Credit Spread Volatility

Nurture strength of spirit to shield you in sudden misfortune.

- Max Ehrmann, 1927

Three Desiderata

Successful portfolio construction involves not only the selection of superior securities and investment themes, but also the assessment of the ranges of outcomes of the portfolio’s exposures. The best portfolios carefully balance risk and reward.

An important part of assessing portfolio risk is the identification of risk factors, which are the variables that will affect many or all of the securities in the portfolio. For fixed-income portfolios, some obvious risk factors are the prevailing key rates of interest in the portfolio’s base currency, general levels of credit spreads and foreign exchange rates.

To estimate the range of future returns of a portfolio—either the total return, or the return relative to a benchmark—we need, among other things, to measure or estimate the following information:

- The exposures the portfolio has to its risk factors, and
- The likely ranges of outcomes of those risk factors.

The ranges of outcomes of risk factors can be described statistically. We know for example that over the 10-year period from 2007 to 2016, daily absolute changes in the US Treasury (UST) 10-year rate were less than 13 basis points (bps), 95% of the time. But over the year 2017, the range of UST 10-year daily absolute moves shrank dramatically to less than 7 bps, 95% of the time.

In this note we’ll talk about our approach to estimating credit spread risk. To do this, we need not only identify statistically reliable phenomena, but we also must identify reliable relationships between our portfolio and those phenomena. As such, we target the three desiderata in Exhibit 1.

We’ll discuss each of these, starting with statistical predictability.

Boredom and Terror

As has been said of war, some credit markets can be characterized as long periods of boredom punctuated by moments of sheer terror. Credit spreads, which capture the higher yields that credit-risky bonds command over non-credit-risky bonds, exemplify this dynamic.
Exhibit 2 shows day-over-day changes in credit spreads for the Bloomberg Barclays U.S. Credit Corporate Index. Each vertical line is a day’s change in spread; for example, the largest upward line in the graph hits 43 bps on the vertical scale. It occurred on September 15, 2008 when the index spread jumped from 327 bps to 370 bps in reaction to the bankruptcy of Lehman Brothers.1

That was sheer terror. The “long periods of boredom” are also apparent here. From November 10, 2011 to January 19, 2016 there wasn’t a single day when spreads changed by more than 6 bps up or down from the previous day.

The credit spread behavior in Exhibit 2 does not satisfy our third desideratum—it is not statistically predictable. One problem is apparent from a visual examination and requires no statistical analysis: there are very large spikes, such as the one on September 15, 2008, that are massively different in magnitude from the usual pattern. The prevalence of such spikes is variously called “fat tails” or “black swans.”2 They are similar to big earthquakes—they don’t happen very often, but when they do they have a huge impact.

Another problem is what financial economists call “volatility clustering.” Again, just by eyeballing Exhibit 2, we can see that daily move amplitude was elevated for about a year before the giant spike on September 15 and continued to be elevated for about a year afterward. This is very much like human emotion: during a period of fear or anxiety, symptoms such as elevated pulse and hyper vigilance persist and possibly lead to panic. But eventually the adrenaline washes out and things cool down.

Standard fixed-income mathematics requires the multiplying of an item’s option-adjusted spread duration3 (OASD) times the item’s credit spread changes (e.g., changes that look like Exhibit 2) to compute the contribution to the item’s rate of return. Although spread duration is a relatively stable quantity, the statistical problems of Exhibit 2 will affect estimation of the range of rates of return through this multiplication.

Exhibit 3 shows an ideal world in which the problems of Exhibit 2 have been removed. Exhibit 3 is in simple ways statistically identical to Exhibit 2, with the same average daily move in spreads, and the same volatility of moves in spreads. But we’ve removed the problematic fat tails and volatility clustering.

