Is All Formulaic Language Created Equal? Unpacking the Processing Advantage for Different Types of Formulaic Sequences

Gareth Carrol
University of Birmingham, UK

Kathy Conklin
University of Nottingham, UK; Max Planck Institute for Psycholinguistics, the Netherlands

Abstract
Research into recurrent, highly conventionalized “formulaic” sequences has shown a processing advantage compared to “novel” (non-formulaic) language. Studies of individual types of formulaic sequence often acknowledge the contribution of specific factors, but little work exists to compare the processing of different types of phrases with fundamentally different properties. We use eye-tracking to compare the processing of three types of formulaic phrases—idioms, binomials, and collocations—and consider whether overall frequency can explain the advantage for all three, relative to control phrases. Results show an advantage, as evidenced through shorter reading times, for all three types. While overall phrase frequency contributes much of the processing advantage, different types of phrase do show additional effects according to the specific properties that are relevant to each type: frequency, familiarity, and decomposability for idioms; predictability and semantic association for binomials; and mutual information for collocations. We discuss how the results contribute to our understanding of the representation and processing of multiword lexical units more broadly.

Keywords
formulaic language, lexical processing, idioms, collocations, binomials, eye-tracking

Introduction
Cheetahs and Ferraris are two examples of members of the category “things that are fast.” Beyond this broad similarity, there is not much that makes the two particularly comparable, and it is clear...
that the mechanisms that make each one fast are very different. In linguistics, formulaic language is an example of something that has sometimes been defined just as broadly. It encompasses a broad range of multiword sequences that fulfill a number of communicative functions (Wray, 2002, 2008), and knowledge of such sequences is an important part of how we use language. For example, in English we implicitly know to describe coffee as “strong” not “powerful,” or to ask for “salt and pepper,” not “pepper and salt.”

Two features are common to all examples of what is generally considered under the heading of formulaic language. The first is that they must be recurrent, in the sense that they occur in natural language more frequently than comparable novel phrases. What counts as the threshold for “frequent” is an open question, but widespread evidence now supports the second feature: faster processing of recurrent sequences compared to “novel” control phrases, or, put another way, frequency effects at the multiword level, in line with usage-based accounts of how language develops and is organized (e.g., Bybee, 2006; Goldberg, 2006; Tomasello, 2003). Often, claims are made about the “holistic” nature of formulaic sequences, suggesting that all recurrent sequences are simply stored in the lexicon and retrieved directly, although the nature of what is meant by this may also be quite variable (c.f. Wray, 2012, p.234).

Beyond this very broad designation, what counts as “formulaic” can differ widely, and a view of “holistic” representation may not tell the whole story. Formulaic sequences vary along a number of important dimensions (Titone, Columbus, Whitford, Mercier, & Libben, 2015), such as their degree of fixedness/conventionalization, their schematicity (whether they allow for internal variation through open “slots”), their semantic unity, the degree of compositionality, and the function they perform (see Buerki, 2016, for a very useful overview). As a result, when we look at the types of phrase that are usually included in the category of formulaic language, we may find that they are just as difficult to compare as cheetahs and Ferraris. For example, idioms are semantically opaque, self-contained figurative phrases, such as kick the bucket. Multiple studies attest to the idiom superiority effect, whereby idioms are processed more quickly than matched control phrases (e.g., Gibbs, 1980; McGlone, Glucksberg, & Cacciari, 1994; Rommers, Dijkstra, & Bastiaansen, 2013; Swinney & Cutler, 1979; Tabossi, Fanari, & Wolf, 2009). In comparison, collocations are very broadly defined as pairs of words that occur together more frequently than we would expect by chance, such as strong coffee. They are generally transparent and have a literal meaning that is the result of combining the component words, but these too have been shown to be processed more quickly than non-formulaic comparators (e.g., Bonk & Healy, 2005; Durrant & Doherty, 2010; Sonbul, 2015; Vilkaite, 2016; Wolter & Gyllstad, 2011). Other types of sequence have been shown to demonstrate a similar processing advantage: binomials (e.g., Arcara et al., 2012; Siyanova-Chanturia, Conklin, & van Heuven, 2011); phrasal verbs (e.g., Blais & Gonnerman, 2013; Kim & Kim, 2012; Matlock & Heredia, 2002; Paulmann, Ghareeb-Ali, & Felser, 2015); and lexical bundles (e.g., Arnon & Snider, 2010; Bannard & Matthews, 2008; Bod, 2001; Hernández, Costa, & Arnon, 2016; Tremblay & Baayen, 2010; Tremblay, Derwing, Libben, & Westbury, 2011).

In this paper we set out to test how far the distributional properties of formulaic sequences can explain the processing advantage, regardless of specific features that may vary from type to type. In other words, is it simply that speakers register occurrences of frequent phrases, or are additional properties important in why formulaic sequences seem to be processed more efficiently than novel strings? Results from studies like those cited above for lexical bundles would suggest that frequency of occurrence, above all else, contributes to faster processing, hence any further distinctions may be fairly arbitrary and of little actual use. For example, Tabossi et al. (2009) showed that although idioms were judged to be meaningful more quickly than control phrases, the same was true for what they called clichés (entirely compositional frequently occurring phrases). They
argued that the *idiom superiority effect* is a property of any frequently occurring phrase, regardless of other aspects such as semantics. On the other hand, Jolsvai, McCauley, and Christiansen (2013) showed that the meaningfulness of a word sequence was an important factor in how it was processed in a phrasal decision task, over and above simply how frequently it occurs. Gyllstad and Wolter (2016) found that frequency but also transparency were important factors in how participants judged word combinations, with shorter reaction times for more transparent combinations (free combinations as opposed to restricted collocations).

In this paper we select three different types of formulaic sequence—idioms, binomials, and collocations—that all qualify as “formulaic” from the point of view of being recurrent phrases. Beyond this, they differ markedly in terms of specific properties that contribute to their formulaic status. For example, idioms are broadly (but variably) non-decomposable, and in all cases the meaning of the whole phrase must be retrieved directly from the lexicon to some degree. In contrast, binomials—sequences of $x$-and-$y$ where a specific word order is highly preferred, such as *salt and pepper*—can be literal or figurative and often constitute semantic associates, but are highly fixed in the sense that the reversed form is rarely if ever used. Collocations are very broadly defined as co-occurring word pairs, and here we define them as combinations of words that are entirely compositional and semantically “free,” but which co-occur in conventional and recurrent patterns. Crucially, based on the established body of evidence, we expect all three types of formulaic language (idioms, collocations, and binomials) to show a processing advantage relative to control phrases. Our first aim is to see how far the broad distributional property of phrase frequency can explain the overall differences, and whether this alone can account for the observed processing effects. We then go on to explore how aspects of predictability, and type-specific constraints, contribute to processing, over and above the effects of frequency. Note that throughout this paper, “predictability” refers to the expectancy for the final word of a formulaic sequence, once the initial word or words have been seen. Various factors, such as context, are likely to determine this, but we do not consider these in detail here.

1.1 Idioms

Idioms are among the most studied of all formulaic phrases, and have been described as “prototypically” formulaic (Siyanova-Chanturia & Martinez, 2014). Titone et al. (2015) suggest that this is because as a class, idioms can vary along all of the dimensions that are relevant to the study of formulaic language more generally, including frequency, familiarity, transparency, decomposability, literalness, etc. Importantly, as well as being clear examples of formulaic sequences, idioms are also often included in the class of figurative (non-literal) language, which may further have a bearing on how they are processed and understood. Early non-compositional models (e.g., Bobrow & Bell, 1973; Gibbs, 1980; Swinney & Cutler, 1979) suggested that idioms are highly lexicalized entries that can be retrieved directly, but subsequent work has shown that idioms are not simply “long words,” and do undergo compositional analysis (Cacciari & Tabossi, 1988). More recent “hybrid” models (e.g., Sprenger, Levelt, & Kempen, 2006; Titone & Connine, 1999) see idioms as both single entries and compositional wordstrings, and are supported by widespread evidence demonstrating that idioms show internal syntax (Cutting & Bock, 1997; Konopka & Bock, 2009; Peterson, Burgess, Dell, & Eberhard, 2001) and that the literal meanings of component words are activated as an obligatory part of processing (Hamblin & Gibbs, 1999; Holsinger & Kaiser, 2013; Smolka, Rabanus, & Rösler, 2007; Titone & Connine, 1994).

