# Regularized and Opposite Spin-Scaled Functionals from Møller-Plesset Adiabatic Connection-Higher Accuracy at Lower Cost 

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#### Abstract

Noncovalent interactions (NCIs) play a crucial role in biology, chemistry, material science, and everything in between. To improve pure quantum-chemical simulations of NCIs, we propose a methodology for constructing approximate correlation energies by combining an interpolation along the Møller-Plesset adiabatic connection (MP AC) with a regularization and spin-scaling strategy applied to MP2 correlation energies. This combination yields $c_{o s} \kappa_{\text {os }}$-SPL2, which exhibits superior accuracy for NCIs compared to any of the individual strategies. With the $N^{4}$ formal scaling, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 is competitive or often outperforms more expensive dispersion-corrected double hybrids for NCIs. The accuracy of $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 particularly shines for anionic halogen bonded complexes, where it surpasses standard dispersion-corrected DFT by a factor of 3 to 5 .




Noncovalant interactions (NCIs) play a crucial role in a variety of fields including biology, chemistry, material science, and everything in between. ${ }^{1-13}$ Second order MøllerPlesset perturbation theory (MP2) has been used extensively to study $\mathrm{NCIs}^{2,14}$ because it includes dispersion interactions and also has more favorable scalings compared to other wave function methods such as $\operatorname{CCSD}(\mathrm{T})$. However, MP2 does have its downsides, because it is known to fail for $\pi-\pi$ stacking complexes and generally for NCIs involving highly polarizable molecules. ${ }^{15}$ More recently, it has been observed that MP2 and even the whole MP series have increasingly large errors for large complexes, ${ }^{16}$ restricting the chemical space of NCIs that MP2 can be meaningfully applied to. Multiple methods have been introduced to increase the accuracy of MP2 by manipulating the underlying MP2 equations, ${ }^{15,17-20}$ such as spin-component-scaled (SCS) MP2, ${ }^{17}$ and spin-oppositescaled (SOS) MP2. ${ }^{18}$ SOS-MP2 does not only often improve the accuracy of MP2, but is also cheaper (scales formally as $N^{4}$ ) than the original MP2 $\left(N^{5}\right)$. Another way to improve MP2 energies is to prevent their divergence when the HOMOLUMO orbital gap closes through regularization, ${ }^{15,20}$ and to mix MP2 with density functional theory (DFT) as in double hybrids (e.g., B2PLYP). ${ }^{21-26}$ However, (with some exceptions ${ }^{23-25}$ ), when there is no dispersion correction, double hybrids typically worsen MP2 for NCIs. ${ }^{10,21,22,27-29}$

In our previous work, ${ }^{30}$ we introduced a new class of functionals that directly approximate the Møller-Plesset adiabatic connection (AC), ${ }^{31,32}$ which is the AC that gives an exact expression for the quantum-chemical correlation
energy and has the MP series as weak coupling Taylor expansion. The idea behind these functionals is that by adding information for the large coupling limit, we can introduce a curvature in the underlying AC curve that is completely missed by MP2. ${ }^{33}$ This curvature has been shown to play a very important role in describing NCIs, particularly as the ratio between dispersion and electrostatics in NCIs grows. ${ }^{33}$ Our MP AC approach is fundamentally different from double hybrids, which rely on error cancellations between exact exchange and MP2 and their approximate semilocal counterparts. Instead, our MP AC approach uses the exact information from the weak coupling limit of MP AC—the full amounts of both exact exchange and MP2. One then recovers correlation energy with MP AC, by performing an interpolation between its weak- and strong-coupling limits. The resulting functionals based on this interpolation strategy either provide major improvements or are on par with dispersion-enhanced (double) hybrids and massively improve over MP2, particularly for large, $\pi-\pi$-stacking, and charge-transfer complexes. ${ }^{30}$ The MP AC interpolation approach introduces a negligible additional cost to MP2 calculations, and because of that, it can

[^0]
also be viewed as a zero-cost correction to MP2 correlation energies. Consequently, the computational cost of current MP AC functionals is the same as that of double hybrids. Nevertheless, their $N^{5}$ formal scaling can be prohibitive for large complexes, especially compared to the more feasible $N^{4}$ formal scaling of standard DFT hybrids.

In this work, we demonstrate the robustness of our MP AC functionals by applying them to different variants of MP2 and show how their cost and accuracy can be improved simultaneously. Utilizing this corrective power, we combine the MP AC approach to only the spin-opposite part of MP2 and the regularized version thereof, yielding massive improvements over SOS-MP2 for NCIs. This scaling-lowering strategy ${ }^{18}\left(N^{5} \rightarrow N^{4}\right)$ not only retains the accuracy of the original MP AC approach that uses "bare" MP2, but also gives major improvements across various classes of NCIs.

Figure 1 demonstrates the corrective power of a successful MP AC interpolation form, called SPL2, ${ }^{30}$ i.e., containing two


Figure 1. MAE (Mean Absolute Error) of MP2, SPL2, $c_{\text {os }} \kappa_{\text {os }}-\mathrm{MP} 2$, and $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 for the halogen bonded complexes of B30 ${ }^{34}$ and $\mathrm{X} 40^{35}$ (see Figure 7 for more detail on the complexes and Table 1 for individual methods).
terms of the Seidl, Perdew, and Levy (SPL) ${ }^{36}$ interpolation formula. The individual bars show the mean absolute errors (MAE) for the interaction energies of halogen-bonded complexes. When "bare" MP2 (blue) is applied to these complexes, we get an MAE of slightly less than $1 \mathrm{kcal} / \mathrm{mol}$. Applying SPL2 (the red arrow) to it, the error reduced to less than $0.6 \mathrm{kcal} / \mathrm{mol}$. This is a substantial change relative to the magnitude of the interaction energies of halogen-bonded complexes. Now we consider a cost-lowering strategy and we take only the spin-opposite part of it, rescale it, and apply to it a specific regularization ${ }^{15,20}$ to obtain $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2, which will be described in detail in the following. Once we go from MP2 (blue; $N^{5}$ scaling) to (light blue; $N^{4}$ scaling), the MAE increases slightly. However, SPL2 corrects $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 to yield $c_{\text {os }} \kappa_{\text {os }}$-SPL2 (orange), which gives now improved results (MAE less than $0.4 \mathrm{kcal} / \mathrm{mol}$ ) even with the more expensive and original SPL2 method (green). At this point, we should remember that both of the red arrows in Figure 1 practically come at no additional cost to the original MP2 calculations, be it MP2 or $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2. In what follows, we will detail the theoretical foundation for the construction of our methods.
Constructing MP AC. The Møller-Plesset adiabatic connection (MP AC) connects the (noninteracting) Hartree-