The world of Exhibit 3 is one in which portfolio managers can anticipate the ranges of outcomes of their decisions quite accurately. This is nice predictable boredom without the terror, which is a good thing when you’re managing portfolios. Is there a way that we can inhabit a world of benign boredom?
The Introduction of Duration Times Spread

In 2005, a group from Lehman Brothers and the Robeco Group outlined an approach to dealing with the moments of boredom and terror—the volatility clustering—that credit spreads display.

The authors noted that credit managers looked at exposure measures such as percentage of portfolio holdings and spread duration to assess their credit risk. But the Lehman/Robeco group adduced empirical evidence in favor of duration times spread (DTS) as a better measure.

To illustrate what DTS is and how it works, consider the following sample portfolio consisting of two securities, each comprising half of the portfolio:

1. Security A has a spread duration (OASD) of 3 years and a spread (SA) of 150 bps;
2. Security B has a spread duration (OASD) of 3 years and a spread (SB) of 450 bps.

The percentage holdings—50% for each Security A and Security B—are fairly stable, so they satisfy our first desideratum above. But they don’t satisfy the second desideratum: percentage holdings do not provide a realistic picture of the relationship between the security position and the risk of general changes in credit spreads. Longer-duration securities will move more per percent holding than shorter-duration securities, and riskier securities will move more than less risky securities. Using percentage as the exposure measure does not take these effects into account.

Spread duration—or, more accurately, contribution to spread duration—is more widely used by practitioners as an exposure measure. The example portfolio’s spread duration is three years since both component securities have three years’ spread duration. If both Security A’s spread and Security B’s spread widen by 10 bps, the spread duration figure tells us that we expect to lose 30 bps from spread behavior in this portfolio.

But it’s unlikely that A’s spread and B’s spread will widen by the same 10 bps. B is riskier than A, as evidenced by its much higher spread. It is more likely—and is confirmed empirically—that in a general spread widening, B’s spread will move more than A’s spread. For example, suppose some index of general credit spreads is at 150 bps (coincidently the same as Security A’s spread), and then the index spread widens by 10 bps. Then Security A’s spread is also likely to widen by 10 bps. But Security B will probably widen by something closer to triple that (10 bps times 450 bps divided by 150 bps), or 30 bps.

DTS is a convenient way of doing the accounting for this phenomenon. The DTS of our 50-50 example portfolio is 0.5x3x150+0.5x3x450=900. The index spread widened by 10 bps on a 150 bps base, or 10/150=6.7%.
If we multiply 900 times 6.7%, we get an expected 60 bps loss rather than the 30 bps we got from the simple spread duration calculation under the assumption of identical spread changes.

The basic idea of DTS is therefore that the percentage change (6.7% in our example) is the relevant figure to capture spread movement. The arithmetic change (10 bps) is not directly relevant.

Does DTS take us to the ideal world of Exhibit 3? In other words, if we look at percentage changes rather than arithmetic changes, do we get the very predictable pattern of Exhibit 3 rather than the problematic (fat-tailed, volatility-clustered) pattern of Exhibit 2? Exhibit 4 suggests an answer:

Visually, Exhibit 4 is somewhere between the problems of Exhibit 2 and the regularity of Exhibit 3. There are clearly still fat tails—big movements that differ from the regular pattern far more than anything in Exhibit 3. For example, September 15, 2008 is still clearly visible as an outlier in Exhibit 3—it represents an unusual 13.2% increase in spread. However, it also seems visually clear that there is more regularity in Exhibit 4 than there was in Exhibit 2.7

So from Exhibit 4 and its associated statistics we see that DTS doesn’t quite get us all the way to the nirvana of perfect boredom. But it is a step in the right direction toward Exhibit 1’s third desideratum.

Where Does the Weirdness Go?8

The “trick” in DTS is to confound an exposure measure with a risk measure. In our example above, DTS was 900 at the beginning of the example period. But by the end of the period, DTS was 960—not because the manager increased exposures, but because the market moved. DTS removes some unpredictability from the risk factor by transferring that unpredictability to the exposure measure. It improves on delivery of desiderata 2 and 3 at the expense of desideratum 1.