A robust finding is that idioms are generally recognized more quickly than matched “novel” phrases. For example, Swinney and Cutler (1979) used a phrasal decision task to show that idioms (e.g., *break the ice*) are judged to be meaningful phrases more quickly than control phrases (e.g.,
break the cup). Recent eye-tracking research has also supported the fast processing of idioms compared to control phrases (Carrol & Conklin, 2017; Carrol, Conklin, & Gyllstad, 2016), regardless of whether idioms are used in figurative or literal contexts (Siyanova-Chanturia, Conklin, & Schmitt, 2011). The literature suggests that such effects stem from the fact that idioms are highly familiar (Schweigert, 1986, 1991; Schweigert & Moates, 1988), and predictable (Fanari, Cacciari, & Tabossi, 2010; Libben & Titone, 2008), hence it seems likely that idioms are recognized quickly primarily because they are well-known phrases (van Lancker Sidtis, 2012).

However, the role of decomposability (how much the figurative meaning of an idiom can be mapped onto the literal meanings of the component words) remains open to debate. For example, the idiom decomposition hypothesis (Gibbs & Nayak, 1989; Gibbs, Nayak, & Cutting, 1989) proposed that only decomposable idioms should be processed quickly, since analysis of the phrase would be consistent with a literal reading. Results here have been mixed, with some studies showing support (Caillies & Butcher, 2007), and others showing the opposite pattern (Cieślicka, 2013; Titone & Libben, 2014) or no difference (Tabossi et al., 2008). Libben and Titone (2008) only found effects of decomposability on meaningfulness judgements for less familiar idioms, suggesting that familiarity may “trump” other aspects in how idioms are recognized, represented, and understood (c.f. Abel, 2003; Carrol, Littlemore, & Dowens, 2018).

For the purposes of the present study, it is useful to make a distinction between familiarity (which is a person-specific, subjective measure) and frequency, although these may often be highly correlated. Idioms as a class are widespread (Brenner, 2003, estimated that English contains over 10,000 idioms), but individual idioms occur relatively infrequently, at least based on corpus evidence (Moon, 1998). We therefore aim to assess how far frequency alone can explain the idiom superiority effect, before considering the additional contributions of specific properties such as decomposability and familiarity.

1.2 Binomials

Binomials have generated less interest in the literature than idioms, and represent an example where their formulaicity comes from an entirely different property. Here we define binomials as combinations of x-and-y where a reversal of the order is entirely possible, but where one word order is highly conventionalized. Examples are most often noun-and-noun (e.g., salt and pepper, king and queen), and a complex set of variables have been identified that determine the order (e.g., Benor & Levy, 2006; Cooper & Ross, 1975; Lohmann, 2012; Mollin, 2012; Morgan & Levy, 2016). These include conceptual factors (e.g., general before specific, animate before inanimate), cultural restrictions (e.g., power relations), and phonological variables (e.g., length and stress patterns of each word), among others. However, for each constraint there are also frequent exceptions (e.g., for “male before female”: man and wife, men and women but bride and groom), and the overriding factor seems to be conventionalization of one order over the other. Morgan and Levy (2016) suggest that for any example there is a trade-off between abstract knowledge of these constraints and direct experience.

In a recent study, Conklin and Carrol (2016) showed that frequency effects emerge rapidly for novel binomials in a natural reading task. For non-attested binomials (e.g., grass and leaves, plates and bowls, where there is no highly conventionalized order) a processing advantage (incrementally faster reading times) was observed in as few as three or four presentations, confirming the importance of phrasal frequency in how binomials become fused in a particular order. Siyanova-Chanturia et al. (2011) argued that frequency only accounts for some of the effects observed for attested binomials. They showed faster reading times for binomials (e.g., bride and groom) compared to reversed forms (groom and bride), but their analysis suggested that phrasal frequency explained
Carrol and Conklin

only part of the effect. They proposed that the configuration itself played a vital role, in that the preferred (binomial) form was privileged compared to the dispreferred (reversed) form, even when overall phrase frequency was accounted for.

An additional consideration with many binomials is that they often represent primary semantic associates \((\text{knife}-\text{fork}, \text{king}-\text{queen}, \text{salt}-\text{pepper})\), which may also contribute to how they are processed. That is, as well as frequency effects for the whole phrase, and direct representation of the configuration for the most common binomials, semantic priming between component words may contribute to faster processing. Carroll and Slowiaczek (1986) found within-sentence priming for semantically related words (e.g., \(\text{author}-\text{book}\)) when these appeared within the same clause. In a similar study, Camblin, Gordon, and Swaab (2007) found that association priming effects were only robust when the overall discourse context was impoverished or not cohesive. Given such results, it is likely that the semantic relatedness of binomial word pairs plays at least some role in how they are processed, although conventionalization is assumed to be the single biggest factor. Note that this does not necessarily equate to frequency, as a phrase may be fairly infrequent but have a highly fixed order, or may be very frequent but occur just as often in the reversed form. Both characteristics will be explored in this study.

1.3 Collocations

Collocations can be very broadly defined as any frequently co-occurring words, or, more accurately, words that co-occur more frequently than we might expect by chance (Biber, Johansson, Leech, Conrad, & Finegan, 1999). They can often be at least partially figurative, and various classifications have been proposed that consider collocations from a phraseological point of view (e.g., Howarth, 1998). From a frequency-based perspective (e.g., Sinclair, 1991), collocations are defined according to certain corpus-derived metrics, such as \(t\)-scores (a test of the null hypothesis that there is no connection between two words) or mutual information (MI), which measures the strength of co-occurrence between two words that form a collocation. Typically, an MI score of 3 is taken as the threshold above which a word pair can be considered to be of linguistic interest (Hunston, 2002).

Importantly, various types of collocations have been shown to demonstrate the same processing advantage as the other formulaic types considered so far. Wolter and Gyllstad (2011) showed that native speakers of English are faster and more accurate when responding to adjective + noun collocations (relative to non-formulaic baseline word combinations) in a primed lexical decision task. In a subsequent study, Wolter and Gyllstad (2013) showed facilitation for verb + noun combinations on a phrase-level grammaticality judgment task. Durrant and Doherty (2010) used a primed lexical decision task for low, mid, and high frequency collocations, as well as for high frequency collocations that were also semantic associates. In an unmasked priming task, they found priming for only high frequency word pairs (MI higher than 6), and this effect was observed for both associated and non-associated word pairs. In a second study using a masked prime, facilitation was only observed for those word pairs that were high frequency collocations and semantic associates. They suggested that, firstly, the threshold at which collocations become psychologically “real” may be much higher than that adopted in the corpus literature, and secondly, the mental representation of collocations may depend on both frequency of encounter and semantic association.

Sonbul (2015) used both offline measures (a rating task asking how typical a phrase is in English) and online measures (eye-tracking) to investigate responses to adjective + noun collocations. She compared synonymous word pairs of high frequency (e.g., \(\text{fatal mistake}\)), lower frequency (e.g., \(\text{awful mistake}\)), and no frequency (i.e., non-collocations, e.g., \(\text{extreme mistake}\)). There
was a clear effect of frequency in the offline task (higher frequency collocations were rated as more typical). In the eye-tracking task, early measures showed an effect of frequency (first pass reading time was shorter for higher frequency collocations), but this disappeared in later measures (overall reading time and fixation count). This suggests that while higher frequency collocations may be more easily recognized, they are not necessarily easier to process and integrate into context than lower frequency synonymous word pairs, hence different factors may be important at different stages of processing.

Finally, Vilkaite (2016) used eye-tracking to show that verb + noun collocations (e.g., provide information) were read more quickly by native speakers than control phrases (e.g., compare information), both in their canonical adjacent configurations, and when they were separated by three words (e.g., provide some of the information). Vilkaite interpreted these result as arguing against a “holistic” hypothesis in how collocations are processed. Instead, they support more general probability-based models (e.g., Barlow & Kemmer, 2000; Kuperberg & Jaeger, 2016). The results are also in line with Hoey’s theory of lexical priming (2005; also Pace-Sigge, 2013), and the notion of “concgrams” described by Cheng, Greaves, Sinclair, and Warren (2009), which are co-occurrences of words regardless of whether they are sequential or adjacent. Such approaches allow for much more flexibility in how multiword sequence are conceived.

Overall, there is clear evidence that frequently occurring collocations are processed quickly. Frequency alone may explain much of this, but other factors such as MI (a measure of strength of co-occurrence, rather than simply how frequent a collocation may be) are also important. Our study will enable us to explore how each of these contributes to overall processing patterns.