Fock (HF) system to the physical system through the following Hamiltonian, ${ }^{32,37-39}$

$$
\begin{equation*}
\hat{H}_{\lambda}=\hat{T}+\hat{V}_{\mathrm{ext}}+\lambda \hat{V}_{e e}+(1-\lambda)(\hat{J}+\hat{K}) \tag{1}
\end{equation*}
$$

with $\lambda \geq 0$ the coupling constant, $\hat{T}$ the kinetic energy, $\hat{V}_{e e}$ the electron repulsion operator, $\hat{V}_{\text {ext }}$ representing the external potential, and $\hat{J}=\hat{J}\left[\rho^{\mathrm{HF}}\right]$ and $\hat{K}=\hat{K}\left[\left\{\phi_{i}^{\mathrm{HF}}\right\}\right]$ being the standard HF Coulomb and exchange operators. The last two are functionals of the HF density, $\rho^{\mathrm{HF}}$, and the occupied orbitals, $\phi_{i}^{\mathrm{HF}}$, and are $\lambda$-independent. The correlation energy (exact energy minus HF energy) can be obtained by applying the Hellman-Feynman theorem to eq $1,{ }^{16,30,32,37-40}$

$$
\begin{equation*}
E_{\mathrm{c}}=\int_{0}^{1} W_{c, \lambda} \mathrm{~d} \lambda \tag{2}
\end{equation*}
$$

where $W_{c, \lambda}$ is the AC integrand,

$$
\begin{equation*}
W_{c, \lambda}=\left\langle\Psi_{\lambda}\right| \hat{V}_{\mathrm{ee}}-\hat{J}-\hat{K}\left|\Psi_{\lambda}\right\rangle-\left\langle\Psi_{0}\right| \hat{V}_{\mathrm{ee}}-\hat{J}-\hat{K}\left|\Psi_{0}\right\rangle \tag{3}
\end{equation*}
$$

with $\Psi_{\lambda}$ being the wave function that minimizes $\hat{H}_{\lambda}$. The small$\lambda$ expansion of $W_{c, \lambda}$ is then the standard MP series,

$$
\begin{equation*}
W_{c, \lambda \rightarrow 0}=\sum_{n=2}^{\infty} n E_{c}^{\mathrm{MP} n} \lambda^{n-1} \tag{4}
\end{equation*}
$$

The large $\lambda$ expansion of the MP AC has the form ${ }^{32,38}$

$$
\begin{equation*}
W_{c, \lambda \rightarrow \infty}=W_{c, \infty}+\frac{W_{1 / 2}}{\sqrt{\lambda}}+\frac{W_{3 / 4}}{\lambda^{3 / 4}}+\ldots \tag{5}
\end{equation*}
$$

with

$$
\begin{equation*}
W_{c, \infty}=E_{e l}\left[\rho^{\mathrm{HF}}\right]+E_{x} \tag{6}
\end{equation*}
$$

where $E_{e l}[\rho]$ is ${ }^{32,38}$ the classical electrostatic energy of $N$ point charges bound by a charge density $\rho(\mathbf{r})$ of opposite sign, with $\int \rho(\mathbf{r}) \mathrm{d} \mathbf{r}=N$. While gradient expansion approximations have recently been introduced for both $E_{e l}$ and $W_{1 / 2}{ }^{39}$ here we will approximate $W_{c, \infty}$ from the following inequality,

$$
\begin{equation*}
W_{c, \infty} \leq W_{\infty}^{\mathrm{DFT}}+E_{x} \tag{7}
\end{equation*}
$$

This inequality links $W_{c, \infty}$ to $W_{\infty}^{\mathrm{DFT}}$ from the strong coupling limit of the DFT density-fixed AC. ${ }^{41-46}$ In our previous work, ${ }^{30}$ we introduced $W_{c, \infty}^{\alpha, \beta}$ based on this inequality as

$$
\begin{equation*}
W_{c, \infty}^{\alpha, \beta}=\alpha W_{\infty}^{\mathrm{DFT}}+\beta E_{x} \tag{8}
\end{equation*}
$$

so that $\alpha$ and $\beta$ can be used as fitting parameters. This has been done to approximate the effect of $W_{1 / 2}$ on the MP AC without including it directly. For $W_{\infty}^{\mathrm{DFT}}$ in eq 8 , we use the accurate PC model approximation, ${ }^{41,47,48}$

$$
\begin{equation*}
W_{\infty}^{\mathrm{DFT}}\left[\rho^{\mathrm{HF}}\right] \approx \int\left[A \rho^{\mathrm{HF}}(\mathbf{r})^{4 / 3}+B \frac{\left|\nabla \rho^{\mathrm{HF}}(\mathbf{r})\right|^{2}}{\rho^{\mathrm{HF}}(\mathbf{r})^{4 / 3}}\right] \mathrm{d} \mathbf{r} \tag{9}
\end{equation*}
$$

where $A=-1.451, B=5.317 \times 10^{-3}$. In our previous work, two new MP AC functionals, ${ }^{30}$ SPL2 and MPACF-1, have been introduced. The SPL2 model for $W_{c, \lambda}$ reads

$$
\begin{equation*}
W_{c, \lambda}^{\mathrm{SPL} 2}=C_{1}-\frac{m_{1}}{\sqrt{1+b_{1} \lambda}}-\frac{m_{2}}{\sqrt{1+b_{2} \lambda}} \tag{10}
\end{equation*}
$$

with $b_{1}, m_{1}$, and $C_{1}$ fixed by the exact constraints: (i) $W_{c, \lambda}^{\mathrm{SPL} 2}$ vanishes at 0 ; (ii) its derivative at $\lambda=0$ is equal to $2 E_{c}^{\mathrm{MP2}}$ (eq 4); and (iii) it converges to $W_{c, \infty}$ at $\lambda \rightarrow \infty$ (eq 5). The
remaining parameters are fitted to the S22 set of NCIs. ${ }^{49}$ MPACF1 (hereinafter F1) approximates $E_{c}$ directly, resulting in

