Predicting Band Gaps with Hybrid Density Functionals

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We compare the ability of four popular hybrid density functionals (B3LYP, B3PW91, HSE, and PBE0) for predicting band gaps of semiconductors and insulators over a large benchmark set using a consistent methodology. We observe no significant statistical difference in their overall performance although the screened hybrid HSE is more accurate for typical semiconductors. HSE can improve its accuracy for large large band gap materials –without affecting that of semiconductors– by including a larger portion of Hartree–Fock exchange in its short range. Given that screened hybrids are computationally much less expensive than their global counterparts, we conclude that they are a better option for the black box prediction of band gaps.

Introduction. Band structure calculations are an important application of electronic structure methods in materials science. Due to the cost of computing electronic properties for extended solids, density functional theory (DFT) methods are most often used for such calculations. However, local and semilocal density functional approximations—the most affordable type of Kohn–Sham functionals for solids—badly underestimate the band gaps of semiconductors (the materials of principal interest in practical applications) and insulators due to self-interaction error [1, 2]. Hybrid functionals that incorporate a fraction of nonlocal Hartree–Fock (HF) exchange overcome this issue; however, computing HF exchange in solids is considerably more expensive than evaluating a semilocal density functional. Methods based on the GW approximation can also be used to compute band gaps more accurately, but these techniques are even more expensive than hybrids. A good compromise between cost and accuracy is provided by short-range screened hybrids: functionals that include HF exchange only for the short-range part of the electron–electron interaction, which significantly reduces the cost of evaluating the nonlocal HF part of the exchange as compared to standard hybrids [3]. This type of functionals have been shown to provide reasonably accurate band gaps for semiconductors, and variants of the Heyd–Scuseria–Ernzerhof (HSE) short-range hybrid have been widely used for the calculation of semiconductor band gaps for many years [8].

A recent study [9] declares the (standard) hybrid B3PW91 as the winner for resolving the band gap prediction problem for materials design. However, that study did not evaluate the performance of other hybrid functionals using a direct comparison based on a consistent methodology. We here carry out such detailed comparison. We also comment upon the pitfalls that arise when benchmarking density functionals [10, 11], as well as aspects of computational efficiency for solids between short-range hybrids like HSE and standard global hybrids (by global, we mean standard full-range hybrids like B3PW91).

The present study focuses on band gap prediction by both short-range screened and global hybrids. We benchmark popular hybrids like HSE06 (henceforth referred to as simply HSE), B3PW91, B3LYP, and PBE0 (also known as PBEh or PBE1PBE). For this comparison, we utilize a variety of error measures in order to avoid the difficulties and ambiguities that emerge when trying to decide which electronic structure method is best for a determined task [10, 11]. We find that these four functionals are similarly adequate for the calculation of band gaps; different error measures for HSE, B3PW91, and B3LYP are very close to each other and, while PBE0 tends to overestimate band gaps, the error is systematic and straightforward to correct with a linear fit. If the benchmark set is narrowed to traditional semiconductors then HSE emerges as the winner. In addition, we show that short-range hybrids can describe even large band gap insulators by including an increased portion of HF exchange in the short range only. Due to their lower cost as compared to global hybrids, greater ease to achieve convergence, and overall good performance, we conclude that screened hybrids are a better option for the prediction of band gaps.

Statistical Evaluation. We have carried out band gap calculations for 41 semiconductors and insulators using the GAUSSIAN suite of programs. The set includes materials from the SC40 dataset [7], as well as transition metal oxides (FeO, CoO, NiO, MnO, and VO2) and large band gap salts (NaCl, LiCl, and LiF); the set is similar to that of Ref. [9] (some compounds have been excluded because spin-orbit effects are very large, and we have not used spin-orbit corrections in our calculations). As in that reference, we employed experimental geome-
We were thus unable to obtain results for all 41 compounds: if the very large band gap (i.e., more than \( \approx 8 \text{ eV} \)) compounds are excluded, the MAEs for HSE, B3PW91, B3LYP, and PBE0 are 0.26, 0.37, 0.36, and 0.57 eV, respectively. Furthermore, if we consider only typical semiconductors for which low-temperature experimental band gaps are available (group 1 of the SC40 set; see Ref. 7, the MAEs (in the same order) are 0.18, 0.32, 0.31, and 0.55 eV, and the MAPEs are 17, 33, 26, and 62. We also report in Table I the standard deviations (SDs) of the error; a perfect approximation would have zero ME and zero SD. However, no significant difference is observed between the SDs of the hybrids: the largest difference is 0.05 eV between HSE and PBE0. Note also that the functional with the worst ME and MAE, PBE0, has the smallest SD. Thus, as has been pointed out by Savin et al. 10, 11, using different error measures to judge the quality of a functional can lead to contradictory results.