### 1.4 Comparing Formulaic Subtypes

Despite the widespread research into specific types of formulaic expressions, there remains a relative lack of work directly comparing subtypes. Columbus (2010) compared reading times for idioms, lexical bundles, and restricted collocations. All three types were read more quickly than non-formulaic controls, and idioms were processed the most quickly, overall. She concluded that these differences may not be the result of the different subtypes per se, but that different variables relevant to each type produce different effects. Columbus (2013) went on to show that both corpus data and human ratings can reliably distinguish between subtypes, using measures such as frequency, familiarity, and perceived transparency. How these factors influence online processing remains to be explored, although as noted earlier, Gyllstad and Wolter (2016) showed that both transparency and frequency affected reaction times in their phrasal decision task. That is, restricted collocations were judged more slowly than free combinations, but within each category more frequent phrases were judged more quickly. Other studies reviewed here have also shown that multiple factors may be at play in how formulaic language is processed, and aspects of conventionalization (of which frequency is a reflection) may only tell us part of the story.

Importantly, little of the work discussed thus far has involved natural reading, as most tasks required overt responses or judgements. We therefore aim to explicitly compare the contribution of a range of factors to understand how different kinds of formulaic sequence are processed in more natural contexts. Our overall research questions are:

1. How far do distributional variables (frequency and predictability) explain the processing advantage for different kinds of formulaic phrase?
2. How do different variables/constraints contribute to the processing of each type of phrase?
2 Experiment

2.1 Materials

Items were selected based on our definitions for each formulaic type introduced in the previous section, using a range of linguistic and distributional criteria. Frequency values were taken from the British National Corpus (BNC).

Idioms were selected from the *Oxford Learner’s Dictionary of Idioms* (Warren, 1994) and previously published idiom studies. They were all of the form Verb-X-Noun (e.g., *kick the bucket, push your luck*) or Preposition-determiner-Noun (*behind the scenes, below the belt*). All items occurred at a phrase frequency of at least 11 per 100 million words (mean = 54, SD = 53). To ensure that they were generally well known to native speakers, they were included in a rating task where native speaker participants (n = 21) rated each for how familiar it was on a seven-point scale. A final list of 45 items that scored highest for familiarity was created (mean = 5.9/7, SD = 0.7).

Binomials were of the form X-and-X, where both words were of the same lexical class. All phrases were from online lists and previously published studies. To trim this list down, we adopted a minimum phrase frequency of 20 per 100 million words (mean = 251, SD = 221), and a minimum ratio of 4:1 in favor of forward to reversed occurrences (i.e., a binomial must occur at least four times as often in the forward form than in the reversed form, mean = 9.3, SD = 0.7). All items were included in a rating study with native English speakers (n = 48) to assess how figurative/literal they were and to assess their reversibility (whether participants thought that the phrase could be reversed and the meaning retained). Figurative/literalness was assessed on a scale from 1 to 3, with 1 being entirely figurative, 3 being entirely literal, and 2 being potentially both (mean = 2.7/3, SD = 0.36). Reversibility was assessed on a scale from 1 to 7 (mean = 4.6/7, SD = 1.1). There was a high correlation here, with more figurative idioms being seen as less reversible (r = 0.64, p < 0.001). A final list of 45 items was created, including only binomials that were considered to be broadly literal and reversible. The majority were noun-and-noun (e.g., *salt and pepper, n = 31*), with some verb-and-verb (e.g., *pick and choose, n = 3*), adjective-and-adjective (e.g., *sick and tired, n = 10*), and preposition-and-preposition (e.g., *out and about, n = 1*).

For collocations we extracted a list of commonly co-occurring adjective-noun combinations from the BNC. We only considered non-idiomatic (non-figurative) examples, and all items occurred at least 10 times per 100 million words (mean = 110, SD = 133), with a minimum MI score of 3 (mean = 6.7, SD = 2.2) and a minimum t-score of 2 (mean = 9, SD = 4.8). Forty-five items were selected based on these criteria.

For all items, we created two control phrases: for control type 1 phrases, the first word was changed (e.g., *spill the beans became drop the beans*); and for control type 2, the second word was changed (e.g., *spill the beans became spill the chips*). This is a potentially important comparison as most studies have generally compared formulaic phrases to controls where the final word is changed (e.g., *break the ice vs. break the cup*, Swinney & Cutler, 1979). By comparing two types of control phrase we will be able to see whether there is a formulaic advantage compared to one, both, or neither. There is also a key difference between the two types of control phrase. For the comparison of formulaic to control type 1 phrases, we are comparing a more frequent phrase with a less frequent one. We are also comparing a phrase where seeing word 1 may generate a strong expectancy for what word should follow (formulaic condition) compared to a phrase where this is not the case. For control type 2 phrases, we are again comparing a more frequent with a less
frequent phrase, but additionally we are comparing phrases where the same expectation generated by word 1 is met (formulaic condition) or not met (control type 2 condition). A list of all formulaic and control items is provided in the Supplementary Materials.

Each control phrase was matched with its formulaic comparator for single word length and frequency. We then created short sentences for all items (see Table 1). These were designed to be as neutral as possible before the critical phrase was seen, to minimize any effects of context that might make a particular completion more or less predictable. The sentences were designed so that the formulaic phrase and both of its controls would make sense in the same context. The immediate post-context (the words following the critical phrase) was also the same to ensure that no contextual information could be extracted in the parafovea during processing of the final word of the phrase, then the final part of the sentence was tailored so that each version was completed in a coherent manner (note: in most cases the binomials and collocations fitted into the same context as the controls, so the same post-context was used for all three). We made sure that the number of words preceding (mean = 3.8, SD = 0.8) and following (mean = 11.9, SD = 2.1) the critical phrase was similar for all items. To ensure that all items (formulaic phrase and controls) were equally plausible sentences, we included all items in a rating study and asked native speakers (n = 25) to judge each sentence for how acceptable it was on a scale from 1 to 5. There were no differences between formulaic units and either control (formulaic, mean = 4.4, SD = 0.4; control type 1, mean = 4.4, SD = 0.4; control type 2, mean = 4.4, SD = 0.4; one-way ANOVA by condition: F = 0.13, p = 0.876).

We collected a range of other data for the experimental items (formulaic and control phrases). To make the frequency counts comparable between phrase types, we converted all raw values to the Zipf scale (van Heuven, Mandera, Keuleers, & Brysbaert, 2014). We then calculated two measures of predictability for the final word of each phrase. The first was a Cloze probability score for all formulaic and control items, which we considered to be a subjective measure of how predictable the phrase was in a short, neutral context. Phrases were presented as the first part of the sentence up to (but not including) the final word of the phrase, for example, *It was hard not to spill the...*, and we asked participants (native speakers of English, n = 69, with participants seeing one of four versions of the materials) to provide the first word that came to mind that could plausibly continue the sentence. It was stressed that these were sentence fragments, and that the word did not have to complete the sentence, simply continue it in a reasonable way. Cloze probability was calculated as the percentage of participants who provided the correct (or intended) completion in each case. We

| Table 1. Examples of context sentences for idioms and control phrases. |
|----------------|----------------|----------------|----------------|
| Variables     | Pre-context   | Phrase         | Post-context   |
| Idiom         | It was hard not to spill the beans                  | when I heard such a juicy piece of gossip. |
| Control type 1| It was hard not to drop the beans                   | when I burned my hand on the hot pan.   |
| Control type 2| It was hard not to spill the beans                   | when I stumbled on my way out of the kitchen. |
| Binomial      | I heard that the king and queen                      | will be visiting the city next week.    |
| Control type 1| I heard that the prince and queen                    | will be visiting the city next week.    |
| Control type 2| I heard that the king and prince                      | will be visiting the city next week.    |
| Collocation   | They were in abject poverty                          | but they seemed to make the best of their situation. |
| Control type 1| They were in total poverty                           | but they seemed to make the best of their situation. |
| Control type 2| They were in abject agony                            | but they seemed to make the best of their situation. |
also calculated a measure of transitional probability for each formulaic and control phrase. This is an objective measure of how likely it is that word 2 follows word 1 based on corpus frequencies. We follow the same formula used by McDonald and Shillcock (2003) and Frisson, Rayner, and Pickering (2005), although this was adapted to calculate the likelihood of the final word following the first word and the determiner or conjunction in idioms and binomials. Transitional probability was calculated as:

\[
\text{Overall Phrase Frequency} \div \text{Frequency of Word 1 ( + determiner or conjunction) } \times 100
\]

For example, \textit{spill the beans} (39) \div \textit{spill the} (93) \times 100 = 42\%. Finally, for all items we obtained semantic association scores between the two content words using the \textit{Edinburgh Associative Thesaurus} (Kiss et al., 1973). A summary of the properties of the items is presented in Table 2.