$$
\begin{equation*}
E_{\mathrm{c}, \lambda}^{\mathrm{F} 1}=-g \lambda+\frac{g(h+1) \lambda}{\sqrt{d_{1}^{2} \lambda+1}+h^{4} \sqrt{d_{2}^{4} \lambda+1}} \tag{11}
\end{equation*}
$$

with $g$ and $h$ given in ref 30 and $E_{\mathrm{c}, \lambda=1}=E_{c}$. In the original F1 formulation, $\alpha$ and $\beta$ were set to 1 to reproduce the HEG correctly. However, here we use the general F 1 -form where $\left[d_{1}, d_{2}, \alpha, \beta\right]$ were fitted to the S22 data set. The advantage of F1 over SPL2 is that the former has the $\lambda^{-3 / 4}$ term in model $W_{c, \lambda \rightarrow \infty}$, which is missed by SPL2. Although the SPL2 and F1 functionals are not size-consistent by themselves, we recover size-consistency at no extra computational by applying the correction of ref 27.

Different MP2 Flavors Yield Different MP ACs. Here we consider the following general form for spin-scaled and regularized MP2 ${ }^{15}$ correlation energy:

$$
\begin{equation*}
E_{c}=c_{\mathrm{os}} E_{c}^{K_{\mathrm{os}}-\mathrm{MP} 2}+c_{\mathrm{ss}} E_{c}^{K_{\mathrm{ss}}-\mathrm{MP} 2} \tag{12}
\end{equation*}
$$

with the opposite-spin (os) and the same-spin (ss) parts given by (closed-shell systems are assumed throughout this work)

$$
\begin{equation*}
E_{c}^{K_{o s}-\mathrm{MP} 2}=-\sum_{a b r s} \frac{\langle a b \mid r s\rangle^{2}}{\Delta_{a b}^{r s}}\left(1-e^{-\kappa_{\mathrm{os}} \Delta_{a b}^{\text {sb }}}\right)^{2} \tag{13}
\end{equation*}
$$

and

$$
\begin{equation*}
E_{c}^{\kappa_{\mathrm{ss}}-\mathrm{MP} 2}=-\sum_{a b r s} \frac{\langle a b \mid r s\rangle[\langle a b \mid r s\rangle-\langle r s \mid b a\rangle]}{\Delta_{a b}^{r s}}\left(1-e^{-\kappa_{\mathrm{ss}} \Delta_{a b}^{r s}}\right)^{2} \tag{14}
\end{equation*}
$$

with $\Delta_{a b}^{r s}=\epsilon_{r}+\epsilon_{s}-\epsilon_{a}-\epsilon_{b}$ being an orbital gap between a pair of occupied ( $a$ and $b$ ) and a pair of virtual orbitals ( $r$ and $s$ ).
When the $\kappa$ parameters are set to $\infty$, eq 12 reduces to the SCS-MP2 scheme, ${ }^{17}$ with the two parameters $c_{\text {os }}$ and $c_{\text {ss }}$ (when these are further set to 1 , eq 12 reduces to the standard MP2). The os part formally scales as $N^{4}$ whereas the ss part as $N^{5}$, where we refer to ref 18 for further clarifications and details on the scalings of SOS-MP2 vs SCS-MP2. Setting $c_{\mathrm{ss}}=0$ defines the cheaper scaled opposite-spin-only (SOS) variant (with, e.g., $\left.c_{\text {os }}=1.3\right) .{ }^{18}$ When the $\kappa$ parameters are set to a finite value, the divergence for vanishing gap is prevented. In general, similar regularization techniques are also employed in CASPT2 ${ }^{50}$ and Green's function methods. ${ }^{51,52}$ For NCI, $\kappa$ MP2 provides significant improvements over MP2, particularly for values $\kappa \sim 1.2{ }^{15,53,54}$

In the general expression of eq 12, we initially have four parameters that can be fitted. However, here we consider an OS variant with two parameters that require fitting, namely, $c_{\text {os }}$ and $\kappa_{\mathrm{os}}$, while $c_{\mathrm{ss}}$ is preset to 0 . Our second MP2 variant has both ss and os spin channels but still two parameters to be fitted ( $\kappa_{\mathrm{ss}}$ and $\kappa_{\mathrm{os}}$ ) with $c_{\mathrm{ss}}$ and $c_{\text {os }}$ preset to 1 . To each of those, we can add the MP AC correction on top, yielding a large number of different functionals for the correlation energy. Here we focus on the three that are most promising (Table 1), and others can be found in Table S1 of the SI. For naming MP AC methods, we use the following notation: The string in front of the functional indicates which version of MP2 is used in that functional. So if regular MP2 is used within MP AC, the functional will just be called SPL2 or F1. If a modified version of MP2 has been used, such as $c_{\text {os }}, \kappa_{\mathrm{os}}-$ MP2, then the functional is called $c_{\mathrm{os}}, \kappa_{\mathrm{os}}$-SPL2 or $c_{\mathrm{os}}, \kappa_{\mathrm{os}}$-F1. Finally, $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}$-SPL2 means

Table 1. The Six Functionals that We Have Studied in This Work (in Addition to MP2 and SPL2) and the Parameters that Were Fitted to the S22 Data Set ${ }^{a}$

that SPL2 interpolation has been used with the $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}$ - MP2 correlation energy (eq 12 with $c_{\mathrm{ss}}=c_{\mathrm{os}}=1$ ).

We acknowledge that employing a modified MP2 correlation energy as half of the initial slope of $W_{c, \lambda}$ deviates from the exact MP AC theory, which dictates the initial slope to be the unaltered MP2. Nevertheless, we adopt this pragmatic approach, recognizing the inherent flexibility of our MP AC construction, reflected in the end point of our MP AC curve, $W_{c, \infty}^{\alpha, \beta}$, which can adjust itself to the modified initial slope through the $\alpha$ and $\beta$ parameters. Moreover, for larger NCI complexes, the radius of convergence of the MP series could be exceptionally small. ${ }^{16,55}$ This could reduce the utility of the actual slope of the MP AC for approximating the true MP AC. Therefore, our approach, which involves regularization and scaling modifications of the initial MP AC slope, offers us a heuristic to capture MP ACs pertaining to NCIs. Simultaneously, this method provides an opportunity to decrease the computational cost of our MP AC construction, for instance, by using only the os part of MP2.