The following three statistics of the deviations from experiment in Table I the median, the median absolute deviation, and the interquartile range, are measures taken from robust statistics, which are more resilient to outliers as compared to the ME, MAE, and SD. In statistics, the median is considered a better descriptor of typical outcome than the mean when the latter is biased by outliers; the median absolute deviation measures statistical dispersion (in this case of the error); and the interquartile range measures the variability in a set of data. For
TABLE I: Different band gap prediction error measures for the four functionals studied here.

| Size of Test Set | Expt. | HSE | B3PW91 | B3LYP | PBE0 |
|------------------|-------|-----|--------|-------|------|
| Mean Error       | 0     | -0.24 | -0.21 | 0.14 | 0.12 | 0.13 | 0.16 | 0.16 | 0.43 |
| Mean Absolute Error | 0 | 0.37 | 0.38 | 0.44 | 0.41 | 0.44 | 0.47 | 0.59 |
| Mean Absolute Percent Error | 0 | 16 | 15 | 27 | 26 | 23 | 25 | 45 |
| Standard Deviation | 0 | 0.55 | 0.54 | 0.53 | 0.51 | 0.56 | 0.56 | 0.50 |
| Median | 0 | -0.06 | -0.07 | 0.33 | 0.29 | 0.19 | 0.27 | 0.63 |
| Median Absolute Deviation | 0 | 0.21 | 0.27 | 0.25 | 0.29 | 0.34 | 0.3 | 0.14 |
| Interquartile Range | 0 | 0.49 | 0.50 | 0.57 | 0.55 | 0.69 | 0.73 | 0.43 |
| Linear Fit Slope | 1 | 0.79 | 0.87 | 0.82 | 0.82 | 0.82 | 0.81 | 0.82 |
| Linear Fit Intercept | 0 | 0.36 | 0.23 | 0.65 | 0.65 | 0.63 | 0.70 | 0.95 |
| Kendall τ Correlation | 1 | 0.85 | 0.88 | 0.85 | 0.88 | 0.84 | 0.85 | 0.86 |

A method that perfectly predicts band gaps, all these values must be zero. We observe, as before, that different functionals provide the best results for different error measures: HSE has the best median and PBE0 the worst, but the latter has the best median absolute deviation and interquartile range. Note, however, that the median absolute deviation and the interquartile range of HSE is only 0.06 eV larger than that of PBE0, while the median of HSE is 0.53 eV closer to zero than the median of PBE0. Thus, overall, measures from robust statistics favor HSE the most.

Apart from the various statistics analyzed so far, an important feature of a quantum chemical method is its ability to correctly reproduce trends. A trend can be defined by a linear fit as follows:

\[ y_{\text{expt}} = my_{\text{calc}} + b. \] (1)

Table I shows the values of the slope \( m \) and the intercept \( b \) from this linear fit for the band gaps calculated by the four hybrid functionals. Ideally, \( m = 1 \) and \( b = 0 \). Judging by the values of the slopes, the four hybrids have a similar quality for predicting trends: the slopes are all around 0.80 and do not differ by more than about 3%. However, PBE0 has a substantially higher intercept (0.95 eV) than other methods, whereas HSE is closest to zero (0.36 eV), and B3PW91 and B3LYP are in between that (≈0.65 eV). Hence, with respect to trends, all functionals are similar but HSE appears to be slightly better again.

