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# INTRODUCING ICBe: AN EVENT EXTRACTION DATASET FROM NARRATIVES ABOUT INTERNATIONAL CRISES

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A PREPRINT

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## Abstract

How do international crises unfold? We conceptualize international relations as a strategic chess game between adversaries and develop a systematic way to measure pieces, moves, and gambits accurately and consistently over a hundred years of history. We introduce a new ontology and dataset of international events called ICBe based on a very high-quality corpus of narratives from the International Crisis Behavior (ICB) Project. We demonstrate that ICBe has higher coverage, recall, and precision than existing state of the art datasets and conduct two detailed case studies of the Cuban Missile Crisis (1962) and the Crimea-Donbas Crisis (2014). We further introduce two new event visualizations (event iconography and crisis maps), an automated benchmark for measuring event recall using natural language processing (synthetic narratives), and an ontology reconstruction task for objectively measuring event precision. We make the data, supplementary appendix, replication material, and visualizations of every historical episode available at a companion website [crisisevents.org](http://crisisevents.org).

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If we could record every international interaction in the realms of diplomacy, conflict, economics, and beyond, how much unique information would this chronicle amount to, and how surprised would we be to see something new? In other words, what is the entropy of international relations? While this record could in principle be unbounded, the central conceit of social science is that there are structural regularities that limit what actors can do, their best options, and even which actors are likely to survive (Brecher 1999; Reiter 2015). If so, then these events can be recorded and systematically measured by social scientists interested in these regularities.<sup>2</sup> Thanks to improvements in natural language processing, more open-ended efforts have begun to capture entire unstructured streams of events from international news reports.<sup>3</sup> This has invited fruitful efforts to evaluate the coverage, quality, and accuracy of attempts to measure international affairs.

We advance existing efforts to identify and structure regularized events and actor in international politics by combining human coding with natural language processing to develop (1) a large, flexible ontology of international affairs and (2) a fine-grained and structured event dataset of international crises from 1918-2017 developed by applying our ontology to an unusually high-quality corpus of historical narratives of international crises (Brecher 1999; Wilkenfeld and Brecher 2000; Brecher et al. 2021). We then develop several methods for objectively gauging how well these event codings reconstruct the information contained in the original crisis narrative. We conclude by benchmarking our event codings against several current state of the art event data collection efforts. The underlying fine-grained variation in international affairs is unrecognizable through the lens of current quantification efforts. We find that existing models produce data on historical episodes that do not contain enough information to reconstruct the underlying event. In focusing this initial effort on international crises as a proof of concept sample, we demonstrate our ontology and method's potential to improve upon existing empirical identifications of patterns of international interactions.

In the proceeding four sections, this measurement paper makes the following arguments. First, there is a real-world unobserved latent concept known as international relations that can and should be systematically measured. Second, we propose a method for systematic large-scale measurement of the actors and behaviors in international affairs and as a proof of concept apply that method to a well-regarded and salient sample of events known as international crises. Third, in doing so we confirm that those measurements exhibit several desirable kinds of internal and external validity and out-perform existing approaches. Fourth, this validation can be evaluated in detail via new event visualizations, with examples provided for case studies of the 1962 Cuban Missile Crisis and 2014 Crimea-Donbas crisis. A final section concludes.

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<sup>2</sup>See work on crises (Brecher and Wilkenfeld 1982; Beardsley et al. 2020), militarized disputes (Palmer et al. 2021; Gibler 2018), wars (Sarkees and Wayman 2010; Reiter, Stam, and Horowitz 2016), organized violence (Ralph Sundberg and Mihai Croicu 2016; Davies, Pettersson, and Öberg 2022), political violence (Raleigh et al. 2010), sanctions (Felbermayr et al. 2020), and international agreements (Kinne 2020; Owsiaik, Cuttner, and Buck 2018), dispute resolution (Frederick, Hensel, and Macaulay 2017), and diplomacy (Moyer, Turner, and Meisel 2020; Sechser 2011).

<sup>3</sup>See Beieler et al. (2016); Boschee et al. (2015); Brandt et al. (2018); Grant et al. (2017); Li et al. (2021). On event extraction from images and social media see Zhang and Pan (2019) and Steinert-Threlkeld (2019).

# 1 Identifying and measuring international relations

## 1.1 Motivation

Our knowledge of any historical episode, including the participants and their preferences, behaviors, and beliefs are only indirectly observed from historical records that most often take the form of unstructured natural language text. Despite its complexity, all international interactions fundamentally involve a finite set of actors expressing their interests through at least theoretically observable behaviors. So how can scholars abstract and measure discrete events about a historical episode in international relations? Scholars have recently sought to produce a structured account of world events by combining the unstructured corpus of historical records with informative priors about international relations. The result can produce novel insight from a systematic account of political events. The easiest way to convey the desired product is with an example. Figure 1 shows a narrative account of the Cuban Missile Crisis (1962) in natural language sentences alongside a mapping to discrete machine-readable abstractive events. From this, scholars can identify similarities and differences across events like what foreign policy actions deter versus inflame (Jervis 1978; Glaser 2000), when third parties mediate (Haffar 2002; Quinn et al. 2006), and how actors communicate resolve (Trager 2016; Lupton 2018). Identifying patterns of international interactions is not just an inherently interesting enterprise; it is a necessary precondition to important efforts to predict where policymakers should turn their attention to improve global welfare (Ward et al. 2013; Beger, Morgan, and Ward 2021).

## 1.2 Existing state of the art measurements

We draw informative prior beliefs about the underlying process of international relations that we expect to govern behavior during historical episodes and their conversion to the historical record. We organize our prior beliefs along two overarching axes: existing efforts to identify the actors/actions of international relations and identifying a corpus that can be used to produce an ontology of the information we hope to recover. Table 1 describes these two axes as columns and rows, respectively.

The rows in Table 1 represent the types of information we expect to find in international relations and forms the basis for our proposed ontology. We created this ontology by performing of the corpus and identifying named entities and verbs mentioned in the text. To identify possible behaviors, we matched verbs to the most likely definition found in Wordnet (Miller 1995), tallied (SI Appendix 1.2), and then aggregated them into a smaller number of behaviors balancing conceptual detail with manageable sparsity for human coding, informed by existing conceptual (“Literature” column) and measurement research (discussed below). We used the ICB project actor level data to identify likely actors for each crisis and location options relative to each actor. For behavior, actor, and location, coders could write-in a value if the given options were insufficient. The codebook lists 11 behaviors added post-coding as coders flagged events that were not captured by the initial ontology (e.g. propaganda).