Returning to Table 1, we distinguish methods with computational scaling of $N^{5}$ (upper part of the table) from those with $N^{4}$ (lower part). We will compare the accuracy of these functionals with other common functionals B2PLYPD3 ${ }^{10,56-61}$ (similar cost to our $N^{5}$ methods) and B3LYP-D3. Note that B3LYP-D3 has the same scaling as our $N^{4}$ methods but likely has a smaller prefactor. Nevertheless, SOS-MP2 can be implemented very efficiently ${ }^{10,57-62}$ ), with its wall times often being shorter than those for the HF calculation. ${ }^{62}$

Training of Our Functionals on S22. Similar to previous work, ${ }^{30}$ we train the empirical parameters in our functional by minimizing their MAEs for the S22 data set. ${ }^{49}$ Notice that for each $\kappa$ value, the MP2 correlation energy needs to be recalculated, and we consider here $\kappa$-values between 0.6 and 1.7 with a 0.1 step. We also use $c_{\text {os }}$ values from 1.0 to 2.5 with a step of 0.1 . Note that the reoptimization of the MP AC parameters is needed any time an alteration to MP2 is made. Thus, for each combination of $\kappa_{\mathrm{os}}$ and $\kappa_{\mathrm{ss}}$ (or $c_{\mathrm{os}}$ ), the corresponding parameters of the functionals are given in Table S2 in the SI.

In Figure 2, we show contour plots with the MAE for the S22 data set with the parameters of our selected methods (functionals). The two left-side panels display the S22 MAE contour plots for two distinct MP2 versions ( $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}-\mathrm{MP} 2$, an $N^{5}$ method; and $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 an $N^{4}$ method) against their respective parameters. Conversely, the two right panels show the same plots after applying the SPL2 correction.


Figure 2. Top: The error between the interacting energy for the S22 data set between the CCSD (T) reference data and $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}-\mathrm{MP} 2$ and $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}-\mathrm{SPL} 2$ for a range of $\kappa$ 's between 0.6 and 1.7. Bottom: The same is true except for $c_{\text {os }} \kappa_{\mathrm{os}}-\mathrm{MP} 2$ and $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 for a range of $c_{\text {os }}$ values from 1 to 2.5 and $\kappa_{\text {os }}$ 's between 0.6 and 1.7. The black dots denote the minimum energy of the contour plot with its corresponding value. A figure containing the curves of $\kappa$-MP2 and $\kappa$-SPL2, so the diagonals of the two top panels, can be found in the SI (figure S2).

Table 2. Total MAE's and Maximum Errors of the 5 NCI Data Sets Considered in This Work for the 8 Studied Functionals ${ }^{a}$

| Method | S22 |  | NCCE31 |  | S66x8 |  | X40 |  | B30 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | MAE | MAX | MAE | MAX | MAE | MAX | MAE | MAX | MAE | MAX |
| MP2 | 0.86 | 3.47 | 0.5 | 2.53 | 0.39 | 4.15 | 0.34 | 2.39 | 0.86 | 3.73 |
| SPL2 | 0.15 | 0.55 | 0.25 | 2.16 | 0.15 | 1.12 | 0.16 | 0.45 | 0.48 | 2.51 |
| $\kappa$-F1 | 0.14 | 0.54 | 0.32 | 2.66 | 0.13 | 1.00 | 0.17 | 0.45 | 0.7 | 2.01 |
| B2PLYP-D3 | $0.15{ }^{61}$ | $0.71{ }^{61}$ | $0.61{ }^{63}$ | $2.41{ }^{63}$ | $\mathbf{0 . 1 1}{ }^{64}$ | $0.90^{64}$ | 0.18 | 0.69 | 0.72 | 3.49 |
| $c_{\text {os }} \kappa_{\text {os }}$-MP2 | 0.19 | 0.89 | 0.4 | 2.92 | 0.2 | 1.17 | 0.22 | 0.74 | 0.93 | 3.96 |
| $c_{\text {os }} \kappa_{\text {os }}$-SPL2 | 0.11 | 0.36 | 0.27 | 2.23 | 0.11 | 0.73 | 0.10 | 0.39 | 0.45 | 1.27 |
| $c_{\text {os }}$-SPL2 | 0.15 | 0.61 | 0.21 | 1.84 | 0.13 | 0.88 | 0.13 | 0.39 | 0.9 | 2.21 |
| B3LYP-D3 | $0.31{ }^{61}$ | $1.16^{61}$ | $0.60{ }^{63}$ | $3.18{ }^{63}$ | $0.15{ }^{64}$ | $1.05{ }^{64}$ | $0.22{ }^{65}$ | $1.21{ }^{65}$ | $1.36{ }^{65}$ | $5.60{ }^{65}$ |

${ }^{a}$ The best result for each column is highlighted in boldface. The S22 results for dRPA@PBE, RPA-SOSEX@PBE, and MP2C can be found in Table S3 from the SI. The NCCE31 results of dRPA@PBE can also be found in Table S4 of the SI. The box-plots of S22, NCCE31, B30, and X40 for MP2, B2PLYP-D3, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2, and B3LYP-D3 can be found in Figure S3, whereas the box-plots of S66x8 for the same functionals can be found in Figure S 4 .

Note that the SPL2 contour plots on the right are the result of reoptimizing the four SPL2 parameters $\left(b_{2}, m_{2}, \alpha, \beta\right)$ for each combination of the $\kappa_{\mathrm{ss}}$ and $\kappa_{\mathrm{os}}$ parameters (upper panel) and $c_{\mathrm{os}}$ and $\kappa_{\mathrm{os}}$ parameters (lower panel). The transition from the uncorrected MP2 versions (left panels) to the SPL2 corrected versions (right panels) yields a narrower range of MAEs, echoing our previous discussion on the adaptability of our MP AC construction to different MP2 inputs. The robustness of the SPL2 correction, regardless of the MP2 input, is further evidenced by the significant areas within both SPL2 contour plots that exhibit minimal MAE variation.