Lastly, let us consider a measure of rank correlation to judge the quality of predicted band gaps: the Kendall \( \tau \) rank correlation coefficient. In order to construct this coefficient, we first need to define the pairs \((g_i^{\text{expt}}, g_i^{\text{calc}})\), where \(g_i^{\text{expt}}\) and \(g_i^{\text{calc}}\) are the experimental and calculated gaps, respectively. The pairs \(i\) and \(j\) and then said to be concordant if \(g_i^{\text{expt}} < g_j^{\text{expt}}\) and \(g_i^{\text{calc}} < g_j^{\text{calc}}\), or if \(g_i^{\text{expt}} > g_j^{\text{expt}}\) and \(g_i^{\text{calc}} > g_j^{\text{calc}}\); tied if \(g_i = g_j\); and discordant otherwise. The Kendall \( \tau \) correlation coefficient is then

\[ \tau = \frac{P - Q}{\sqrt{(P + Q + T)(P + Q + U)}} \] (2)

where \( P \) is the number of concordant pairs, \( Q \) the number of discordant pairs, and \( T \) and \( U \) the number of ties in the experimental and calculated set of band gaps, respectively. Thus, \( \tau \) measures the similarity in the orderings of two datasets: a \( \tau \) value of 1 indicates perfect agreement in the ordering, whereas \( \tau = -1 \) indicates complete disagreement. It can, in a way, be considered a measure of qualitative agreement. We see in Table I that all four hybrids provide essentially the same \( \tau \) value (≈0.85), and thus have all similarly good qualitative agreement with experiment.

One could also analyze the distribution of the error of pairs of methods, and ask whether the distributions differ statistically. This is done in Table II using the Wilcoxon signed-rank test for the absolute error data. The null hypothesis of the test is that the difference between pairs (of absolute errors for two methods) follows a symmetric distribution around zero because each set has the same statistical distribution, resulting in similar positive and negative differences; the null hypothesis is rejected when the test statistic \( W \) is greater than a certain \( W_{\text{crit}} \) determined by the sample size and level of confidence. A \( p \)-value associated with the test represents the probability of obtaining the observed distribution, or a more asymmetric distribution around zero, assuming that the null hypothesis is true; \( p \)-values larger than 0.05 have been related with a lack of meaningful difference in the context.

TABLE II: Wilcoxon signed-rank test \( W \) values for the absolute errors of pairs of hybrid functionals using the data that appears in Figure 1. The associated \( p \)-values are included in parenthesis. The value for \( W_{\text{crit}} \) with a 95% confidence interval is 107; \( W < W_{\text{crit}} \) for a pair of methods indicates that their absolute errors show the same statistical distribution around zero.

|       | HSE | B3PW91 | B3LYP | PBE0 |
|-------|-----|--------|-------|------|
| Expt. |     | -227(0.01) | 214(0.01) | 128(0.14) |
| HSE   | 42(0.61) | - | 18(0.83) | 227(0.01) |
| B3PW91| 84(0.31) | 18(0.83) | - | 214(0.01) |
| B3LYP | 84(0.31) | 18(0.83) | - | 214(0.01) |
| PBE0  | 128(0.14) | 227(0.01) | 214(0.01) | - |
of method comparison [15]. Based on both the W test statistic and its associated p-values, the results in Table [1] indicate no significant statistical difference (considering a commonly used 95% confidence interval) in the absolute error distributions of HSE, B3PW91, and B3LYP. The W statistics suggest, however, that differences exist between these three functionals and PBE0; the p-values also imply significant difference between PBE0 and both B3PW91 and B3LYP. If the same analysis is carried out for signed error and absolute percentage error, all pairs of functionals come out as being statistically different except for the B3PW91/B3LYP pair. The results of the Wilcoxon signed-rank test support the more qualitative comparison of the ME, MAE, and MAPE discussed above, and put such a discussion in a rigorous statistical framework.

On the underestimation of large band gaps. As shown here and in other works [10], even hybrid functionals underestimate band gaps when these are considerably large. No single hybrid with a fixed amount of HF exchange can accurately describe small and large band gap materials simultaneously: the fraction of nonlocal exchange needed in a solid depends on its dielectric function [15–17]. Increasing the percentage of HF exchange in a hybrid functional increases the band gap, which allows for the description of larger band gap insulators, but causes overestimation of semiconductor band gaps. A way to optimize the fraction of HF exchange in hybrid functionals based on a local estimator of the dielectric function has been proposed in Ref. [16]. With this procedure, one obtains, for example, that HSE predicts a band gap 22.29 eV for solid Ne (which has an experimental gap of 21.70 eV), while still giving good semiconductor band gaps (e.g., 0.82 eV for Ge, compared to 0.74 eV from experiment) [10]. The method is applicable to global hybrids too: an optimized PBE0 gives gaps of 21.88 and 0.68 eV for Ne and Ge, respectively, compared to 15.14 and 1.31 eV predicted by standard PBE0 [10].