### 2.2 Participants

Thirty-six English native speaker undergraduate students took part in the experiment for course credit. They were randomly assigned to one of the three presentation lists.

### 2.3 Procedure

The experiment was administered on an Eyelink 1000+ eye-tracking system from SR Research. Stimuli were presented on a 1920 \times 1080 computer monitor (refresh rate = 60 Hz). Eye movements (left-eye, monocular recording) were monitored using a desk-mounted camera (sample rate = 500 Hz). After initial setup, a nine-point calibration and validation procedure was used to verify accuracy, and was repeated at regular intervals throughout the experiment.
Participants were asked to read each sentence for comprehension and to press the spacebar when they had finished. Each sentence was preceded by a fixation cross to allow for trial by trial drift checking. One-third of the sentences were followed by a simple yes/no comprehension question to ensure that participants paid attention throughout. Accuracy was high overall and comparable across all items (mean = 93%; scores ranged from 88% to 98% for individual sub-types/conditions). Participants saw a total of 180 sentences, including 45 filler sentences, with a short break and recalibration after every 60 items. Participants were randomly assigned to one of the three presentation lists (A, B, and C).

3 Results

Two participants were removed because of technical problems, leaving data from 34 subjects (12 from List A, 11 from Lists B and C). As List was not a significant factor in any of the subsequent analysis, data were collapsed across lists. All eye-tracking data was cleaned according to the default settings in the four-stage procedure within the Eyelink Data Viewer program. Here, very short fixations are first merged with neighboring fixations within a specified distance (first stage = fixations <40 ms and within 0.5 degrees; second stage = fixations <40 ms and within 1.25 degrees), then any instances of three consecutive fixations <140 ms are merged into one fixation. Finally, any remaining fixations <80 ms or >800 ms are removed entirely, as these are assumed to represent, respectively, minor location errors rather than true fixations, and momentary losses of concentration. Data were also visually inspected and any individual trials where data was clearly unusable because calibration was poor, track loss had occurred, or in which the whole phrase was skipped (received no fixations at all), were discounted. In total 4.6% of data were excluded based on these criteria, leaving 4379 data points for analysis.

We concentrate our analysis on both whole phrases and the final words (see Carrol & Conklin, 2014, for a rationale of this with formulaic sequences). Phrase-level effects might reflect the processing and integration of the phrase as a whole, while the final word is likely to be the locus of any formulaic advantage (Columbus, 2010). We consider a range of early and late measures in our analysis. Broadly speaking, early measures are thought to reflect immediate lexical processing during an initial parse of a sentence, while late measures reflect postlexical processes and integration of meaning into the overall sentence context (Altarriba, Kroll, Scholl, & Rayner, 1996; Conklin, Pellicer-Sánchez, & Carrol, 2018; Inhoff, 1984; Paterson, Liversedge, & Underwood, 1999; Staub & Rayner, 2007). For whole phrases and final words, we consider first pass reading time (early measure: the sum duration of all fixations on the phrase the first time it is encountered in the sentence, before gaze exits to the left or right) and total reading time (late measure: the sum of all fixations on the phrase over the course of the trial, including rereading time). For final words, we also considered likelihood of skipping (also an early measure: how likely is it that the final word receives no fixations at all during first pass reading?).

We constructed linear mixed effects models using the lme4 package (version 1.1-13, Bates et al., 2014) in R (version 3.5.1, R Development Core Team, 2018). Each eye-tracking measure was considered in its own model. We included random intercepts for subject and item, and adopted the maximal random effects structure warranted by the dataset (Barr, Levy, Scheepers, & Tily, 2013). All duration measures were log-transformed to reduce skewing. For analysis of likelihood of skipping we used a logistic linear model, and skipped items were discounted from any subsequent word-level durational analysis. In all models we included single word length and frequency (on the Zipf scale) for both content words as covariates to control for any word-level differences.
3.1 Omnibus Analysis of Distributional Variables

Table 3 reports the overall reading patterns (mean and SD) for each measure, for the data as a whole, and according to subtype. For each measure we constructed a model including fixed effects of phrase type and condition, and compared the difference between levels of condition (formulaic vs. control 1 and formulaic vs. control 2) for each phrase type using the “difflsmeans” function in the lmerTest package (version 3.0-1; Kuznetsova, Brockhoff, & Christensen, 2016) in R. Significant differences are indicated in Table 3, and the full output of this model is provided in the Supplementary Materials. Note that the data suggests that fixation durations in general are quite low, compared to averages reported in the wider literature. In fact, the average fixation duration during the experiment as a whole (considering all fixations made) was 197 ms (SD = 76), which is somewhat lower than the mean fixation duration of 225 ms for silent reading reported in Rayner (1998). Crucially, the data for the regions of interest are not markedly different than reading patterns for the sentences in general. In Table 3, whole-phrase data includes items where the final word was skipped (but where at least one fixation was made elsewhere in the phrase). The data for the final word reports the proportion of phrases where the final word was skipped entirely, then duration measures are reported for only those items that were not skipped.

Table 3 shows the expected formulaic advantage, relative to control phrases, although this is inconsistent across the three phrase types. For idioms, there is a clear advantage across all measures, and for phrase-level total reading time (RT), final-word skipping and final-word first pass RT for control type 2 phrases. For binomials, there are differences relative to control type 2 phrases across all measures, and relative to control type 1 phrases for all measures except phrase-level first pass RT. For collocations, there are significant differences relative to

<table>
<thead>
<tr>
<th>Variables</th>
<th>Whole phrase</th>
<th>Final word</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>First pass RT</td>
<td>Total RT</td>
</tr>
<tr>
<td>All types</td>
<td>335 (164)</td>
<td>428 (227)</td>
</tr>
<tr>
<td>Control 1</td>
<td>355 (180)*</td>
<td>490 (285)***</td>
</tr>
<tr>
<td>Control 2</td>
<td>357 (176)*</td>
<td>485 (261)***</td>
</tr>
<tr>
<td>Idioms</td>
<td>351 (149)</td>
<td>441 (215)</td>
</tr>
<tr>
<td>Control 1</td>
<td>392 (186)**</td>
<td>539 (299)***</td>
</tr>
<tr>
<td>Control 2</td>
<td>371 (173)</td>
<td>491 (230)***</td>
</tr>
<tr>
<td>Binomials</td>
<td>340 (180)</td>
<td>430 (234)</td>
</tr>
<tr>
<td>Control 1</td>
<td>364 (183)</td>
<td>484 (289)*</td>
</tr>
<tr>
<td>Control 2</td>
<td>359 (172)*</td>
<td>473 (257)***</td>
</tr>
<tr>
<td>Collocations</td>
<td>315 (161)</td>
<td>413 (231)</td>
</tr>
<tr>
<td>Control 1</td>
<td>310 (162)</td>
<td>445 (257)</td>
</tr>
<tr>
<td>Control 2</td>
<td>339 (182)</td>
<td>491 (294)***</td>
</tr>
</tbody>
</table>

Duration measures are reported in milliseconds; skipping is reported as a probability; duration values for the final word exclude zero values for skipped items. Values in parentheses are SDs. Significant differences between formulaic and control conditions (based on the model reported in the Supplementary Materials – Table S1) are indicated with the convention of *p < 0.05, **p < 0.01, ***p < 0.001.

RT: reading time; SD: standard deviation.
control type 2 phrases for all measures except phrase-level first pass RT, but compared to control type 1 phrases, only final-word skipping showed an advantage.

The different measures reflect different processes, and the one where all three subtypes showed an advantage compared to both types of control was final-word skipping. Skipped words are assumed to have been at least partially recognized and processed in the parafovea, hence words that are part of a known sequence may be more likely to be skipped entirely, or may subsequently receive shorter fixations (as seen for idioms and binomials relative to both types of control phrase, and collocations relative to control type 2 phrase). Consistent effects were also seen for phrase-level total RT, which is a later measure that may reflect overall effort required to integrate the meaning of the phrase into the sentence. This suggests that the overall meaning of the formulaic phrases was easier to understand, while the non-formulaic controls require relatively more consideration (even when the meaning of the formulaic phrase was entirely literal, as in the case of binomials and collocations).