As an example, Exhibit 5 shows the DTS exposure of Western Asset’s Corporate Bond Fund since 2007. This fund’s option-adjusted spread duration (OASD, blue line) was reasonably smooth over the 10 years shown in Exhibit 5. But the DTS jumped and subsided due to market shocks. The procyclical (lagging indicator) nature of DTS is apparent: DTS would have indicated that exposures were incredibly high just after the worst incidents of spread widening, and that exposures were low during periods of unusually tight spreads.

Thus the increased stability of Exhibit 4 over Exhibit 2 comes at the expense of less stability in the exposure metric, and less clarity about which changes are due to active portfolio manager decisions and which exposures are due to market movements.

Exhibit 4: Daily Percentage Changes in US Investment-Grade Corporate Credit Spreads

Source: Bloomberg, Barclays, Lehman, WISER. As of 28 Nov 17

Note: Exhibit 4 shows daily percentage changes in US investment-grade corporate credit spreads from 2005 to 2017. The changes are displayed on a logarithmic scale, showing the volatility clustering over time.
The standard spread duration approach fails to deliver two of the desiderata, while DTS only fails on one. We are mindful of market strategist M. Loaf’s sage advice: “Two Out of Three Ain’t Bad.” Would we be over-reaching if we tried to satisfy all three of our desiderata?

**Adjusted Spread Duration**

There’s another way to do the accounting that gets to the same place as our DTS example without violating desideratum 1. We can look at the *adjusted spread duration (ASD)* of each security. Since Security A has the same spread as the index, we expect changes in Security A to be similar in magnitude to changes in the index. But Security B has a much higher spread, so we adjust its spread duration upward to reflect the likelihood of more intense moves as the index changes.

For our example, we would say ASD =OASD =3 years; that is, no adjustment to Security A’s spread duration is needed. But for Security B we would say ASD = 450*OASD /150=450*3/150=9 years; in other words, we would adjust the spread duration up from 3 years to 9 years. The adjusted spread duration of the 50-50 portfolio is then 0.5x3+.5x9=6 years.

If we apply the portfolio ASD of 6 years to the 10 bps index spread move in our example, we get the same 60 bps loss that we got from DTS. So DTS and ASD attempt to answer a question about what is likely to happen to our portfolio if a general spread level changes, while unadjusted spread duration attempts to answer a question about a not very realistic situation in which every spread in the portfolio changes by the same amount.

ASD is stable and generally not procyclical (desideratum 1). Some adjusted spread durations may change because of market movements, but the fact that the adjustment is made by taking a ratio of credit spreads means that general spread changes will tend not to have too big an effect on overall ASD. Further, ASD’s relationship to risk factor movements is realistic (desideratum 2). Exhibit 6 adds ASD to Exhibit 5, demonstrating ASD’s relative stability.

The ASD is reliably higher than the OASD of the portfolio because the credit spread curve is upward-sloping through virtually the entire period shown, so longer-duration securities get heavier adjustments than shorter-duration securities.

Unfortunately for the return calculation to work properly, we’ve had to revert to arithmetic spread changes (10 bps in the example) as a risk factor. So here, we’ve lost desideratum 3.
Time-Varying Volatility

Fortunately statisticians have studied the nature of time-varying volatility in financial markets. In fact, one of them, Robert Engle, won a Nobel Memorial Prize in Economic Sciences for his work in this area. The model that Engle developed is called ARCH; this was refined to GARCH by Tim Bollerslev and extended by hundreds of researchers over the last three decades.

DTS hard-codes a belief that the immediate current volatility environment (as represented by spread levels) will continue into the future. The GARCH-type approach carefully balances three components when making volatility forecasts:

- The current level of volatility;
- The most recent observation—which in our example is a change in credit spreads; and
- A long-term equilibrium level of volatility.