As another example of how even screened hybrids can describe large band gap insulators, consider the HSE-HF functional of Ref. [18], which is simply HSE06 but with 100% exact exchange in the short range. Figure 2 shows the band gap of NaCl calculated by HSE-HF as a function of the range separation parameter, ω. The smaller this parameter is, the larger the fraction of HF exchange incorporated. At small ω values, the band gap is overestimated; at ω ≈ 0.2–0.3 au the value is close to experiment; and at higher ω values the gap is underestimated. This is just one example, but this behavior is very general and this observation is in agreement with previous works showing that long-range HF exchange is seldom necessary and that the physics behind the more realistic properties computed with hybrid functionals is the reduction of the self-interaction error by incorporation of substantial amounts of exact exchange [15, 19]. Thus, a short-range screened hybrid can describe even large band gaps insulators and one does not need to resort to more expensive global hybrids. Note also that although, in principle, long-range HF exchange should be included when describing solids, its effect should be mostly canceled by correlation. Inadequate treatment of correlation is the reason why the HF method and typical global hybrid functionals cannot describe metallic behavior. Hence, it is convenient to eliminate long-range exchange and use a screened hybrid to more adequately describe the physics of an extended solid.

Additional Considerations. We have focused on band gaps calculated using experimental geometries and observed little overall difference between the different hybrids tested. However, other properties, including geometries, can be significantly different. For example, the MAE of B3LYP for lattice constants (0.053 Å) is more than twice larger than the MAEs of HSE and PBE0 (0.024 and 0.022 Å, respectively) [20]. Atomization energies of solids are also significantly worse when calculated with B3LYP (MAE = 0.59 eV) as compared to HSE and PBE0 (MAE ≈ 0.26 eV); B3PW91 improves over B3LYP, but it is not as good as the latter two functionals (MAE = 0.39 eV) [20]. The failure of B3LYP is attributed to not correctly reproducing the homogeneous electron gas behavior, and B3PW91 improves upon it by fulfilling this limit [20].

Even when considering only band gaps and using experimental geometries, another important consideration arises: the quality of the experimental data. In particular for the band gaps, there can be a significant variability depending on factors such as the method of measurement and temperature (which sometimes is not reported). For example, band gap values of 3.6, 2.6, 2.5, 2.1, 2.7, and 5.43 eV have been reported for CoO (see Supporting Information of Ref. [9]). This variability leads to the question: What value should one use as reference? One can average of all these values (3.16 eV) and obtain very good agreement with B3PW91 (calculated here as 3.06 eV) [9]. However, this approach may not be appro-
appropriate in this case: the experimentally reported 5.43 eV value is most likely the direct gap, whereas the smaller values correspond to the indirect gap \cite{21}. Even if one decided to follow this methodology, the standard deviation of the six experimental band gap values for CoO is 1.22 eV, and hence all functionals predicting values in the wide range between 1.90 and 4.34 eV would have to be considered to be in agreement with experiment. The often cited value of 2.5 eV for the CoO band gap is in reasonable agreement with the HSE indirect gap of 2.82 eV. An additional consequence of the variability in the reference data is that small differences in the average errors of different functionals are not really meaningful. Thus, based on, say, the MAEs in Table \ref{table1}, it would be hard to tell whether HSE, B3PW91, or B3LYP are truly closer to experiment. Furthermore, adding to the level of uncertainty, experimental reference band gaps are assumed to be an ideal measurement of the fundamental band gap (\textit{i.e.}, the difference between the ionization potential and electron affinity). However, this is not always the case \cite{10}, whereas the calculation of generalized Kohn–Sham band gaps (as done here) is a rigorous and honest approximation to the fundamental band gap \cite{22}. Considering all of this, and the fact that the use of different, finite basis sets leads to further changes in predicted band gaps, it would be unreasonable to assert that a certain functional is decidedly better than another for band gap predictions if the differences in the error measures are small (\textit{e.g.}, less than about 0.1 eV or, depending on the case, even more than this, as shown by the CoO example above). Statistical methods such as the ones used here should be considered for assessing whether the differences are significant.