To address our first research question, we added phrase frequency and predictability into the model for each measure, to see firstly whether this improved the fit, and secondly whether it accounted for the formulaic advantage. In other words, were there still between-condition differences once these were included? For all duration measures, phrase frequency made a significant improvement: phrase-level first pass RT: $\chi^2 = 18.55, p < 0.001$ and total RT: $\chi^2 = 2.95, p < 0.001$; final-word first pass RT: $\chi^2 = 15.79, p < 0.001$ and total RT: $\chi^2 = 17.29, p < 0.001$. There was no further improvement for any model by adding phrase frequency as an interaction rather than a fixed effect. In all cases, higher phrase frequency led to shorter durations (all $t$-values $> 2.0$, all $p$-values $< 0.05$). However, phrase frequency did not significantly improve the model for likelihood of skipping as a fixed effect ($\chi^2 = 5.80, p = 0.446$) or as an interaction with type and condition.

We next considered whether predictability adds anything, over and above the effects of phrase frequency. As might be expected, phrase frequency, Cloze probability, and transitional probability are highly correlated (all $r$-values $> 0.50$; all $p$-values $< 0.001$), suggesting that broadly they reflect similar properties. We ran a Principal Component Analysis that confirmed this: All three predictors loaded onto the first component in a similar way, and this accounted for 72% of the variance in these variables. Cloze probability and transitional probability loaded onto the second component (accounting for 15% of the variance) in the same way, while phrase frequency operated in a different direction. Based on this, and to avoid issues of collinearity, we removed transitional probability from any further analysis and instead included only Cloze probability as a measure of predictability.

We added Cloze probability to the models including phrase frequency to determine whether this made any further improvement. This made no improvement to the model for phrase-level first pass RT ($\chi^2 = 0.71, p = 0.397$) or skipping rates ($\chi^2 = 1.10, p = 0.295$), but did for phrase-level total RT ($\chi^2 = 11.86, p < 0.001$), final-word first pass RT ($\chi^2 = 4.12, p = 0.042$) and final-word total RT ($\chi^2 = 8.80, p = 0.003$). Table 4 reports the differences between condition once phrase frequency, and phrase frequency and Cloze probability, are included in the models. With the exception of final-word skipping for idioms, inclusion of these two variables eliminates the formulaic advantage for all subtypes on all measures.

### 3.2 Individual Constraints for Each Subtype

We next considered each phrase type separately, to consider how type-specific constraints may influence the processing of the different subtypes of formulaic language. For example, MI is a measure of the strength of a collocation, and therefore is only relevant to this type of phrase. Similarly, idiom-specific variables such as decomposability are not relevant to binomials or collocations. We included type-specific variables and the main variables of condition, phrase frequency, and Cloze probability.
For each subtype we again constructed separate models for each eye-tracking measure. We began with a model including a fixed effect of condition, and word-level length and frequency for both content words as covariates. We then added in each of phrase frequency, Cloze probability, and the type-specific predictors outlined below individually, then used log-likelihood tests to see whether any of these improved the model, either as fixed effects or interactions with condition. Next, we constructed cumulative models using forward model selection, where if any variable made a significant improvement this was retained, then subsequent variables were added to this (as fixed effects and interactions with condition), and the models were compared using log-likelihood tests. This gave us an indication of the combination of predictors that exerted an effect for each subtype. Finally, we constructed a maximal model for each measure, where all variables were included as fixed effects and interactions with condition. We compared this to the best fitting cumulative model to ensure that the pattern of significant variables was the same and found no notable differences in terms of significant effects. Below, we report first the results of the individual models, then the results of the cumulative models, for each subtype and each eye-tracking measure. We include coefficient values from the models to give an indication of the size of any significant effects. (Note: coefficients relate to log-RTs for duration measures, and log-odds for skipping rates). At the end of the section an overall summary of predictor variables for each subtype (based on the best-fitting cumulative model) is presented in Table 5.

### 3.3 Idioms

For idioms, as well as phrase frequency and Cloze probability, we also included familiarity and decomposability to assess their effect. Familiarity was rated by a set of native speakers of English ($n = 21$) as indicated in the Experiment section. Mean rating was 5.9 ($SD = 0.7$, range = 3.5–7).
Decomposability was rated by a separate set of native speakers \((n = 19)\), who were asked to judge how much the component words contributed to the figurative meaning that was provided for them (e.g., If you make peace with someone you bury the hatchet). Mean rating was 4.1/7 \((SD = 1.3, \text{range } 1.8–6.2)\).

At the phrase level, the model for first pass RT was improved by the addition of fixed effects for phrase frequency \(\chi^2 = 16.43, p < 0.001\) and decomposability \(\chi^2 = 6.30, p = 0.012\). In a cumulative model, both remained significant, hence higher phrase frequency \((\beta = -0.11, t = -4.01, p < 0.001)\) led to shorter reading times and higher decomposability \((\beta = 0.03, t = 2.44, p = 0.019)\) led to longer reading times. In this model, differences between idioms and control type 1 \((t = -0.22, p = 0.827)\) and control type 2 \((t = -1.75, p = 0.082)\) were not significant. For total RT, the addition of phrase frequency \(\chi^2 = 16.29, p < 0.001\) and an interaction between condition and familiarity \(\chi^2 = 10.89, p = 0.012\) improved the initial model. In the cumulative model, this meant that more frequent phrases were read more quickly \((\beta = -0.11, t = -3.68, p < 0.001)\). The interaction between condition and familiarity for control type 2 phrases \((t = 2.80, p = 0.005)\) meant that for control type 2 phrases only, greater familiarity with the corresponding idiom led to marginally longer RTs \((\beta = 0.08, t = -1.89, p = 0.061)\), while there was no effect on idioms or control type 1 phrases. In this model, there was a marginal difference between idioms and control type 1 phrases \((t = 1.90, p = 0.059)\) but no difference compared to control type 2 phrases \((t = 0.33, p = 0.745)\).

For final-word skipping, the initial model was improved by the addition of an interaction between condition and familiarity \(\chi^2 = 13.05, p = 0.005\) and condition and decomposability \(\chi^2 = 10.17, p = 0.017\), but neither phrase frequency nor Cloze probability made any improvement. In the cumulative model, the best fit included the interaction of condition and familiarity, and a fixed effect of decomposability (addition of this as an interaction with condition made no further improvement once familiarity was also included in the model). This meant that skipping rates were higher for more familiar idioms \((\beta = 0.46, z = 2.69, p = 0.007)\), while higher decomposability led to less skipping for all phrases \((\beta = -0.17, z = -2.56, p = 0.010)\). In this model there were significant differences between conditions for idioms compared to control type 1 phrases \((z = -4.61, p < 0.001)\) and compared to control type 2 phrases \((z = -4.12, p < 0.001)\).

For final-word reading times, phrase frequency \(\chi^2 = 10.17, p = 0.017\) and decomposability marginally \(\chi^2 = 5.74, p = 0.057\) improved the model for first pass RT as fixed effects. In the cumulative model, once phrase frequency was included, decomposability made no further improvement \(\chi^2 = 0.45, p = 0.501\). Here, higher phrase frequency led to shorter reading times \((\beta = -0.05, t = -2.30, p = 0.022)\) and there were no differences between idioms and control type 1 \((t = 0.96, p = 0.338)\) or control type 2 \((t = 0.10, p = 0.922)\) phrases. For total RT, only phrase frequency improved the initial model as a fixed effect \(\chi^2 = 8.54, p = 0.003\). This meant that higher frequency led to shorter reading \((\beta = -0.08, t = -2.92, p = 0.004)\), and there were no differences between idioms and control type 1 \((t = 0.25, p = 0.806)\) or control type 2 \((t = -0.56, p = 0.578)\) phrases.

### 3.4 Binomials

For binomials, as well as the main variables, we included semantic association strength and the ratio of forward to backward occurrences in the BNC.