Figure 1: Comparison of a natural language and machine-readable abstractive account of the Cuban Missile Crisis (1962). The text on the left is a summary of the event from the ICB Crisis Narrative. The mapping on the right shows the corresponding ICBe coding

As we are not the first to attempt to measure international relations in a structured manner, the columns of Table 1 compare the ontological coverage of ICBe to existing state of the art systems in production and with global coverage. We choose these datasets and models as they represent frequently used and reputable efforts to structure and describe historical events of interest to scholars of international politics. The first column starts with our contribution, ICBe, alongside other event-level datasets including CAMEO dictionary lookup-based systems (Historical Phoenix (Althaus et al. 2019); ICEWS (Boschee et al. 2015); Terrier (Grant et al. 2017)), the Militarized Interstate Disputes Incidents dataset, the UCDP-GED dataset (Davies, Pettersson, and Öberg 2022; Sundberg and Melander 2013), and ACLED (Raleigh et al. 2010).<sup>4</sup> The final set of columns compares episode-level datasets beginning with the original ICB project (Brecher et al. 2021; Brecher and Wilkenfeld 1982; Beardsley et al. 2020); the Militarized Interstate Disputes dataset (Palmer et al. 2021; Gibler 2018), and the Correlates of War (Sarkees and Wayman 2010). We include episode-level datasets as they remain a common and trusted tool for analyzing international relations, and because ICBe is unique among event-level datasets as events are matched to crises and so can be aggregated to the episode level. There is imperfect overlap concerning their intended depth and scope of coverage; ‘international crises’ are similar, but not identical to, ‘interstate wars’ and ‘militarized interstate disputes’ which differ yet again from ‘individual events of organized violence’ and ‘non-violent action’. Even like-concepts require care in comparison, as an ‘aim’ in ICBe is the same as in MIPS, but an ‘alert’ in ICBe is not the same as an ‘alert’ in MID.

This comparison is not intended to fault existing data and models for not including every variable in ICBe’s ontology, as some of these variables fall outside the scope of a particular dataset’s intended purpose. Rather, it serves as an initial basis for identifying the heterogeneity in existing efforts to abstract and measure discrete historical events of interest and to provide theoretical justifications from existing research about what is included in our dataset’s ontology and where ICBe’s detail about historical events can be compared to the current state of the art.

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<sup>4</sup>Other related datasets that insufficiently overlap ICBe’s domain for comparison include BCOW (Leng and Singer 1988), WEIS (McClelland 1978), CREON (Hermann 1984), CASCON (Bloomfield and Moulton 1989), SHERFACS (Sherman 2000), Real-Time Phoenix (Brandt et al. 2018), and COFEE (Balali, Asadpour, and Jafari 2021) (see histories in Merritt (1994) and Schrodt and Hall (2006)).

Table 1: Ontological coverage of ICBe versus existing State of the Art

	Concept	Literature	Ev	Ev	Ev	Ev	Ev	Ev	Ep	Ep	Ep	COW
			1918	1945	1977	1995	1993	2015	1918	1816	1816	MIDs
Domain	Type (Episode or Event)											ICB
	Start		1918	1945	1977	1995	1993	2015	1918	1816	1816	
	End		2017	2019	2018	2020	2010	2022	2017	2014	2007	
	N		32K	8.5M	28.4M	17.5M	9.6K	128K	1M	1K	5.9K	1K
	Coders (Hand or Automated)		H	A	A	A	H	H	H	H	H	
	Corpus		ICB	News	News	News	Mix	News	Mix	Mix	Mix	
	Date source (Event or Article)		E	A	A	A	E	A	E	E	E	
	Location source (Event or Actor)		E	E	E	E	A	E	E	A	E	A
	States	Fazal (2011), Spruyt (1996)	✓	✓	✓	✓	✓	✓	✓	✓	✓	
	Subnational Actors	Haffar (2002)	✓	✓	✓	✓		✓	✓			✓
Players	IGO/NGO	Bush and Hadden (2019)	✓	✓	✓	✓		✓		✓		✓
	Civilians	Ben-Yehuda and MishaliRam (2006)	✓	✓	✓	✓		✓	✓			
	Fatalities	Lacina (2006), McNabb Cochran and Long (2017)	✓				✓	✓	✓	✓	✓	
	Force Size	Carafano (2014), Goertz and Diehl (1986)		✓								
	Force Domain	Gartzke and Lindsay (2019), Lindsay and Gartzke (2020)		✓	✓	✓	✓					
	Geography (location, territorial change)	Carter (2010)		✓				✓				
	Alert (Start/End Crisis)	Brecher and Wilkenfeld (1997)	✓							✓		
	Wishes (Desire/Fear)	Goldgeier and Tetlock (2001)	✓							✓		
	Evaluation (Victory/Defeat)	Stein and Russett (1980)	✓							✓		
	Aims (Territory, Policy, Regime, Preemption)	Sullivan (2007)										
Think	Awareness (Discover, Become Convinced)	Ramsay (2017), Yarhi-Milo (2013)	✓									
	React to past event (Praise, Disapprove, Accept, Reject, Accuse)	O'Neill (2018)	✓	✓	✓	✓				✓		
	Request future event (Appeal, Demand)	Zartman and Faure (2005)	✓	✓	✓	✓	✓					
	Predict future event (Promise, Threaten, Express Intent, Offer Without Condition)	Sechser (2011)	✓	✓	✓	✓	✓	✓			✓	
	Predict with condition (Offer, Ultimatum)	R. Powell (2002)	✓									
	Government (Leadership/Institution Change, Coup, Assassination)	Goemans, Gleditsch, and Chiozza (2009), J. M. Powell and Thyne (2011)	✓	✓	✓	✓			✓			
	By Civilians (Protest/Riot/Strike)	Chenoweth, Hendrix, and Hunter (2019)	✓	✓	✓	✓			✓			
	Against Civilians (Terrorism, Domestic Rights, Mass Killing, Evacuate)	Eck and Hultman (2007), LaFree and Dugan (2007)	✓	✓	✓	✓			✓			
	Diplomacy (Discussion, Meeting, Mediation, Break off negotiations, Withdraw/Expel Diplomats, Propoganda)	Beardsley (2011)	✓	✓	✓	✓			✓			
	Legal Agreements (Sign Agreement, Settle Dispute, Join War on Behalf of, Ally, Mutual Defense Pact, Open Border, Cede Territory, Allow Inspections, Political Succession, Leave Alliance, Terminate Treaty)	Gibler and Sarkees (2004), Owsiaik, Cuttner, and Buck (2018)	✓	✓	✓	✓			✓	✓	✓	
Say	Violate Agreement (Violate Terms of Agreement)	Leeds (2003)	✓									

