over all 471 energetic data in Database 2015B, MN15-L is the best functional among all the local approximations, and it even does better than all of the hybrid functionals in Tables 7 and 8 except for the MN15 functional presented here. The MN15 functional adds a portion of Hartree–Fock exchange to the ingredients of the MN15-L functional, and improves the mean unsigned error on the 471 data from 2.36 kcal mol<sup>-1</sup> to 2.08 kcal ml<sup>-1</sup>. The MN15 functional gives best results for single-reference systems (SR313) and second best results for multi-reference systems (MR54). It has outstanding performance for noncovalent interactions. It also has the best performance for the ten bond distances in the MS10 database. Perhaps most significantly of all, it improves the average ranking over 28 diverse energetic databases from 17 to 9, which means it is a more universal functional than any previous functional.

In order to further test the broad applicability of the new functional, we tested it on the following additional databases: S492, EE69, TMBH21, WCCR10, SBG31, TMBDL7, and SE47. This gives a total of 823 data that are not in our training set in addition to 481 pieces of data in Database 2015B that was used for training and broader testing. MN15 has good performance on the data not included in the training set as well as on the data used for training. It is especially noteworthy that it has good accuracy for both valence and Rydberg electronic excitations, which is not achieved by most other functionals; such simultaneous accuracy is important for applications such as photochemistry.

Many applications of density functional theory, such as understanding and designing new energy materials and catalysts, require more than one property to be accurately modeled. The broad accuracy of MN15 can be very useful for studying these applications.

# Competing financial interests

The authors declare no competing financial interests.

# Acknowledgements

We thank Xiaoyu Li, Xuefei Xu, and Xiaoqing You for sharing the data in ref. 35 prior to publication. This research is supported by the U.S. Department of Energy and the Inorganometallic Catalyst Design Center at the University of Minnesota under Award DE-SC0012702. H.Y. acknowledges a Doctoral Dissertation Fellowship. S.L.L acknowledges support from the Frieda Martha Kunze Fellowship, University of Minnesota. Edge Article<br>
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