At the phrase level, the model for first pass RT was improved by the addition of fixed effects for each of phrase frequency \(\chi^2 = 4.02, p = 0.045\), association strength \(\chi^2 = 7.68, p = 0.006\), and marginally Cloze probability \(\chi^2 = 3.05, p = 0.081\) and ratio \(\chi^2 = 3.19, p = 0.074\). In a cumulative model, however, the inclusion of association strength removed any effects of phrase frequency or Cloze probability. The best fitting model included a fixed effect of semantic association...
(β = −0.02, t = −2.78, p = 0.006) and a marginal effect of ratio (β = −0.08, t = −1.77, p = 0.077), where both variables contributed to shorter reading times. In this model, no between-condition differences were observed for binomials compared to control type 1 phrases (t = −0.76, p = 0.448) or control type 2 phrases (t = −0.82, p = 0.414). For total RT, the initial model was improved by the addition of fixed effects for phrase frequency (χ² = 9.20, p = 0.002) and Cloze probability (χ² = 7.91, p = 0.005), and an interaction of condition and association strength (χ² = 14.23, p = 0.003). In a cumulative model, combinations of both phrase frequency and Cloze, and phrase frequency and association strength were significant, but only phrase frequency remained as a significant predictor when all three were included in the model. These models suggested that phrase frequency was always facilitative (led to lower overall RTs), and Cloze was also facilitative, but only when semantic association strength was not included. When association strength was included, the effects on binomials and control type 1 phrases were non-significant, but there was a significant facilitative effect for control type 2 phrases, whereby more strongly associated phrases had overall less reading time. Comparison of the two models (phrase frequency + Cloze and phrase frequency + association) showed that they were very similar in terms of their fit, suggesting that Cloze and association strength may be in part reflecting a similar property for binomials. When all three variables were included, the difference between binomials and control type 1 phrases disappeared (t = −0.92, p = 0.358) but there remained a significant difference compared to control type 2 phrases (t = −2.83, p = 0.005).

For final-word skipping, addition of an interaction with condition made an improvement to the initial model for association strength (χ² = 12.44, p = 0.006), and marginally for Cloze probability (χ² = 7.39, p = 0.060). Ratio also made a marginal improvement as a fixed effect (χ² = 2.91, p = 0.088). The cumulative model showed that once an interaction with association strength was included, no other variables made any further improvement. In this model, a higher level of association between the component words led to higher rates of skipping in binomials (β = 0.13, z = 2.57, p = 0.010), and there were differences between binomials and control type 2 phrases (z = 2.01, p = 0.043) but not control type 1 (z = −1.18, p = 0.237).

For final-word reading, the initial model for first pass RT was improved by the addition of a fixed effect for Cloze probability only (χ² = 9.22, p = 0.002). In a cumulative model, no other variable further improved this, so the final model showed a facilitative effect of Cloze probability for all phrases (β = −0.01, t = −3.03, p = 0.003). In this model, there were no differences between binomials and either control type 1 (t = −0.45, p = 0.653) or control type 2 phrases (t = 0.48, p = 0.635). For total RT, the initial model was improved by the addition of fixed effects for phrase frequency (χ² = 34.21, p < 0.001) and Cloze probability (χ² = 38.22, p < 0.001), and by the addition of interactions with condition for association strength (χ² = 38.44, p < 0.001) and ratio (χ² = 35.68, p < 0.001). The cumulative model suggested that a combination of factors were important here. Once phrase frequency was included, Cloze probability made an additional improvement as a fixed effect (χ² = 7.32, p = 0.007), and association strength made a marginal improvement as an interaction with condition (χ² = 6.75, p = 0.080). In the model including Cloze probability, phrase frequency made a marginal improvement as a fixed effect (χ² = 3.31, p = 0.069) and ratio made a marginal improvement as an interaction with condition (χ² = 7.64, p = 0.054). Association strength made no contribution once other factors were included. The final model included fixed effects of phrase frequency (β = −0.08, t = −2.09 p = 0.038) and Cloze probability (β = −0.02, t = −2.65, p = 0.008), where both led to shorter overall RTs, and a marginal interaction of condition and ratio, whereby for control type 1 phrases (but not binomials or control type 2) a higher ratio (therefore a more strongly conventionalized order) for the corresponding binomial led to longer reading times (β = 0.16, t = 2.67, p = 0.008). In this model, there remained overall marginal differences between binomials and control type 1 (t = −1.83, p = 0.068) and control type 2 phrases (t = −1.94, p = 0.052).
3.5 Collocations

For collocations, we included semantic association strength and MI score as predictors in all models. At the phrase level, none of the predictors made any improvement to the model for first pass RT, either on their own or in combination with other variables. In the model including only condition and the word-level covariates, there were marginal differences between formulaic and control type 2 phrases (t = 1.75, p = 0.083) but not control type 1 phrases (t = -0.80, p = 0.430). The model for total RT was improved by the addition of Cloze probability (X² = 15.19, p < 0.001), MI (X² = 9.84, p = 0.002), and semantic association (X² = 9.24, p = 0.002) as fixed effects. A cumulative model included fixed effects of both Cloze probability (β = -0.03, t = -3.49, p < 0.001) and MI (β = -0.02, t = -2.63, p = 0.009). The addition of association strength made no further improvement to this, but the addition of an interaction between condition and phrase frequency did make a marginal improvement (X² = 7.11, p = 0.068). Here, phrase frequency did not have an effect on collocations themselves (t = 0.48, p = 0.631), but was inhibitory for both control type 1 (t = 2.37, p = 0.018) and control type 2 phrases (t = 2.11, p = 0.035). In this model, there was a significant difference between collocations and control type 1 phrases (t = -1.97, p = 0.049) but not control type 2 phrases (t = -1.60, p = 0.111).

Table 5. Summary of predictor variables for each subtype, where + indicates a facilitative effect and - indicates an inhibitory effect.

<table>
<thead>
<tr>
<th>Idioms</th>
<th>Variables</th>
<th>Advantage</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Phrase Zipf</td>
<td>Cloze</td>
</tr>
<tr>
<td>Phrase first pass RT</td>
<td>+</td>
<td>-</td>
</tr>
<tr>
<td>Phrase total RT</td>
<td>+</td>
<td>-</td>
</tr>
<tr>
<td>Word skip</td>
<td>+</td>
<td>-</td>
</tr>
<tr>
<td>Word first pass RT</td>
<td>+</td>
<td>-</td>
</tr>
<tr>
<td>Word total RT</td>
<td>+</td>
<td>-</td>
</tr>
</tbody>
</table>

Binomials

| Phrase first pass RT | Phrase Zipf | Cloze | Semantic association | Ratio | n.s. | n.s. |
| Word first pass RT   | +           | +     | +                   | n.s.  | **   |
| Word total RT        | +           | +     | #                   | #     |      |

Collocations

| Phrase first pass RT | Phrase Zipf | Cloze | Semantic association | MI   | n.s. | #    |
| Word first pass RT   | (+)         | +     | (+)                 | (+)  | n.s. | n.s. |
| Word total RT        | +           | +     | n.s.                | n.s. |      |      |

Effects in parentheses are significant only as individual predictors and not in a cumulative model. Advantage columns indicate whether the formulaic advantage (relative to each control type) remains once all significant variables are included in the model (not significant (n.s.), otherwise p-values are reported as #p < 0.10; *p < 0.05; **p < 0.01; ***p < 0.001). RT: Reading time; Decomp: decomposability; MI: mutual information.
For final-word skipping, no variable made an improvement to the initial model on its own, either as a fixed effect or interaction term, and no combination of predictors made any improvement in a cumulative model. This meant that in the basic model including only condition and word-level covariates, the second word of collocations was skipped more often than in control type 2 phrases ($z = -2.98, p = 0.003$) and marginally in control type 1 phrases ($z = -1.74, p = 0.081$).

For final words that were not skipped, the model for first pass RT was improved by the addition of both phrase frequency ($X^2 = 4.11, p = 0.042$) and MI ($X^2 = 6.42, p = 0.011$) as fixed effects. The cumulative model suggested that addition of both variables caused a confound, as neither was significant when they were both included, hence we looked at separate models for phrase frequency and MI. Phrase frequency was not improved by the addition of either Cloze probability or association strength, and in the final model led to shorter first pass RTs ($\beta = -0.47, t = -0.01, p = 0.636$), with no differences between collocations and control type 1 ($t = -0.47, p = 0.642$) or control type 2 phrases ($t = -0.47, p = 0.639$). The model for MI similarly included a facilitative fixed effect ($\beta = -0.01, t = -2.54, p = 0.011$), with no differences between collocations and control type 1 ($t = -0.47, p = 0.639$) or control type 2 phrases ($t = -0.52, p = 0.604$). The model for total RT was improved by the addition of fixed effects for each of Cloze probability ($X^2 = 5.31, p = 0.021$), MI ($X^2 = 5.29, p = 0.021$), and association strength ($X^2 = 4.19, p = 0.040$). A cumulative model showed that once Cloze and MI were included, association strength made no further improvement, hence the final model included fixed effects of Cloze probability ($\beta = -0.02, t = -1.96, p = 0.051$) and MI ($\beta = -0.02, t = -1.98, p = 0.048$), with no differences between collocations and control type 1 ($t = -1.24, p = 0.216$) or control type 2 phrases ($t = -0.11, p = 0.914$). When Cloze probability was not included, both MI ($\beta = -0.02, t = -2.14, p = 0.033$) and marginally association strength ($\beta = -0.06, t = -1.58, p = 0.065$) had a facilitative effect on all phrases.