Table 1: Ontological coverage of ICBe versus existing State of the Art

	Concept	Literature	ACLED UCDP-GED MID Incidents ICEWs	Terrier Phoenix ICBe	COW MIDs ICB
Do - Unarmed	Mutual Cooperation (Economic cooperation or Aid, Military Cooperation, Intelligence Cooperation, Unspecified)	Leeds (1999)	✓ ✓ ✓ ✓		
	Directed Aid (General Political Support, Economic Aid, Humanitarian Aid, Military Aid, Intelligence Aid, Unspecified Aid)	Yarhi-Milo, Lanoszka, and Cooper (2016)	✓ ✓ ✓ ✓		✓
Do - Armed	Preparation (Alert, Mobilization, Fortify, Exercise, Weapons Test)	Lai (2004)	✓ ✓ ✓ ✓ ✓ ✓		✓
	Maneuver (Deployment, Show of Force, Blockade, No Fly Zone, Border Violation)	Allen, Flynn, and Martinez Machain (2021)	✓ ✓ ✓ ✓ ✓ ✓		✓
	Combat (Battle/Clash, Attack, Invasion/Occupation, Bombard, Cease Fire, Retreat)	Fortna (2018), Min (2021)	✓ ✓ ✓ ✓ ✓ ✓	✓ ✓ ✓	
	Strategic (Declare War, Join War, Continue Fighting, Surrender, End War, Withdraw from War, Switch Sides)	Sarkees and Wayman (2010), Reiter, Stam, and Horowitz (2016)	✓ ✓ ✓ ✓ ✓ ✓		✓
	Autonomy (Assert Political Control Over, Assert Autonomy Against, Annex, Reduce Control Over, Decolonize)	Frederick, Hensel, and Macaulay (2017)	✓ ✓ ✓ ✓ ✓ ✓	✓	

With the exception of large-scale CAMEO dictionary-based systems (the first grouping of columns), our ontology improves upon the the existing state of the art quantitative datasets that ignore important information content about international interactions.<sup>5</sup> We highlight two particular innovations. First, we separate the ‘chess pieces’ from the ‘chess players’ in distinguishing between different actors within a state. By virtue of our ontology coding military versus civilian actors and national leaders versus bureaucrats, our data can be used to explore important questions concerning civilian-military relations (Narang and Talmadge 2017), Track Two diplomacy and the role of sub-national actors (Hsu et al. 2020), and the evolution of what actors are engaged in crises - a topic of increasing interest as states engage in gray zone conflict by employing the coast guard or paramilitary mercenaries instead of internationally recognized state militaries (Gannon et al. 2022). Second, we add information about the domains in which actors behave - whether in land, air, sea, space, or cyber - since they differ in their technology, tactics, geography, and purpose (Gartzke and Lindsay 2019). Doing so allows researchers to identify and explain patterns in escalation conditional on the military means states use in conflict. Recent concerns about cross-domain conflict and the effect of new domains of conflict like space and cyber have made this an endeavor of increased interest to practitioners. (Gannon 2022)

## 2 Methodology and data

### 2.1 Corpus

For our corpus, we select a set of unusually high-quality historical narratives from the International Crisis Behavior (ICB) project ( $n = 471$ ) with coverage spanning 1918-2017 (Supplementary Information (SI) Appendix 1.1)(Brecher et al. 2021; Brecher and Wilkenfeld 1997). ICB defines a crisis as meeting three conditions: (1) an actor perceives a threat to one of more of its basic values, (2) the actor has a finite time horizon for responding to the perceived threat, and (3) the probability of military hostility has increased (Brecher and Wilkenfeld 1982). Crises are a significant focus of detailed single case studies and case comparisons because they provide an opportunity to examine behaviors in international relations short of, or at least prior to, full conflict (Holsti 1965; Paige 1968; Allison and Zelikow 1971; Gavin 2014; Brecher and Wilkenfeld 1982; Iakhnis and James 2019). The corpus is also unique in being designed to be used in a downstream quantitative coding project, meaning each narrative was written by consensus by a small number of scholars using a uniform coding scheme where things like word choice, writing style, and level of specificity were deliberately done in a consistent manner (Hewitt 2001). Case selection was exhaustive based on a survey of world news archives and region experts, cross-checked against other databases of war and conflict, and non-English sources (Kang and Lin 2019; Brecher et al. 2021, 59).

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<sup>5</sup>See Balali, Asadpour, and Jafari (2021) for a recent review of ontological depth and availability of Gold Standard example text.

## 2.2 Coding Process

The ICBe ontology follows a hierarchical design philosophy where a smaller number of significant decisions are made early on and then progressively refined into more specific details (Brust and Denzler 2020).<sup>6</sup> Each coder was instructed to first thoroughly read the full crisis narrative and then presented with a custom graphical user interface (GUI) (SI Appendix 2.1). Coders then proceeded sentence by sentence, choosing the number of events (0-3) that occurred, the highest behavior (thought, speech, or action), a set of players, whether the means were primarily armed or unarmed, whether there was an increase or decrease in aggression (uncooperative/escalating or cooperative/de-escalating), and finally one or more specific and non-mutually exclusive activities. Some additional details were always collected (e.g. location and timing) while other details were only collected if appropriate (e.g. force size, fatalities, domains, units). While each event was matched to a sentence, coders could fill in details outside that sentence (e.g. antecedents to pronouns). We reviewed, standardized, and normalized where coders listed a behavior, actor, or location outside the ontology.<sup>7</sup>

A unique feature of the ontology is that thought, speech, and do behaviors can be nested into combinations, e.g. an offer for the U.S.S.R. to remove missiles from Cuba in exchange for the U.S. removing missiles from Turkey. Through compounding, the ontology can capture what players were said to have known, learned, or said about other specific fully described actions. Two other novel features of our data – its re-programmable coding and its inclusion of thought, speech, and action – intersect to provide a unique contribution.

First, re-programmable coding means scholars can make modifications to the ontology we provide and re-run the code script to produce a modified dataset. For example, if thoughts and speech acts are immaterial for a particular research question, those higher-level categories can be modified in a way that does not produce inevitable downstream effects on the rest of the coding. This is necessary because unlike existing datasets where variable values are largely independent, the nested nature of the ICBe variables means filtering out speech acts also requires filtering out variables about the actors, locations, and timing of those speech acts.