The top-left panel focuses on $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}-\mathrm{MP} 2$ with regularizers $\kappa_{\mathrm{ss}}$ and $\kappa_{\text {os }}$ as parameters. Here, the MAE is considerably more sensitive to the $\kappa_{\text {os }}$ regularizer than to its same-spin counterpart. The minimum MAE is achieved at $\kappa_{\mathrm{os}}=1.4$ and
$\kappa_{\mathrm{ss}}=0.9(0.16 \mathrm{kcal} / \mathrm{mol})$. If we restrict $\kappa=\kappa_{\mathrm{ss}}=\kappa_{\mathrm{os}}$ as done in previous work, ${ }^{15}$ the minimum is at $\kappa=1.1$ with an MAE of $0.21 \mathrm{kcal} / \mathrm{mol}$ (see Figure S2 in the SI). Adding the SPL2 correction to $\kappa_{\text {ss }} \kappa_{\mathrm{os}}$-MP2 (moving from the top-left to top-right panel) shifts the minimum to $\kappa_{\text {os }}=1.1$ and $\kappa_{\mathrm{ss}}=1.7$, reducing it to $0.11 \mathrm{kcal} / \mathrm{mol}$. While the reduction is modest, it is likely approaching the accuracy limit of MP2-like methods for the S22 data set. Importantly, we observe a relatively flat area around the minimum, suggesting that the MAE is almost constant as long as $\kappa_{\mathrm{ss}}$ and $\kappa_{\mathrm{os}}$ exceed approximately 1.1.

The lower-left panel illustrates the MAE for the opposite-spin-only $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 with its parameters. Now the MAE depends substantially more on the $c_{\text {os }}$ scaling factor than on the $\kappa_{\mathrm{os}}$ regularizer. The minimum is achieved at $c_{\mathrm{os}}=2.1$ and $\kappa_{\mathrm{os}}=$ 0.9 , resulting in an MAE of $0.19 \mathrm{kcal} / \mathrm{mol}$. This value is only
marginally higher than the one from the upper panel, despite a significant reduction in computational cost. With the addition of the SPL2 correction to $c_{\mathrm{os}} \kappa_{\mathrm{os}}-\mathrm{MP} 2$ (transition from the lower-left to lower-right panel), the minimum MAE dips to $0.11 \mathrm{kcal} / \mathrm{mol}$, slightly shifted to $c_{\text {os }}=2.1$ and $\kappa_{\text {os }}=1.3$. This MAE yields a significant improvement over $c_{o s} \kappa_{\text {os }}$-MP2, and is on par with the more expensive $\kappa_{\mathrm{ss}} \kappa_{\mathrm{os}}$-SPL2 (top-right panel). Thus, through these explorations, we see the robustness of SPL2 in correcting various MP2 inputs.
$c_{o s} K_{o s}$-MP2 as Our Workhorse. After training our functionals on S22, we test them on a variety of NCI data sets, with the results shown in Table 2. Alongside S22, we include NCCE31, ${ }^{66-68}$ S66x8, X40, and B30. From Table 2, we can see that $c_{\text {os }} \kappa_{\text {os }}$-MP2 has the overall best performance (even when compared with more expensive $N^{5}$ methods).

In Figure 3, we use radar plots to show the performance of various functionals across five interaction-specific subsets in


Figure 3. A radar plot for NCCE31 containing the MAEs in $\mathrm{kcal} / \mathrm{mol}$ of the 5 subgroups (HB6, CT7, DI6, WI7, and PPS5) as well as the total MAE of MP2, $c_{\text {os }} \kappa_{\text {os }}$-SPL2, B3LYP-D3, and B2PLYP-D3. A table containing the MAEs of these methods as well as the other 4 methods studied in the paper can be found in the SI (Table S3). HB stands for hydrogen bonds, CT for charge transfer interaction, DI for dipole interactions, WI for weak interactions, and PPS for $\pi-\pi$-stacking interactions.
the NCCE31 data set, including hydrogen bonds (HB6), charge-transfer interactions (CT7), dipole interactions (DI6),
weak interactions (WI7), and $\pi-\pi$-stacking interactions (PPS5). We can see that $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 exhibits superior performance in CT7, DI6, and PPS5 subsets, as well as the NCCE31 data set as a whole. While B2PLYP-D3 outperforms $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 for the WI7 subset, the differences in the MAEs are very small. In the HB6 subset, MP2 is the top performer, but the accuracy of $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 is still satisfactory and is similar to that of B2PLYP-D3.

S66x8-Accuracy beyond Equilibrium. Turning our attention to the $\mathrm{S} 66 \times 8$ data set, ${ }^{6,70,71}$ we assess the performance of our functionals across NCIs beyond their geometric equilibrium (Table 3). We can see from the table that $c_{\text {os }} \kappa_{\text {os }}$-SPL2 ( $N^{4}$ scaling) and B2PLYP-D3 ( $N^{5}$ scaling) deliver twice the accuracy of MP2. For dispersion and mixed interactions, B3LYP-D3 shows a slightly better performance than our $c_{\text {os }} \kappa_{\text {os }}{ }^{-}$ SPL2, but for the hydrogen-bonded complexes, B3LYP-D3 yields the poorest results.

From the box-plots shown in Figure 4, we observe the impact of SPL2 addition on the error distributions of MP2 and $c_{\text {os }} \kappa_{\text {os }}$-MP2 across individual categories within the S66x8 data set.

As expected, ${ }^{30}$ SPL2 considerably improves "bare" MP2 across all categories. Using SPL2 to correct $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 can be considered more challenging, as $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 already gives significant improvements over MP2. Nevertheless, SPL2 also does a very good job here, as $c_{\text {os }} \kappa_{\text {os }}$-SPL2 surpasses $c_{\text {os }} \kappa_{\text {os }}$-MP2, both in terms of median values and the range of errors across all S66x8 categories. Moreover, as demonstrated in Figure 4, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 (orange) significantly outperforms the original SPL2 (green), highlighting the dual advantage of our newly developed SPL2 formulation; it provides superior accuracy at a reduced computational cost.