The importance of careful inspection of reference data from the literature and the use of a consistent methodology in benchmark calculations applies not only to density functionals. The GW method, for example, is quite sensitive to a number of variables: the input (reference) wavefunction, number of empty bands included, dielectric matrix energy cut-off, etc. To illustrate this, consider the GW band gaps of ZnO used as reference to compare against B3PW91 in the Supporting Information document of Ref. \cite{14}. The experimental band gap is 3.44 eV; the GW gaps in the literature vary from 0.1 to 4.61 eV. The worst value, 0.1 eV, was computed using the static Coulomb hole and screened exchange (COHSEX) approximation, which is generally not recommended for quantitative calculations \cite{24}. Shih \textit{et al.} \cite{24} report that conventional GW approaches can provide accurate estimates of the ZnO band gap if convergence in the evaluation of the Coulomb-hole self-energy is carefully resolved. Further examples with large variance (more than 1 eV) in the GW gaps cited in Ref. \cite{14} include Si (0.56–1.91), GaAs (1.09–3.77), SiC (1.8–2.88), AlP (1.88–3.1), CdS (2.11–3.41), CoO (2.4–4.78), GaN (2.75–3.82), MnO (2.34–4.39), ZnS (1.52–4.15), NiO (1.74–5.0), diamond (5.59–6.99), LiCl (8.75–10.98), and LiF (13.13–16.17) among others. It is worth pointing out here that Ref. \cite{14} concluded that B3PW91 was superior to other functionals by comparing against GW, assuming that GW is more accurate than these (other) functionals. However, as shown here, such comparisons are not straightforward and B3PW91 does not appear to have any decisive advantage over other commonly used hybrids when the direct comparison is done using identical geometries, basis sets, etc.

The effects of spin-orbit coupling should also be mentioned here. We have not included these effects in our calculations, as such corrections are not widely available in commercial codes. Nevertheless, our conclusions are not affected by neglecting spin-orbit coupling because of, mainly, two reasons. The first one is that we have not included compounds for which spin-orbit coupling effects are important in our test set. The second one is related to the fact that spin-orbit corrections apply in the same way to all functionals: \textit{i.e.}, if we use experimental spin-orbit coupling values to correct the band gaps, the shift in the error is the same for every functional. Here, we observe that various hybrids provide rather similar results, with no clear winner in all cases and a different method being “better” depending on the error measure. After spin-orbit coupling corrections, the functionals will still provide similar results because these corrections are identical for all of them.

Lastly, an important consideration for calculations on extended systems is computational cost. The increase in CPU time when going from a semilocal functional, to a screened hybrid, to a global hybrid depend on the type of basis set employed. Here, we have used codes based on localized gaussian basis sets, which are not as efficient for semilocal functionals but have a less dramatic increase in cost for hybrids as compared to plane-wave based codes. This means that, in plane-wave implementations of DFT, screened hybrids are more strongly favored over global hybrids concerning CPU time.

\textbf{Conclusions.} After carrying out a thorough comparison between four popular hybrid functionals for predicting band gaps using a consistent methodology in the calculations, we did not find truly significant differences in the overall performance of three of them: HSE, B3PW91, and B3LYP (the latter two in particular give very similar gaps). PBE0 tends to overestimate band gaps but the error is systematic and the statistical analysis shows that all four functionals provide good qualitative agreement with the reference data. Nevertheless, our results do suggest that HSE is more accurate for typical semiconductors. The analysis here shows that there is nothing particularly remarkable about the band gaps predicted by B3PW91 \cite{14} when they are compared to other global hybrids using a similar amount of HF exchange. We have also shown how screened hybrids can incorporate larger fractions of HF exchange in the short range to compensate for the effects that result from screening in large gap insulators, and ways to tune the mixing parameter that solve this problem have also been proposed in the literature \cite{15,17}. Global hybrids also underestimate very
large band gaps, and can too be modified to include more exchange to solve this problem, but at a higher computational cost. Thus, based on cost and accuracy considerations, we recommend HSE for semiconductors and tuned versions of HSE for larger gap cases.

In summary, screened hybrids can describe the same type of problems that global hybrids do, but in addition, they provide access to modelling metallic behavior, have a much lower computational cost, and are easier to converge. Our conclusion is therefore that a screened hybrid like HSE is preferable over global hybrids for the black box prediction of band gaps of novel materials.

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Supporting Information Available. Details about the calculations, as well as band gap data used in this work.

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