Second, no existing event data distinguishes thoughts, speeches, and actions. In fact, most only try to code actions and entirely omit thoughts and speech acts despite recognition of their importance in international politics (Smith 1998). Scholars have opted against coding thoughts and speech acts because of a lack of confidence the full universe could be readily observed and consequently at least theoretically be included.<sup>8</sup> But the perfect should not be the enemy of the good, and measurement challenges are only overcome after an initial attempt to estimate difficult-to-observe concepts of interest. The ICB narratives are one of the better sources for this endeavor due to the consistent use of high-quality primary source material that takes

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<sup>6</sup>This process quickly focuses the coder on a smaller number of relevant options while also allowing them to apply multiple tags if the sentence explicitly includes more than one or there is insufficient evidence to choose only one tag. The guided coding process also allows for the possibility that earlier coarse decisions have less error than later fine-grained decisions.

<sup>7</sup>See the full codebook on Github Repository ICBEEventData.

<sup>8</sup>Even the coding of overt actions like MIDs is not without contention (Gibler 2018).

advantage of qualitative methods well-suited to identifying thoughts and speech acts like archival work and expert interviews. Important questions about potential bias remain like whether a thought was known at the time by those involved or only recorded long afterwards by historians, but data validation required there to be data to be validated. By design, existing datasets largely omit thoughts and speech acts and also treat the resulting dataset as the static product of hardwired coding decisions. In allowing scholars to identify actions, thoughts, and speech acts as distinct characteristics of international crises while also granting flexibility about the inclusion of a non-random sample of novel variables, our approach is a notable improvement over the state of the art.

Each crisis was typically assigned to 2 expert coders and 2 novice coders with an additional tie-breaking expert coder assigned to sentences with high disagreement.<sup>9</sup> For the purposes of measuring intercoder agreement and consensus, we temporarily disaggregate the unit of analysis to the Coder-Crisis-Sentence-Tag ( $n=993,731$ ), where a tag is any unique piece of information a coder can associate with a sentence such as an actor, date, behavior, etc. We then aggregate those tags into final events ( $n=18,783$ ), using a consensus procedure (SI Appendix 2.2) that requires a tag to have been chosen by at least one expert coder and either a majority of expert or novice coders. This screens noisy tags that no expert considered possible but leverages novice knowledge to tie-break between equally plausible tags chosen by experts. Requiring sentence-tag matching may underestimate agreement but minimizes the inclusion of noise and allows for additional validation. Once filtered for agreement, we find 472 actors and 119 different behaviors: 12 thought, 13 speech, and 94 actions.

### 3 Performance comparison

#### 3.1 Internal consistency

We evaluate the internal validity of the coding process in several ways. For every tag applied we calculate the observed intercoder agreement as the percent of other coders who also applied that same tag (SI Appendix 2.3). Across all concepts, the Top 1 Tag Agreement was low among novices (31%), moderate for experts (65%), and high (73%) following the consensus screening procedure.

We attribute the remaining disagreement primarily to three sources. First, we required coders to rate and justify their confidence in the coding. They reported low confidence for 20% of sentences; 45% of those were due to a mismatch between the ontology and the text (“survey doesn’t fit event”) and 46% were from a lack of information or confused writing in the source text (40% “more knowledge needed”, 6% “confusing sentence”). Observed disagreement varied predictably with self-reported confidence (SI Appendix 2.4). Second, as intended, agreement is higher (75-80%) for questions with fewer options near the root of the ontology compared to agreement for questions near the leaves of the ontology (50%-60%). Third, individual coders

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<sup>9</sup>Expert coders were graduate students or postgraduates who collaboratively developed the ontology and documentation for the codebook. Undergraduate coders were students who engaged in classroom workshops.

exhibit nontrivial coding styles, e.g. some more expressive coders applied many tags per concept while others focused on only the single best match. We further observed unintended synonymity, e.g. the same information can be framed as either a threat to do something or a promise not to do something.

### 3.2 Improvement over existing efforts

To evaluate our coding process relative to existing datasets, we measure the recall and precision of ICBe events in absolute terms and relative to other existing systems. Recall measures the share of desired information recovered by a sequence of coded events while precision measures the degree to which a sequence of events correctly and usefully describes the information in history. To aid in subjective evaluation of the precision and recall of ICBe for each event, we provide full ICB narratives, ICBe coding in an easy-to-read iconographic form, and a wide range of visualizations for every case on the companion website.

Recall for historical episodes is poorly defined for two reasons. History may or may not be written by the victors but by virtue of being written by *someone* there is no genuine ground truth about what occurred, only surviving texts about it (Turberville 1933). Second, there is no *a priori* guide to what information is necessary detail and what is ignorable trivia. History suffers from what is known as the Coastline Paradox (Mandelbrot 1983) — it has a fractal dimension greater than one such that the more you zoom in, the more detail you will find about individual events as well as in between any two discrete events. The ICBe ontology is a proposal about what information is important, but we need an independent benchmark to evaluate whether that proposal is a good one and that allows for comparing proposals from event projects that had different goals. We need a yardstick for history.

Our strategy for dealing with both problems is a plausibly objective yardstick called a synthetic historical narrative. We collect a large diverse corpus of narratives spanning timelines, encyclopedia entries, journal articles, news reports, websites, and government documents. Using natural language processing (fully described in SI Appendix 3.1), we identify details that appear across multiple accounts. A detail refers to the smallest textual unit (a word or n-gram) for which we can calculate similarity across corpora to identify whether sentences semantically refer to the same broader observed event (Narayan, Cohen, and Lapata 2018). The more accounts that mention a detail, the more central it is to understanding the true historical episode. The theoretical motivation is that authors face word limits which force them to pick and choose which details to include, and they choose details that serve the specific context of the document they are producing. With a sufficiently large and diverse corpus of documents, we can vary the context while holding the overall episode constant and see which details tend to be invariant to context. Sufficiently similar details were binned together and then summarized so they could be compared to the coding in ICBe. This presents a harder evaluation baseline than comparing ICBe’s recall to just that of ICB since there are non-crisis aspects of these events that may be included in other narratives but are out of the scope of our data. For example, the nationalization of businesses in Cuba may be included as important context in the Cuban Missile Crisis

in documents that do not focus on the crisis dimensions like ICB. Using this hard case, a recall measure of ICBe on the synthetic narratives thus serves as a way to evaluate the breadth of ICBe's ontology and potential application to non-crisis international events.