Continuing with the S 66 x 8 data set, Figure 5 shows dissociation curves for four S66 complexes: AcOH-Uracil (representing hydrogen bonds), Ethyne-Pyridine (mixed interactions), and Benzene-Ethene I and Pentane-Ethyne (both representing dispersion interactions). The dissociation curves of $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 align closely with those of the $\operatorname{CCSD}(\mathrm{T})$ reference. While B2PLYP-D3 also provides accurate curves, it is outperformed by $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 in the case of the mixed Ethyne-Pyridine complex. The $c_{o s} \kappa_{o s}$-SPL2 method effectively corrects the MP2 curves, both when MP2 underbinds (AcOH-Uracil), and when it overbinds (remaining complexes). In addition to errors in interaction energies, MP2 also displays significant errors in the equilibrium distances between dispersion-bonded fragments ${ }^{72,73}$ (position of the minimum in

Table 3. Total MAEs and Maximum Errors of the S66x8 Data Set of 8 Functionals as well as the MAEs for Only the Hydrogen Bonding Complexes, the Dispersion Interaction Dominated Complexes, and the Mixed Set ${ }^{a} \mathrm{I}$

| Method | Total |  | H-bonds |  | Dispersion |  | Mixed |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | MAE | MAX | MAE | MAX | MAE | MAX | MAE | MAX |
| MP2 | 0.39 | 4.15 | 0.14 | 0.71 | 0.67 | 4.15 | 0.34 | 1.59 |
| SPL2 | 0.15 | 1.12 | 0.12 | 0.42 | 0.21 | 1.12 | 0.1 | 0.75 |
| $\kappa$-F1 | 0.13 | 1.00 | 0.09 | 0.44 | 0.19 | 1.00 | 0.11 | 0.60 |
| B2PLYP-D3 | 0.11 | 0.90 | 0.09 | 0.40 | 0.14 | 0.90 | 0.08 | 0.52 |
| $c_{\text {os }} \kappa_{\text {os }}$-MP2 | 0.2 | 1.17 | 0.14 | 1.00 | 0.32 | 1.17 | 0.14 | 0.64 |
| $c_{\text {os }} \kappa_{\text {os }}$-SPL2 | 0.11 | 0.73 | 0.08 | 0.41 | 0.16 | 0.73 | 0.1 | 0.48 |
| $c_{\text {os }}$ SPL2 | 0.13 | 0.88 | 0.11 | 0.58 | 0.16 | 0.88 | 0.1 | 0.53 |
| B3LYP-D3 | 0.15 | 1.05 | 0.22 | 1.05 | 0.13 | 0.65 | 0.08 | 0.50 |

[^1]

Figure 4. Box-plots including outliers as dots containing the errors of MP2, SPL2, and $c_{\mathrm{os}} \kappa_{\mathrm{os}}-\mathrm{MP} 2$ and $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-MP2 for respectively the dispersion interactions (top left), hydrogen bonds (top right), mixed interactions (bottom left), and all interactions in total (bottom right), which show that the SPL2 form corrects the base MP2 form for the different kind of interactions of S66x8. Medians are denoted by a thin white line inside of each box, and the underlying median values are also given above each box. The box-plots of the remaining functionals can be found in Figure S5 in the SI. Box-plots showing how errors of different methods vary with the distances between fragments can be found in Figure S6 in the SI.


Figure 5. Dissociation curves of a dimer containing hydrogen bonds (top left) and containing mixed interactions (top right) and two dimers containing dispersion interactions (bottom) for conventional wave function methods ( $\mathrm{CCSD}(\mathrm{T}$ ) and MP2), dispersion corrected DFT functionals (B3LYP-D3 and B2PLYP-D3), and our new $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 functional. The dissociation curves of all of the functionals can be found in the SI (Figure S7).
the dissociation curves), and this is also fixed by $c_{\text {os }} \kappa_{\text {os }}$-SPL2 (lower panels of Figure 5).
NCIs in Complexes Involving Halogens, Chalcogens, and Pnictogens. We extend our analysis to the NCI between
molecules with halogen, chalcogen, and pnictogen atoms. Simulations of halogen bonds pose significant challenges to dispersion-corrected DFT, with difficulties often ascribed to the delocalization error inherent to density-functional approx-

Table 4. Total MAEs and Maximum Errors of 31 Complexes from the X40 Data Set of 9 Functionals Split as well as the MAEs of the Separate Hydrogen Bonding, Halogen Bonding, and the Remaining NCIs ${ }^{a}$

| Method | Total |  | H-bonds |  | HGB |  | Rest |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | MAE | MAX | MAE | MAX | MAE | MAX | MAE | MAX |
| MP2 | 0.34 | 2.39 | 0.26 | 1.08 | 0.27 | 0.87 | 0.48 | 2.39 |
| SPL2 | 0.16 | 0.45 | 0.17 | 0.31 | 0.2 | 0.45 | 0.1 | 0.18 |
| $\kappa$-F1 | 0.17 | 0.45 | 0.26 | 0.45 | 0.16 | 0.37 | 0.1 | 0.18 |
| B2PLYP-D3 | 0.18 | 0.69 | 0.25 | 0.69 | 0.13 | 0.18 | 0.16 | 0.43 |
| $c_{\text {os }} \kappa_{\text {os }}$-MP2 | 0.22 | 0.74 | 0.24 | 0.74 | 0.23 | 0.53 | 0.2 | 0.31 |
| $c_{\text {os }} \kappa_{\text {os }}$-SPL2 | 0.1 | 0.39 | 0.1 | 0.21 | 0.12 | 0.27 | 0.09 | 0.39 |
| $c_{\text {os }}$-SPL2 | 0.12 | 0.39 | 0.15 | 0.38 | 0.11 | 0.39 | 0.1 | 0.25 |
| B3LYP-D3 | 0.22 | 1.21 | 0.46 | 1.21 | 0.1 | 0.20 | 0.13 | 0.31 |
| M06-2X | 0.27 | 0.90 | 0.24 | 0.40 | 0.40 | 0.90 | 0.17 | 0.35 |