A short-term estimate of volatility will tend to be dominated by the recent environment, while a long-term estimate of volatility will tend to be dominated by the equilibrium level of volatility.

Exhibit 7 shows what happens when we apply GARCH to Exhibit 2.

As with Exhibit 4, Exhibit 7 does not quite bring us to the full predictability of Exhibit 2. There are still sur-
prises. But the statistics\(^6\) show that Exhibit 7 mitigates the fat-tail problem even better than Exhibit 4 while virtually eliminating volatility clustering.

With the use of the more powerful time-varying volatility estimation provided by GARCH-type models, we can use adjusted spread duration to realistically assess credit spread risk in portfolios. We can do so in a way that allows portfolio managers to see clearly the effects of their active decisions, avoiding procyclicality. We can separate the activity of describing current risk exposures from the activity of estimating future volatility levels.

**Multidimensional Risk**

Credit spreads are not monolithic. Short-term credit spreads sometimes part ways with long-term credit spreads; technology spreads don’t always go the same way as financial sector spreads; US credit spreads may diverge from European credit spreads, etc. Thus it’s unrealistic to think that credit spread exposure can be captured with a single number—whether that number is (unadjusted) spread duration, DTS or ASD.

To fully capture credit spread exposures and the associated risks, a multidimensional approach with a full risk model—such as the Western Asset Information System for Estimating Risk (WISER)—is necessary. A full risk model estimates relationships among many types of credit spreads, along with other risk factors.

So when we talk about “the” (unadjusted) spread duration, or “the” DTS, or “the” ASD of a portfolio, we are oversimplifying. It’s like talking about “the” skill of a human being: do we mean skill at playing the piano, or breaking into bank safes, or sailing upwind or managing portfolio risk? We could try to rank a person, say on a scale of 1 to 100, on each of these skills. But it would not make much sense to add these ranks together.

The urge to cram multidimensional effects into a single linear dimension is, however, quite strong. So when we can’t go nonlinear (full-risk system) and we can’t resist the urge to use a single number, we look for a single broad risk factor to use as the denominator of the ASD calculation for all spread-risky securities in a portfolio.

For a broad market US portfolio, we might adjust credit spread durations to give us an idea how that portfolio would react to the risks represented by changes in the Bloomberg Barclays U.S. Credit Corporate Index. For a European high-yield credit portfolio, we might make the adjustment with respect to the Bloomberg Barclays Pan-European High Yield 2% Issuer Constrained Index. This means that the same high-yield security might show a different ASD depending on the benchmark of the portfolio it’s in. So adjusted spread durations are properties of a pair (security, benchmark) and not of the security alone.

Internally at Western Asset, we have assigned reference indices used for the ASD adjustment on a strategy-by-strategy basis. For example, for long duration portfolios a long investment-grade credit index OAS is used for the adjustment. For high-yield portfolios, a high-yield index OAS is used.

**Summary**

We’d like to have an approach to assessing credit spread risk that satisfies the desiderata in Exhibit 1. This would give us better information as we seek to estimate risk in portfolios.

The standard approach to estimating credit spread risk is to compute spread duration exposures, while using arithmetic changes in general spread levels to represent overall credit spread risk. But a change in general spread levels will not flow through realistically to individual securities using this approach, and the statistical properties of arithmetic changes in spreads are not as steady as we’d like.

The DTS approach improves on some of the failings of the standard approach, but at the expense of creating unstable procyclical exposures with a hardwired view of future volatility.

Using an adjustment to spread duration along with a powerful statistical technique to predict time-varying
volatility gets us the best of all worlds, satisfying all the desiderata. The evolution of approaches is shown in Exhibit 8.

In fact any combination of quantitative and qualitative assessments can be used to assess likely future risk levels. For example, in the low market volatility environment that prevails at the time of this writing, we believe that portfolios will display unusually low volatility over the short term. But we also believe that over the longer term, there will be a return to a higher volatility environment. The adjusted spread duration method allows us to separate this assessment from the assessment of our exposures.