We find substantive variation in recall across existing state of the art methods. Mentions of a detail across accounts are exponentially distributed with context-invariant details appearing dozens to hundreds of times more than context-dependent details.<sup>10</sup> Furthermore, crisis start and stop dates are arbitrary, and the historical record points to many precursor events as necessary detail for understanding later events. Figure 2 compares ICBe's recall with that of existing datasets for the two case studies detailed in Section 4. ICBe strictly dominates all of the systems but ICEWs in recall though we note that the small sample sizes mean these systems should be considered statistically indistinguishable. Across all existing datasets and ICBe, recall increases with the number of document mentions which is an important sign of validity for both them and our benchmark. The one outlier is Phoenix which in the Cuban Missile Crisis case is so noisy that its recall curve is flat to decreasing as mentions increase. The two episode-level datasets (MIDs and ICM) have low coverage of contextual details. The two other dictionary systems ICEWs and Terrier have higher coverage, with ICEWs outperforming Terrier. Importantly our corpus of ICB narratives has high recall of frequently mentioned details giving us confidence in how those summaries were constructed, and ICBe lags only slightly behind showing that it left little additional information on the table.<sup>11</sup>

The second component of event measurement validation is precision. It does little good to recall a historical event but too vaguely (e.g. MIDs describes the Cuban Missile Crisis as a blockade, a show of force, and a stalemate) or with too much error to be useful for downstream applications (e.g. ICEWS records 263 “Detonate Nuclear Weapons” events between 1995-2019). ICBe's ontology and coding system are designed to strike a balance so that the most important information is recovered accurately but also abstracted to a level that is still useful and interpretable.

How does ICBe's precision compare to the existing state of the art? A researcher should be able to lay out the events of a crisis on a timeline and read off the macrostructure of an episode from each individual move. We call this visualization a crisis map, a directed graph intersected with a timeline. A crisis map using ICBe for the Cuban Missile Crisis case study is provided in Figure 2, and crisis maps for the two case studies using existing event datasets can be found in SI Appendix 4.3 and 4.4 and crisis maps for all crises using all datasets can be found on the companion website. The crisis maps reveal the episode-level datasets like

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<sup>10</sup>As the ICB narratives are intended to explain conflictual behavior in a political context, many of the missing events concern more economic components of conflict (eg. nationalizing a foreign business). Even when they occur in the context of a crisis, these events largely fall outside the sample of information on which ICBe's ontology is currently trained. Even with this limitation, ICBe is more comprehensive than the existing datasets that do try to code the economic dimensions of these crises. We see expanding the ontology to broader international phenomenon as a promising future implementation of our model.

<sup>11</sup>Although Figure 2 focuses only on two crises, the synthetic narrative approach and recall comparison can, and should, be more broadly applied to all international crises in a way that could reveal systematic blindspots across datasets.

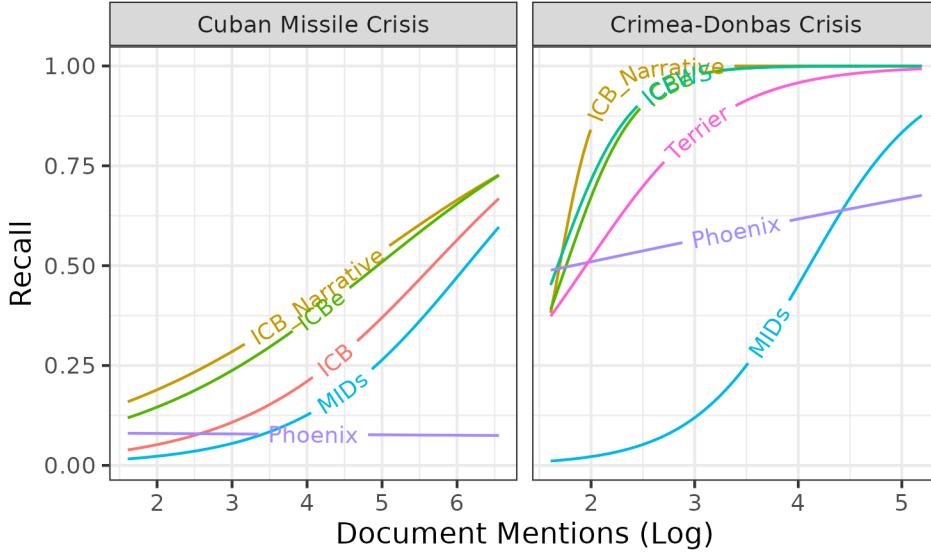


Figure 2: Recall comparison of two cases across existing state of the art efforts. Higher y-axis values represent higher recall and higher x-axis values represent number of times that detail is mentioned across the full corpus used to construct the synthetic narrative.

MIDs or the original ICB are too sparse and vague to reconstruct the structure of the crisis (SI Appendix 4.3 and 4.4). On the other end of the spectrum, the high recall dictionary-based event datasets like Terrier and ICEWs produce so many noisy events (several hundred thousand) that even with heavy filtering their crisis maps are completely unintelligible. Further, because of copyright issues, none of these datasets directly provide the original text spans making event-level precision difficult to verify.

We further want to verify individual event codings, which we can do in the case of ICBe because each event is mapped to a specific span of text. We develop the iconography system for presenting event codings as coherent statements that can be compared side by side to the original source narrative for every case on the companion website. We further provide a stratified sample of event codings alongside their source text (SI Appendix 4.2). We find both the visualizations of macrostructure and head-to-head comparisons of ICBe codings to the raw text to strongly support the quality of ICBe.

Figure 3 shows the location of every sentence from the ICBe corpus in semantic space as embedded using the same large language model as before, and the median location of each ICBe event tag applied to those sentences.<sup>12</sup> Labels reflect the individual leaves of the ontology and colors reflect the higher level coarse branch nodes of the ontology. If ICBe has high precision, substantively similar tags ought to have been applied to substantively similar source text, which is what we see both in two dimensions in the main plot and via hierarchical clustering on all dimensions in the dendrogram along the right-hand side.<sup>13</sup>

<sup>12</sup>We preprocess sentences to replace named entities with a generic Entity token.

<sup>13</sup>Hierarchical clustering on cosine similarity and with Ward’s method.