${ }^{a}$ As in ref 65, we have removed the iodine complexes as they require relativistic corrections. This leaves the set with 31 different halogen bonded complexes, which have been also studied in the context of double hybrids ${ }^{74,75}$ and the $\kappa$-regularizers. ${ }^{15,20}$ The best result for each column is highlighted in boldface.
imations, ${ }^{26,76-79}$ which cannot be fixed by empirical dispersion corrections such as D3 or D4. ${ }^{26,76-79}$ However, accurately capturing halogen bonds is crucial, given their significant role in numerous biomolecular (protein-ligand) complexes and diverse crystalline materials. ${ }^{35}$ We test our methods against the X40 data set from Řezáč et al., ${ }^{35}$ which includes a broad range of NCIs, spanning from hydrogen to pure halogen bonds in molecules containing halogens. Table 4 provides a summary of the results for the X40 data set, including the MAE for the overall X40 and for the hydrogen-bonded, halogen-bonded, and remaining complexes in the set. Once again, $c_{\mathrm{os}} \kappa_{\mathrm{os}}-\mathrm{SPL} 2$ is the most accurate method, outperforming MP2 by factors ranging from 2 to 5 depending on the X40 category. In comparison with B3LYP-D3, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 exhibits comparable accuracy for halogen bonds and the "remaining" complexes, but surpasses B3LYP-D3 by a factor of 4.6 for hydrogen bonds. We have added M06-2 $\mathrm{X}^{80}$ to Table 4, a functional that was assessed reliable for halogen-bonded complexes. ${ }^{77,81}$ Our results demonstrate that $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 significantly outperforms M06-2X, both for the entire X40 data set and for its subsets.

Before delving into more intricate halogen-bonded complexes, we note that $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 displays remarkably low errors for both S22 and X40. The two sets include NCIs of very different natures, and yet, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 attains MAEs of 0.11 and $0.1 \mathrm{kcal} / \mathrm{mol}$, respectively (approaching the uncertainties in the $\operatorname{CCSD}(\mathrm{T})$ reference itself $\left.{ }^{82}\right)$. This level of accuracy, unequaled among methods of similar or even higher computational cost, is highlighted in Figure 6. From this figure we can see that while $\kappa$-MP2.5 $5^{20}$ comes close to the performance of $c_{\mathrm{os}} \kappa_{\mathrm{os}}{ }^{-}$ SPL2, it also comes at a higher computational cost $\left(N^{6}\right.$ scaling).

The B30 data set, comprising halogen, chalcogen, and pnictogen bonds, ${ }^{34}$ has proven to be an even more difficult challenge for DFT than X40, ${ }^{65,74}$ being classified as a "difficult NCI" data set. In Table 5, the MAE of the full B30 set and the three separate subsets are displayed, and we can see that $c_{\text {os }} \kappa_{\text {os }}$ SPL2 outperforms other methods for halogen bonds (aHGB), with an MAE of $0.73 \mathrm{kcal} / \mathrm{mol}$, significantly surpassing all competing methods. In the case of chalcogen bonds, however, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 is surpassed by SPL2, and by MP2 for pnictogen bonds. Nevertheless, for the full B30 set, $c_{\mathrm{os}} \kappa_{\mathrm{os}}-\mathrm{SPL} 2$ is still a top performer.
The performance of $c_{0 S} K_{05}$-SPL2 and the discrepancy in the accuracy of other methods for six B30 and eleven X40 halogen bonds are showcased in Figure 7. While the electron-donor in


Figure 6. Correlation plot for the X40 and S22 MAE for methods with $N^{4}, N^{5}$, and $N^{5+}$ scalings. The results for all the hybrid functionals were taken from ref 65, the results for B2PLYP-D3 for S22 were taken from ref 61 and for x40 were calculated using Turbomole (see Computational Details for more information), and the results for MP2.5, кMP2.5, and MP3 were taken from ref 20.

X40 is always a neutral species, in B30 it also can be an atomic anion. From Figure 7, we observe that both DFT-D3 and MP2 methods exhibit substantial errors for the more "extreme" B30 halogen-bond interactions. Yet, these methods demonstrate excellent performance for the remaining complexes, barring the MP2's performance for the last two, dominated by halogen $-\pi$ interactions. For the B30 complexes in Figure 7, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 is by far the most accurate, and its accuracy does not deteriorate once we move to the X40 complexes, suggesting that unlike other methods, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 can accurately predict halogen interactions of different nature. The good performance of DFT-D3 for X40 halogen bonds and very bad for B30 is a consequence of large density-driven errors associated with the anions and charge-transfer prevalent in B30 halogen-bonded complexes. These errors have been extensively dissected in several studies. ${ }^{26,76,83,84}$ Different RPA functionals with DFT orbitals struggle for the same reasons as the standard DFT functionals, with the density corrected variant of RPA giving improvements, albeit still performing worse (see Table S6 from the SI). ${ }^{85}$

Furthermore, in Figure 8, we show the dissociation curves of $\mathrm{NH}_{3}-\mathrm{FBr}$ (complex 6 in Figure 7). Here, $c_{\mathrm{os}} \kappa_{\mathrm{os}}-\mathrm{SPL} 2$ again outperforms the other functionals and closely matches the CCSD $(T)$ reference. Even though B3LYP-D3 and B2PLYP-D3

Table 5. Total MAEs and Maximum Errors of the B30/Bauza30 Data Set of 9 Functionals Split as well as the MAEs of the Separate Halogen Bonding (Containing Mostly Anions as Donors), Chalcogen Bonding, and Pnictogen Bonding Complexes ${ }^{a}$

| Method | Total |  | aHGB |  | CHB |  | PNB |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | MAE | MAX | MAE | MAX | MAE | MAX | MAE | MAX |
| MP2 | 0.86 | 3.73 | 2.22 | 3.73 | 0.63 | 1.21 | 0.17 | 0.22 |
| SPL2 | 0.48 | 2.51 | 1.23 | 2.51 | 0.2 | 0.66 | 0.6 | 1.00 |
| $\kappa$-F1 | 0.84 | 2.01 | 2.27 | 2.01 | 0.58 | 1.25 | 0.2 | 0.77 |
| B2PLYP-D3 | 0.72 | 3.49 | 2.02 | 3.49 | 0.23 | 0.83 | 0.89 | 1.78 |
| $c_{\text {os }} \kappa_{\text {os }}$-MP2 | 0.93 | 3.96 | 2.6 | 3.96 | 0.59 | 1.34 | 0.26 | 0.58 |
| $c_{\text {os }} \kappa_{\text {os }}$-SPL2 | 0.45 | 1.27 | 0.73 | 1.27 | 0.4 | 0.92 | 0.31 | 0.61 |
| $c_{\text {os }}$-SPL2 | 0.9 | 2.21 | 1.53 | 2.21 | 0.64 | 1.95 | 1.03 | 1.68 |
| B3LYP-D3 | 1.36 | 5.60 | 3.50 | 5.60 | 0.98 | 2.10 | 0.36 | 0.91 |
| M06-2X | 0.99 | 3.22 | 1.80 | 3.22 | 0.59 | 1.80 | 1.36 | 1.71 |

${ }^{a}$ The best result for each column is highlighted in boldface. The MAEs and MAXs for 3 different RPA variants can be found in Table S6 in the SI.