Table 5 summarizes the constraints that are relevant for each of the formulaic subtypes. We indicate whether each variable has a facilitative effect (speeds up processing/leads to shorter reading) or inhibitory effect (slows down processing/leads to longer reading times). Effects in parentheses are only significant when other predictors are not included. We also indicate which between-condition differences remain when significant predictors were included in a cumulative model.

As Table 5 shows, the advantage for idioms was generally explained by phrase frequency, and when this was included the between-condition differences disappeared for all duration measures. Increased frequency led to shorter first pass RT (initial recognition of the sequence) and total RT (overall integration of the meaning). Decomposability, which is often seen as an important factor in how idioms are processed, had a limited role whereby less decomposable (i.e., more opaque) idioms were read more quickly during the first pass, and were more likely to have the final word skipped. Familiarity (but not frequency) also made an important contribution to skipping rates, but Cloze probability was not a significant predictor for any measure, either on its own or when phrase frequency was included. The only measure where the included variables did not account for the formulaic advantage was skipping rates, where, in the best fitting model, the differences between idioms and both control conditions remained.

For binomials, a wider spread of variables was implicated. Phrase frequency was important for later measures (phrase-level and final-word total RT). Cloze probability contributed to final-word reading patterns on early and late measures, and was important at the phrase level only when semantic association was not accounted for. Similarly, semantic association contributed to phrase-level reading times (for total RT, and only when Cloze was not included), and was the only variable that contributed to final-word skipping. Ratio, which might be seen as a measure of fixedness (in terms of how much more often the binomial is seen compared to its reversed form), contributed to facilitated reading only during initial recognition of the whole phrase.
For collocations, on two measures our variables failed to explain the formulaic advantage at all, although for first pass reading time this advantage was not strongly apparent in our data. For final-word skipping there was a clear advantage compared to control type 2 phrases (and a marginal advantage compared to control type 1) that was not explained by any of our variables. Both Cloze probability and MI contributed to shorter overall reading times, suggesting that these might be a good indicator of how easily a collocation can be integrated into the overall sentence, rather than simply how quickly it is recognized. Phrase frequency had a minimal effect (and when it did, it seemed to reflect the same property as MI), and semantic association was also minimal in its contribution.

4 Discussion

We compared reading patterns for three distinct types of formulaic sequence (idioms, binomials, collocations), in three conditions (conventional formulaic form, control phrase where the first content word was changed, control phrase where the second content word was changed). We found evidence that for all three types there was an advantage for the formulaic phrases compared to both types of control phrase, across a range of eye-tracking measures. Based on the previous research indicating a processing advantage for formulaic language in general, this was expected. Our primary questions were: How far can phrase frequency and predictability alone account for this advantage? And what contribution do type-specific variables make for each subtype?

4.1 Distributional Variables: Omnibus Analysis

The omnibus analysis suggested that both frequency and predictability jointly explain why formulaic expressions are processed (recognized during initial reading and then integrated into the surrounding context) more quickly than non-formulaic control phrases. Put simply, formulaic language is processed quickly largely because it is known, hence phrases that have been encountered more often (as measured by their overall frequency) are assumed to be more strongly encoded in the lexicon, independently of their component words. Note that what this means could cover a multitude of things, from a truly “holistic” entry for some phrases (such as non-decomposable idioms), to something more akin to a lexical priming mechanism whereby links between co-occurring words become strengthened through experience (e.g., Wray, 2012; Hoey, 2005). Our results suggest that broad distributional properties do a fairly good job of explaining the formulaic processing advantage, and when these are accounted for, differences between formulaic and control conditions largely disappear.

If these findings seem straightforward, it is worth remembering that some formulaic phrases—and in particular idioms—are actually not particularly frequent, at least compared to other sequences. In our study, the control phrase “see the film” had a higher phrase frequency (84) than the majority of idioms (only seven idioms had a higher frequency), hence frequency alone does not equate to formulaicity. Predictability clearly contributes as well, and adding this to the analysis improved the models for three of the five measures we included here. We considered two measures of predictability—Cloze probability and transitional probability—but our initial analysis suggested that as these are highly correlated, they are likely to reflect very similar properties (Frisson, Rayner, & Pickering, 2005; Janssen & Barber, 2012). Where these may differ is in the sensitivity to context: Cloze probability may vary according to the strength of bias provided by a preceding context, whereas transitional probability will not. Taken together, we can consider frequency and both measures of predictability to reflect the
overall experience with each phrase, and it is clear that in broad terms, this does a good job of explaining how and why formulaic language is processed in the efficient way that it is.

The question of interest then becomes: To what extent do additional, phrase-specific factors such as those we have considered here represent something over and above experience-based effects of frequency and predictability? If so, are such features unique to formulaic language, or simply manifestations of features of the language processing system that are brought into focus by the particular subtypes we have looked at? We consider these questions for each of the subtypes and their respective variables in turn.

4.2 Idioms

Despite being relatively infrequent (at least compared to the binomials and collocations in our study), idioms were the subgroup where phrase frequency had the most consistent effect. On all duration measures, phrase frequency had a main effect, and when this was included in the analysis any between-condition differences all but disappeared. There were additionally effects of decomposability, whereby greater decomposability was inhibitory (led to longer reading times and lower likelihood of skipping the final word). As we addressed in the literature review, the role of decomposability has been variable in previous studies: in terms of activation of idiom meaning, some find that greater decomposability leads to faster activation than for less decomposable phrases (e.g., Caillies & Butcher, 2007), whereas others find that greater decomposability interferes with the activation of figurative meaning (e.g., Cieślicka, 2013; Titone & Libben, 2014). One proposal is that all idioms are to some degree represented as unitary entries, at least in terms of the meaning of the phrase as a whole (c.f. Caillies & Declerq, 2011; Titone & Libben, 2014), and other aspects such as decomposability or literal plausibility dictate the extent to which the literal meaning interferes with retrieval of the figurative. In this sense, “knowing” an idiom is the key driver of how it will be processed (like all fixed phrases, e.g., Tabossi et al., 2009), and we highlighted in the Introduction that a crucial fact about idioms is that they are both formulaic and figurative. If an idiom is not known (i.e., has never been encountered before), aspects such as transparency or decomposability (along with context) will be essential in determining whether the meaning can be inferred. However, once an idiom is known, these properties may serve only to modulate the ease or difficulty with which the figurative meaning is selected, relative to the competing activation of the literal meaning. Importantly, these do not drive or override the overall advantage for idioms compared to non-formulaic phrases, since this is based primarily on the recognition of a known combination of words.

The key difference between frequency and familiarity was demonstrated in our results only in terms of final-word skipping, where greater familiarity led to a higher rate of skipping (but did not entirely account for the between-condition differences). Our results are in line with studies such as Carrol and Conklin (2017), where native speakers skipped the final words of idioms 31% of the time compared to 9% for control phrases. In eye-tracking research, word skipping is known to be affected by both visual and linguistic factors. Very short words are often skipped (Rayner & McConkie, 1976), as are function words (Carpenter & Just, 1983) and very high frequency content words (Rayner, Sereno, & Raney, 1996). Of relevance here, words that are highly predictable are also skipped more often (Drieghe, Rayner, & Pollatsek, 2005; Rayner, Slattery, Drieghe, & Liversedge, 2011; Rayner & Well, 1996). Cloze probability in itself did not contribute to skipping, but we could posit that this is subsumed within the variable of familiarity, in that once an idiom is well known, it is highly predictable. In turn, for all
idioms, but especially very familiar ones (and note that all of the idioms in this study were selected to ensure that they were generally familiar), recognition of a “known” configuration may increase the chances that the final word is skipped entirely, although the negative effect of decomposability reinforces the degree to which other variables might interfere with this.