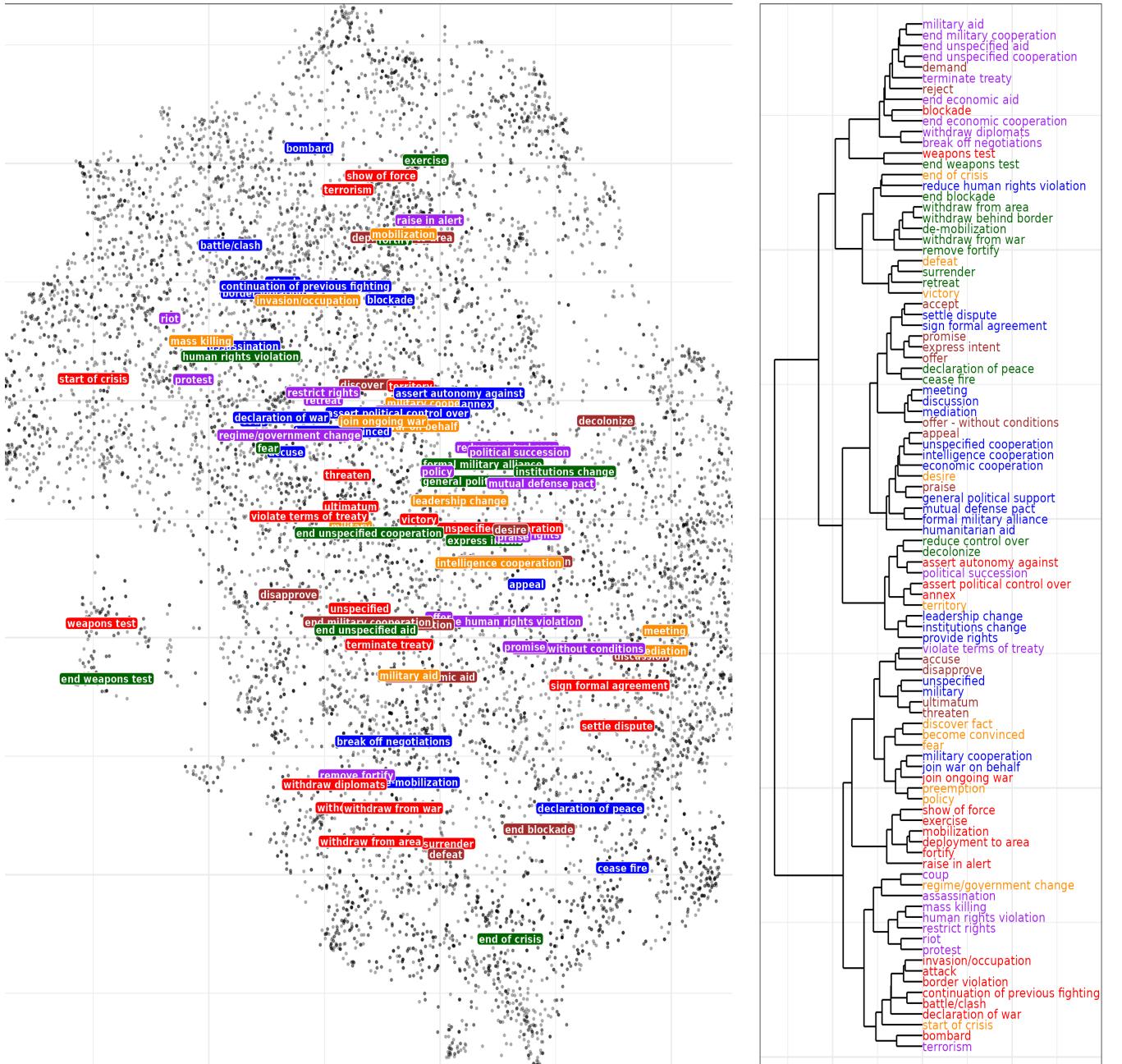


Figure 3: Computational evaluation of the precision of ICBe event codings. The plot on the left is a map of the semantic meaning of every sentence in the corpus (black points) as assigned by a large language model (Paraphrase-MPNET-base-v2) and projected down into two dimensions (UMAP). Overlaid are the average semantic locations of each label assigned by ICBe coders (colored labels). The labels with similar meaning are assigned to sentences with similar semantic meaning, creating an observable structure and pattern we would not observe with low quality coding where tag location would instead appear random. The plot on the right shows a hierarchical dendrogram clustering labels into groups by their average semantic location with more similar labels being more closely connected on the tree. The clustering by color indicates it closely mirrors the intended ICBe ontology, suggesting high precision in the coding.

## 4 Case illustrations

In this section, we focus our validation on two case studies for which we have produced synthetic narratives using the method described in Section 3.2. Our proposed measure is a reconstruction task to see whether our intended ontology can be recovered through only unsupervised clustering of sentences they were applied to. The first (Figure 1) is the Cuban Missile Crisis (hereafter Cuban Missiles) which took place primarily in the second half of 1962, involved the United States, the Soviet Union, and Cuba, and is widely known for bringing the world to the brink of nuclear war. The second (SI Appendix 4.1) is the Crimea-Donbas Crisis (hereafter Crimea-Donbas) which took place primarily in 2014, involved Russia, Ukraine, and NATO, and within a decade spiraled into a full-scale invasion. We choose these cases because they are significant in contemporary international relations, are widely known across academic disciplines as well as among the public, and are sufficiently brief to evaluate in depth. They are similar in that both cases involve a superpower in crisis with a neighbor that changed from a friendly to a hostile regime, both held implications for the economic and military security for the superpower by risking full-scale invasion, and both eventually invited intervention by an opposing superpower.

### 4.1 Cuban Missile Crisis (1962)

A synthetic historical narrative for Cuban Missiles appears in Figure 4, with 51 events drawn from 2,020 documents. Each row represents a detail that appeared in at least five documents along with an approximate start date, a handwritten summary, the number of documents it was mentioned in, and whether it could be identified in the text of the original ICB corpus, our ICBe events, and any of the competing existing models.

ICB's improved recall of Cuban Missiles was relative to the state of the art was summarized in Section 3.2 (Figure 2), but the events that explain that improvement can now be seen. Our ground truth ICB narrative contains 17/51 of the events from the synthetic narrative of a case that includes high-level previously classified details. ICBe captures nearly all details included in ICB as well as more details from the synthetic narrative than any competing dataset. Phoenix includes some earlier information than ICBe like the nationalization of businesses and back channel negotiations, but the crisis narrative has a clean canonical end with the Soviets agreeing to withdraw missiles. ICBe stands out in including more communicative behavior (do – speech) than existing datasets like US threats to attack and later promises not to invade. Given the recognized importance of threat credibility for understanding international conflict, the addition of this information is a substantively important improvement over the existing state of the art (Slantchev 2011).