Figure 7. Errors between the $\operatorname{CCSD}(\mathrm{T})$ reference data ${ }^{35,77}$ and MP2, $c_{\mathrm{os}}, \kappa_{\mathrm{os}}-\mathrm{SPL} 2$, B3LYP-D3, and B2PLYP-D3, respectively, for the halogen bonded complexes of the $\mathrm{B} 30^{34,77}$ and $\mathrm{X} 40^{35}$ data sets. The underlying complexes are shown on the right, with the first four containing negatively charged halogen atoms. The curves for $c_{o s} \kappa_{o s}-\mathrm{MP} 2$ and SPL2 can be found in the SI (left panel of Figure S8).


Figure 8. Dissociation curve of $\mathrm{NH}_{3}-\mathrm{FBr}$ (complex 6) for $\operatorname{CCSD}(\mathrm{T})$, MP2, $c_{\text {os }} \kappa_{\text {os }}$-SPL2, B3LYP-D3[BJ], and B2PLYP-D3. Another dissociation curve of the B30 set can be found in the SI (right panel of Figure S8).
at small bond distances and MP2 at larger ones display good accuracy, $c_{\text {os }} \kappa_{\text {os }}$-SPL2 is the only method accurate around equilibrium. Similar conclusions can be drawn from the dissociation curve of another anionic halogen-bonded complex (see the right panel of Figure S8 in the SI).
In summary, $c_{\text {os }} \kappa_{\text {os }}$-SPL2 offers excellent performance for NCIs of various types. Building upon its SPL2 predecessor, $c_{\text {os }} k_{\text {os }}$-SPL2 offers substantial improvements over both MP2 and dispersion-corrected (double) hybrid methods. Simultaneously, $c_{\mathrm{os}} \kappa_{\mathrm{os}}$-SPL2 achieves even greater accuracy than SPL2,
while costing less ( $N^{4}$ vs $N^{5}$ scaling). Further, we demonstrate the robustness and versatility of the SPL2 correction, showing its capability to significantly correct not only "bare" MP2, for which it was initially designed, but also various MP2 modifications. This flexibility is maintained while retaining, or even improving upon, the original SPL2 accuracy.

While we used a large aug-cc-pVQZ basis set in this study, ${ }^{86}$ encouraged by SPL2's ability to absorb errors of both MP2 and its alterations, in the spirit of refs 87 and 88, we will also apply SPL2 to correct MP2 within smaller basis sets (e.g., that from ref 89). Finally, we will also use machine learning to create SPL2-like interpolations of the MP AC curve between its small and large $\lambda$ limits. This approach could offer an even higher accuracy for NCIs.

## COMPUTATIONAL DETAILS

For all calculations, PYSCF $^{90}$ is used to calculate the exchange energy, the HF density, and the MP2 correlation energy, which have been used for MP2 and SPL2 I\& F1 forms. To calculate the $\kappa$-MP2 correlation energies, we used a python, PYSCF-like code with numba parallelization. This code and the full code for running MP AC calculations can be found in ref 91. Unless otherwise specified, we have used the aug-cc-pVQZ. ${ }^{92}$ The exceptions are the bromide complexes of $\mathrm{X} 40^{35}$ for which we run Br with a aug-cc-pVQZ-pp basis $\operatorname{set}^{93}$ and the corresponding pseudopotential ${ }^{94}$ as done by Mardirossian et al. ${ }^{65}$ A frozen core was also applied in these cases. The iodine complexes were removed from X40 due to the need for relativistic corrections.

The reference data of S22 and S66x8 for B3LYP-D3 and B2PLYP-D3 can be found in refs 61 (def2-QZVP) and 64 (haVQZ), respectively. The NCCE31 data for both functionals are taken from ref 63 (aug-cc-pVTZ) for B3LYP-D3 and B2PLYP-D3. The B3LYP-D3 and M06-2X data for B30 are from refs 65 (aug-cc-pVTZ) and 20 (aug-cc-pVTZ), whereas the data for B2PLYP-D3 have been calculated using TURBOMOLE ${ }^{95,96}$ with an aug-cc-pVQZ basis. For X40 the B3LYP-D3 and M06-2X data are from ref 65. (aug-cc-pVTZ) and the B2PLYP-D3 data have been calculated from the aug-cc-pVTZ basis set in TURBOMOLE. The data from the dissociation curves for MP2 and $\operatorname{CCSD}(\mathrm{T})$ have been also calculated with the same basis set with TURBOMOLE and extrapolated to the CBS limit. The B2PLYP-D3 and B3LYPD3 dissociation curves have been calculated using a standard aug-cc-pVQZ basis set in TURBOMOLE, whereas the $c_{\text {os }} \kappa_{\text {os }}-$ SPL2 has been run within the same basis set in PYSCF. For MP2C we have taken the reference data from ref 97 for S22 (extrapolated to the CBS limit). The dRPA data are obtained from ref 85 for S22 (cc-pVTZ) and B30 (cc-pVQZ) and from ref 98 for NCCE31 (ATZ). For RPA-SOSEX we got the data for S22 (cc-pVTZ) and B30 (cc-pVQZ) also from ref 85, but the S66x8 data (CBS limit) is from ref 69. Lastly the density corrected dRPA data, C(HF)-dRPA, were taken from ref 85. All of the RPA reference data were run using PBE orbitals. The fitted parameters of all the functionals, other than the ones from Table 1, can be found in Table S2 of the SI.

## - ASSOCIATED CONTENT

## Data Availability Statement

The data that support the findings of this study are available within the main manuscript and its Supporting Information. Additional raw data and the code can also be found on Zenodo in ref 99 and ref 91. respectively.

## (s) Supporting Information

The Supporting Information is available free of charge at https://pubs.acs.org/doi/10.1021/acs.jpclett.3c01832.

Additional S66x8 plots, additional comparison data, parameters for the new methods, additional dissociation curves, and results for halogen bonds (PDF)

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## Notes

The authors declare no competing financial interest.

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