4.3 Binomials

Binomials represent a very different case to idioms, since successful comprehension was always a simple case of combining the component words, rather than recognizing an additional meaning at the level of the whole phrase. The items we used are highly frequent (on average much more frequent than idioms) and are in theory entirely reversible (in its non-idiomatic sense, black and white has the same propositional meaning as white and black; note: all of the items in our study were non-idiomatic). As we summarized previously, a range of linguistic factors have been proposed for how this order is determined, but convention/experience seems to be the most important in processing terms (Morgan & Levy, 2016). Our results confirmed this, with overall phrase frequency and Cloze probability accounting for most of the phrase- and word-level reading patterns. Ratio, which we might interpret as an index of how conventionalized or fixed a word order is, was important only for initial recognition of the phrase as a whole (first pass RT at the phrase level). Siyanova-Chanturia, Conklin, and van Heuven (2011) found a consistent advantage for how binomials were read, and their analysis suggested that something over and above phrase frequency contributed to this. In a follow-up study using EEG, Siyanova-Chanturia, Conklin, Caffarra, Kaan, and van Heuven (2017) found evidence to suggest that the configuration itself was an important part of processing, in that some element of pattern recognition as well as semantic expectancy was involved in how they were processed. Since conventionalization might be seen as the result of very high levels of exposure (that is, repeated encounters lead to the formation of something akin to a “template” for very high frequency items), these results all point to the same conclusion, whereby frequency plays a very significant role in how binomials are assimilated into the lexicon, and how they are subsequently processed.

The effect of semantic association may represent something additional to these distributional characteristics. Binomials are very often also primary semantic associates, and low-level semantic priming between words in natural reading has been demonstrated previously (e.g., Carroll & Slowiaczek, 1986; Camblin et al., 2007), in particular in the early stages of processing. We saw similar effects, whereby stronger association between component words led to shorter first pass RTs for the phrase, and made skipping of the final word more likely. These between-word priming effects are a part of language processing more generally (reflecting well-established properties such as spreading activation, e.g., Collins & Loftus, 1975), with automatic semantic priming thought to be driven by a combination of properties, such as feature overlap and association strength (e.g., Hutchison, 2003; Lucas, 2000). In binomials, these properties may serve to reinforce the one-way relationship between words, although this is driven primarily by having encountered the word combination multiple times in one particular configuration (which may be determined in the first place by a range of linguistic constraints discussed in the Introduction). As with idioms, for the most frequent (and fixed) examples, something akin to a “lexical entry” may eventually form, or at least connections between words may become so automatized that the end result (highly speeded, activation of the second coordinate word) is the same.
4.4 Collocations

The collocations in the current study showed some evidence of a formulaic advantage, but even in the initial omnibus analysis this was less marked and consistent than for idioms or binomials. Phrase frequency did not do a good job of explaining these effects in the individual analysis, and was largely superseded by MI, which is a measure not simply of occurrence but of co-occurrence for two given words. MI and phrase frequency were highly correlated, and hence may be reflecting the same broad patterns. However, it may be that MI is both more nuanced, and more specific than phrase frequency. To take an example, strong X may be a combination that occurs highly frequently, but where the noun slot can be filled with several plausible options (many of which may in themselves be frequent, such as strong tea, strong coffee, strong feeling, etc.). In contrast, abject X is more restricted in that only a small number of words (poverty, failure) are likely candidates. Although the overall frequency may be low, the likelihood of co-occurrence is high, hence MI reflects a more nuanced knowledge of language experience than the coarser measure of phrase frequency. In line with this, Ellis, Simpson-Vlach, and Maynard (2008) found that for lexical bundles derived from academic corpora, MI rather than phrase frequency determined speed of processing for native speakers, while language learners were more sensitive to overall frequency of occurrence. Our results suggest that Cloze probability and MI (both reflective of the expectation created by seeing the first word) are both better explanatory variables for collocations than phrase frequency, although ultimately all are derived from experience and therefore support the broad conclusion that distributional factors are the primary drivers of collocational processing. Interestingly, on two measures (first pass RT for the phrase and likelihood of skipping the second word altogether), none of the variables considered here explained the differences between collocations and control phrases (although these were marginal). Vilkaite (2016) also found no specific effects of phrase frequency or MI in her analysis of verb-noun collocations (which did show an overall advantage), and suggested that both variables were subsumed within the overall status of collocation.

We found little evidence that semantic association affected the processing of collocations, in contrast to, for example, Durrant and Doherty (2010), who found a clear difference for collocations that were also semantic associates. Hutchison (2003) concluded that there is evidence of pure associative priming in the absence of any semantic overlap, but that the strongest effects of priming are often seen when both criteria are met (semantic overlap and association). In our stimuli, semantic association was in general low (much lower than for the binomials), and was routinely close to zero for both sets of control items, so it may be that we did not have enough variability here to see any effects. As above, any evidence of semantic priming for associated words would be consistent with language processing more generally, and would not necessarily reflect any aspect of the formulaic nature of the items themselves.

4.5 Overall Conclusions

Our data support the view that formulaic expressions, regardless of fundamental differences in the properties that constitute categories such as idioms, binomials, and collocations, are all processed quickly, primarily because they are known phrases that have been encountered multiple times as part of the language experience of native speakers. Note that “processed” here refers to two aspects: both the recognition of a “known” combination of words, and the analysis and integration of that sequence into the surrounding sentence context. Both aspects seem to be easier for formulaic expressions than literal expressions, even when the task of deriving propositional meaning is ostensibly straightforward (as in the case of binomials and collocations). This frequency of past occurrence may, in the
most frequent or most fixed examples, lead to the formation of something akin to a “template,” and evidence from the EEG literature supports this for a range of different types of expression: idioms (Vespignani, Canal, Molinaro, Fonda, & Cacciari, 2011; Zhang, Yang, Gu, & Ji, 2013); figurative collocations (Molinaro & Carreiras, 2010); and binomials (Siyanova-Chanturia et al., 2017).

As Siyanova-Chanturia and Martinez (2014) and Siyanova-Chanturia (2015) argue, much of the evidence that claims to show “holistic” processing in formulaic sequences doesn’t actually speak directly to this claim, but instead simply shows a consistent speed advantage for formulaic phrases. A better way to conceive of formulaic sequences might be as distributed representations at a lexical level, with multiple connections both between words and with other levels (e.g., the lexical conceptual level postulated by the superlemma theory—Sprenger et al., 2006; see also the construction-integration account of idioms described in Caillies & Butcher, 2007). Experience and frequency of past encounter are the primary drivers here, as with language processing in general, and this view is not incompatible with a lexical priming account (Hoey, 2005; Pace-Sigge, 2013), whereby all examples of a word combination either serve to reinforce an existing link, or dilute it.

Our data suggest that distributional characteristics do account for most of the formulaic advantage, while other aspects serve to modulate the ease or difficulty with which the phrase as a whole might be interpreted. For example, semantic links provide low-level facilitation where these exist, and since binomials happen to be primary associates more often than not, these present an example of a more general phenomenon. Similarly, if idioms are not familiar, there can be no formulaic advantage, and variables such as decomposability serve to make the process of working out the figurative meaning more or less straightforward (e.g., Carrol et al., 2018). When idioms are well-known, they are recognized quickly and easily, and decomposability has little effect on the most familiar examples (Libben & Titone, 2008). There are obvious linguistic differences between disparate subtypes (e.g., the difference between idioms and lexical bundles), but within the broad class of formulaic language, aspects of conventionalization (frequency of past occurrence, and predictability of the sequence based on co-occurrence probabilities) are the main driver of the faster processing reported in the literature. Specific aspects of different phrase “types” (e.g., idioms, which are inherently ambiguous) serve to underpin processing in various subtle ways, which do not differ markedly from how language is processed more generally.

Funding
This research received no specific grant from any funding agency in the public, commercial, or not-for-profit sectors.

Supplemental material
Supplemental material for this article is available online.

Notes
1. We exclude what have been called irreversible binomials (e.g., Arcara et al., 2012)—phrases such as hit and run where the order is not only iconic and logical, but also where the phrase itself has a meaning over and above the constituent parts, therefore effectively operating as an idiom.
2. Many of the studies on collocations discussed here also investigate non-native processing, but we concentrate here only on results for native speakers as these are more relevant to our study.

ORCID iD
Gareth Carrol  [https://orcid.org/0000-0002-1886-4881]
References