Figure 5 shows the crisis map for the Cuban Missile Crisis. Looking at the crisis on a timeline, one can now identify the structure of actors and the environment, along with its supporting details, in a way that validates the precision of ICBe. Although harder to measure objectively, this crisis map provides face valid evidence that ICBe's account is not too vague, but also not unnecessarily detailed. We include much of

YMD	Ground Truth Events	ICB Corpus	ICBe	ICB	MIDs	Phoenix
Docs						
1958-12-31	communist coup	128				
1959-06-08	nationalizes 🇺🇸 owned businesses	28				✓
10-26	backchannel negotiates with 🇺🇸	18				✓
1960-01-01	begins recon flights over 🇺🇸	10				
03-17	prepares for invasion of 🇺🇸	159	✓	✓		
05-01	U-2 downed over 🇺🇸	8				✓
07	establishes diplomatic and trade relations with 🇺🇸	10				
07-08	embargos 🇺🇸	21				
09-14	attempts assassination 🇺🇸	9				
1961-01-03	breaks diplomatic ties 🇺🇸	40				
04-17	attempts coup in 🇺🇸	16				
	invades 🇺🇸	192	✓	✓	✓	✓
19	Invasion 🇺🇸 fails	5	✓	✓	✓	
06-01	provides economic and military aid to 🇺🇸	43	✓	✓		
08-13	begins construction of Berlin Wall	62	✓	✓	✓	✓
11-30	covert destabilization efforts against 🇺🇸	44				
1962-04-01	places nuclear missiles in 🇸🇮	93				
05-21	begins placing nuclear missiles in 🇺🇸	31				
07-01	asks 🇺🇸 for weapons	5				
	deploy troops to 🇺🇸	5				
	places nuclear missiles in 🇺🇸 as response to 🇺🇸 placing nuclear missiles in 🇸🇮	347				
08-10	begins to suspect 🇺🇸 nuclear missiles will be placed in 🇺🇸	29	✓			
09-04	meets 🇺🇸 to denies presence of missiles	14	✓			
	demands withdrawal and threatens nuclear response on 🇺🇸 if attacked from 🇺🇸	125				
11	threatens war with 🇺🇸 if attack on 🇺🇸 or ships	8				
10-01	deploy nuclear armed submarines 🇺🇸	5				
14	discovers 🇺🇸 nuclear missiles in 🇺🇸 starts crisis	705	✓	✓	✓	
17	mobilizes troops for invasion of 🇺🇸	13	✓	✓		
18	meets 🇺🇸 to denies presence of missiles	21				
22	blockades 🇺🇸	400	✓	✓	✓	✓
	demands withdrawal	7				
	raises nuclear alert	5				
	threatens military attack 🇺🇸	10				
23	OAS statement of support for 🇺🇸	7	✓	✓		
	meets 🇺🇸 denies offensive intention	8				
24	respects 🇺🇸 blockade of 🇺🇸	23			✓	
	raises nuclear alert	14				
25	confronts 🇺🇸 at the 🇺🇸	30				✓
26	offers remove missiles for 🇺🇸 no invasion pledge 🇺🇸	60	✓	✓	✓	
27	nuclear missiles in 🇺🇸 operational	5				
	accidentally violates 🇺🇸 airspace	25				
	interdicts 🇺🇸 submarine	16				
	offers withdrawal of nuclear missiles in 🇸🇮 for 🇺🇸 missiles in 🇺🇸	115				
	promises to not invade 🇺🇸	32	✓	✓		✓
	U-2 shot down over 🇺🇸	119	✓	✓	✓	
28	withdraws nuclear missiles in trade for 🇺🇸 promise to not invade 🇺🇸 ends crisis	647	✓	✓	✓	✓
30	refuses 🇺🇸 observers	7	✓	✓		
11-20	ends blockade 🇺🇸	22	✓	✓	✓	
1963-04-01	removes missiles from 🇺🇸	5				
08-05	sign Nuclear Test Ban Treaty	94				
30	install hotline	55				

Figure 4: Synthetic narratives combine several thousand accounts of each crisis into a single timeline of events, taking only those mentioned in at least 5 or more documents. Checkmarks represent whether that event could be hand matched to any detail in the ICB corpus, ICBe dataset, or any of the other event datasets (SI Appendix 3.2 and 3.3).

the geopolitically important details like Soviet deployment, US discovery of that deployment, heightened alert levels, a blockade, and negotiations that ended with a formal agreement. At the same time, the crisis map indicates that ICBe does not include unnecessary nuances that preclude useful comparison to other international events.

#### 4.2 Crimea-Donbas (2014)

A synthetic historical narrative for the 2014 Crimea-Donbas crisis (30 events drawn from 971 documents) appears in Figure 6. As in the earlier case, rows represent details that appeared in at least five documents and whether it is identified in ICBe and existing datasets.

Again quantitatively summarized earlier in Section 3.2 (Figure 2), our ground truth ICB narrative contains 23/30 of the events from the synthetic narrative. Like the gray zone precursor to the Cuban Missile crisis (Cormac and Aldrich 2018), Ukraine provided several security guarantees to Russia that were potentially undone, e.g. a long term lease on naval facilities in Crimea. But unlike the Cuban Missile crisis, the end of this crisis is unclear, with the event meekly ending with a second cease-fire agreement (Minsk II) but continued fighting. ICBe again recalls more important information about the crisis than any existing dataset, particularly information concerning the behavior of non-state separatist groups like the Donetsk People’s Republic (DPR) and Luhansk People’s Republic (LPR).

As this more recent case reflects primarily public reporting rather than the previously classified details relevant for the Cuban Missile Crisis, ICBe’s improvement relative to the global and real-time coverage of dictionary-based event systems is still present, but less pronounced. We want to take seriously the possibility that some functional transformation could recover the precision of ICBe. For example, Terechshenko (2020) attempts to correct for the mechanically increasing amount of news coverage each year by de-trending violent event counts from Phoenix using a human-coded baseline. Others have focused on verifying precision for ICEWs on specific subsets of details against known ground truths, e.g. geolocation (Cook and Weidmann 2019), protest events (80%) (Wüest and Lorenzini 2020), anti-government protest networks (46.1%) (Jäger 2018).

We take the same approach here in Figure 7, selecting four specific CAMEO event codings and checking how often they reflect a true real-world event from the Crimea-Donbas synthetic narrative. We choose four event types around key moments in the crisis. The start of the crisis revolves around Ukraine backing out of a trade deal with the EU in favor of Russia, but “sign formal agreement” events act more like a topic detector with dozens of events generated by discussions of a possible agreement but not the actual agreement which never materialized. The switch is caught by the “reject plan, agreement to settle dispute”, but also continues for Viktor Yanukovych even after he was removed from power because of articles retroactively discussing the cause of his removal. Events for “use conventional military force” capture a threshold around the start of hostilities and who the participants were but not any particular battles or campaigns. Likewise,