Three out of three ain't bad.

ENDNOTES

1 Ironically, the index that captured this behavior was created and computed by Lehman Brothers itself. The fixed-income index business originated by Lehman was taken over by Barclays in 2008 and sold to Bloomberg L.P. in 2016.

2 The use of the phrase “Black Swan” to indicate rare but impactful phenomena was popularized by Nicholas Nassim Taleb in his 2007 book *The Black Swan: The Impact of the Highly Improbable*. The actual statistic that describes tail thickness is kurtosis; fat tails are called *leptokurtosis*. Virtually all financial time series data are leptokurtic.

3 See Appendix for a definition of option-adjusted spread duration.

4 Exhibit 2 contains 3,100 daily observations of spread changes. Exhibit 3 was generated by taking 3,100 random numbers drawn from a normal distribution with the same mean and standard deviation as the observations in Exhibit 2.


6 Of course percentage holdings do change in buy-and-hold portfolios as mark-to-market prices change, but this effect is relatively slow compared to other effects we will discuss.

7 For the statistical reader, the kurtosis of the observations in Exhibit 2 is 50.6, while Exhibit 4’s kurtosis drops to 15.8. The serial correlation of squared observations in Exhibit 2 is 27.6%, while the serial correlation of squared observations in Exhibit 4 is 23.7%.

8 *Where Does the Weirdness Go?* is the title of a 1996 book about quantum mechanics by David Lindley.


10 The adjustment leading to the “A” in “ASD” is the proportional spread multiplier defined more precisely in the Appendix. This is in addition to the industry convention of adjusting for optionality in securities, usually denoted “OASD.” ASD is thus both spread-proportion adjusted and option-adjusted.

11 ARCH stands for Auto-Regressive Conditional Heteroskedasticity.


13 In response to the kinds of problems we have noted, Barclays eventually suggested modifications to DTS. Instead of using “spot” spread, in some cases they use a smoothed (over time) spread. They also recommend restricting the range of spreads over which DTS is used. Our view is that Occam’s Razor applies here: if you have to make a lot of changes to the original idea to get it to work, it’s time to scrap it and start over.

14 Exhibit 7 was generated by applying an exponentially weighted moving average (EWMA) variance adjustment to Exhibit 2. This is a simplified form of GARCH.

15 Kurtosis was 7.3; serial correlation of squared observations was 0.04.
Appendix

To be more precise about the quantities we discussed, let’s suppose we have a portfolio “x” where weight $w_1^x$ is invested in the first security in the universe of investable securities, $w_2^x$ in the second investable security... and so forth up to $w_n^x$ in the n-th and final security in the universe of investable securities. These weights add to 1 (100%)—mathematically that is the budget constraint $\sum_{i=1}^{n} w_i^x = 1$.

In the standard fixed-income framework, we are able to list all planned future cash flows of an instrument and discount them to the present with an appropriate default-free discount curve. The standard framework further assumes that we can add a spread to the discount rates in order to match the market price of the instrument; this spread accounts for credit risk, liquidity and other factors. The sensitivity of the instrument’s price to changes in this spread is called the spread duration.

We will assume that spread durations are option-adjusted: for example, if a bond has a call schedule that gives the issuer the option to pay off the bond before its maturity date, then there is optionality associated with the bond. The calculations of spread and spread duration need to take this optionality into account. Sometimes the option adjustment is done by generating multiple random (Monte Carlo) paths into the future; in some of those paths the options will be in the money, and in others the options will be out of the money. The option adjustment would average the behavior across the multiple paths.

The resulting spreads and spread durations are called option-adjusted and are abbreviated OAS and OASD, respectively. If there is no optionality in the instrument, then option-adjusted spread duration is just the same as spread duration. We will assume that everything is option-adjusted, so if we refer just to “spread” or “spread duration,” we mean “option-adjusted spread” and “option-adjusted spread duration,” respectively.

Security is duration-times-spread $DTS_i$ is simply $DTS_i = SD_i OAS_i$. The duration-times-spread of portfolio $x$, $DTS_x = \sum_{i=1}^{n} w_i^x DTS_i$, i.e. it is the weighted sum of the DTS’s of the individual securities in the portfolio.

The (unadjusted) spread duration of portfolio $x$ is $SD_x = \sum_{i=1}^{n} w_i^x SD_i$, and the option-adjusted spread of portfolio $x$ is $OAS_x = \sum_{i=1}^{n} w_i^x OAS_i$.

Adjusted spread duration assumes there is an index or benchmark that is appropriate for adjustments. We define $ASD_i = SD_i \left(\frac{OAS_i}{OAS_b}\right) = \left(\frac{DTS_i}{OAS_b}\right)$, where b is the appropriate benchmark. This means that the adjustment to spread duration might differ by context if the security is in two different portfolios with two different benchmarks.

The adjusted spread duration of a portfolio is $ASD_x = \sum_{i=1}^{n} w_i^x ASD_i = \frac{1}{OAS_b} \sum_{i=1}^{n} w_i^x DTS_i = \frac{DTS_x}{OAS_b}$. In other words we can get an adjusted spread duration by adding up the weighted adjusted spread durations of each security in the portfolio, or we can just divide the DTS of the portfolio by the OAS of its benchmark. Note that the adjusted spread duration of the benchmark is NOT the spread duration of the benchmark, i.e. $ASD_b = \frac{DTS_b}{OAS_b} \neq \sum_{i=1}^{n} w_i^b SD_i$. The use of ratios makes the adjusted spread duration metric less vulnerable to the current market environment.

Contributions to adjusted spread durations can also be computed for segments of portfolios and subtracted to get benchmark-relative exposures of segments. Contributions to adjusted spread duration are better than contributions to DTS, again because of the avoidance of procyclical.

GARCH(1,1) models estimate three parameters $\alpha, \beta, \text{and } \gamma$ so that $\sigma_t^2 = \gamma + \alpha y_{t-1}^2 + \beta \sigma_{t-1}^2$ is a good predictor of the volatility of the time series $y_t$. Generally we would like to see that the de-GARCHED time series $y_t / \sigma_t$ moves in the direction of white noise, with low kurtosis and a good Ljung-Box test statistic.
Past results are not indicative of future investment results. This publication is for informational purposes only and reflects the current opinions of Western Asset Management. Information contained herein is believed to be accurate, but cannot be guaranteed. Opinions represented are not intended as an offer or solicitation with respect to the purchase or sale of any security and are subject to change without notice. Statements in this material should not be considered investment advice. Employees and/or clients of Western Asset Management may have a position in the securities mentioned. This publication has been prepared without taking into account your objectives, financial situation or needs. Before acting on this information, you should consider its appropriateness having regard to your objectives, financial situation or needs. It is your responsibility to be aware of and observe the applicable laws and regulations of your country of residence.

Western Asset Management Company Distribuidora de Títulos e Valores Mobiliários Limitada is authorised and regulated by Comissão de Valores Mobiliários and Banco Central do Brasil. Western Asset Management Company Pty Ltd ABN 41 117 767 923 is the holder of the Australian Financial Services Licence 302160. Western Asset Management Company Pte. Ltd. Co. Reg. No. 200007692R is a holder of a Capital Markets Services Licence for fund management and regulated by the Monetary Authority of Singapore. Western Asset Management Company Ltd is a registered Financial Instruments Business Operator and regulated by the Financial Services Agency of Japan. Western Asset Management Company Limited ("WAMCL") is authorised and regulated by the Financial Conduct Authority ("FCA"). In the UK and EEA countries, this communication is a financial promotion intended for distribution to Professional Clients only as defined by the FCA or MiFID II rules.