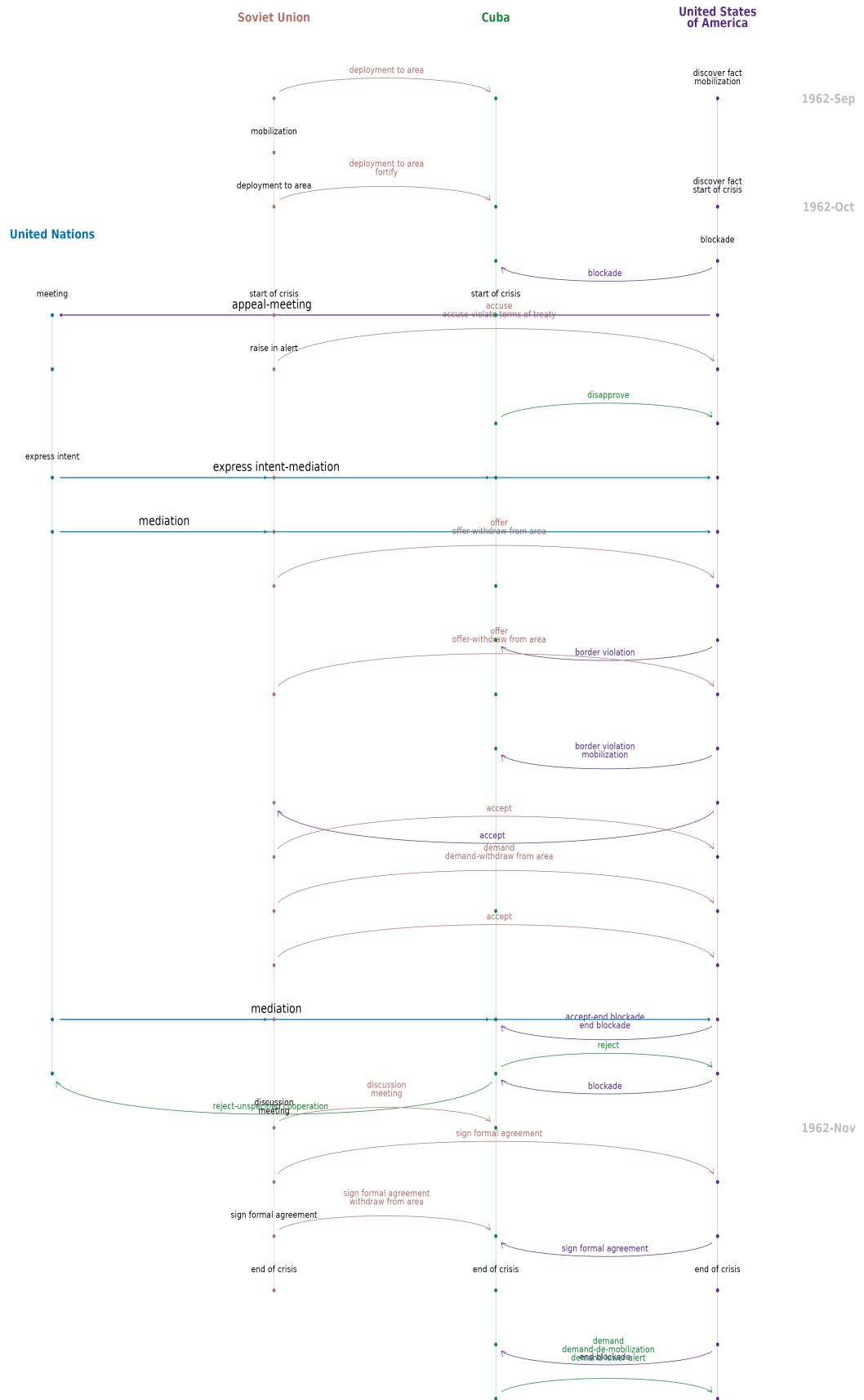


Figure 5: Crisis map for the Cuban Missile Crisis. The start of the crisis is at the top and end of the crisis is the bottom, with each actor in a column with labeled points identifying their speeches, actions, and thoughts.

Day	Ground Truth Events	Docs	ICB Corpus	ICBe	MIDs	phoenix	terrier	icews
1994-12-05	Ukraine nuclear disarmed in exchange for Russia promise to never invade	7						
2004-04-02	NATO expands to include Estonia, Latvia, and Lithuania	5						
2008-01-01	NATO provides incredible offer of membership to Ukraine	7	✓	✓				
08-01	Russia invades Georgia and annexes Abkhazia and Ossetia	10	✓	✓	✓			
2010-04-21	Russia and Ukraine ratified the Russian Ukrainian Naval Base for Gas treaty, extending the Russian Navy's lease of Crimean facilities for 25 years after 2017	5						
2013-11-01	Russia plans to join EU trade agreement	6	✓			✓		✓
21	Russia rejects EU trade agreement	42	✓	✓		✓	✓	✓
12-17	Russia offers debt relief and discounted energy to Ukraine	6	✓	✓		✓	✓	✓
2014-02-01	USA provide econ and military aid to Ukraine	5	✓			✓	✓	✓
21	Russia backs Ukraine political settlement	26	✓	✓		✓	✓	✓
22	Russia leader removed and flees to Russia	10	✓	✓		✓	✓	✓
27	Russia backed gunmen begin seizing government buildings in Crimea	65	✓	✓		✓	✓	✓
03-01	Russia votes to deploy military to Crimea	10	✓	✓		✓	✓	✓
16	Crimea independence referendum	11	✓	✓		✓	✓	✓
	Russia mobilizes forces to Crimea and Ukraine border	12	✓			✓	✓	✓
17	Russia sanctions Ukraine	11	✓	✓		✓	✓	✓
18	Russia annexes Crimea	6	✓	✓		✓	✓	✓
04-06	Russia backed separatists begin fighting in Ukraine (Donetsk and Luhansk)	180	✓	✓	✓	✓	✓	✓
23	Russia military exercises at Ukraine border	146	✓	✓	✓	✓	✓	✓
05-11	DPR and the LPR declare independence	7	✓	✓		✓	✓	✓
21	DPR requests Russian military intervention	5	✓			✓		
07-17	Russia shoots down passenger jet (Malaysia Airlines Flight 17)	5	✓	✓		✓	✓	✓
	Russia sanctions Ukraine	5	✓	✓		✓	✓	✓
09-05	Russia, the DPR and the LPR signed a ceasefire agreement, the Minsk I	5	✓	✓		✓	✓	✓
06	Russia, the DPR, and the LPR, continue fight with Ukraine	10	✓	✓		✓	✓	✓
2015-02-12	Russia, the DPR, and the LPR signed a ceasefire agreement (Minsk II)	15	✓	✓		✓	✓	✓
21	Russia recognizes Donetsk People's Republic and the Luhansk People's Republic	10	✓	✓		✓	✓	✓
		6						

Figure 6: Synthetic narratives combine several thousand accounts of each crisis into a single timeline of events, taking only those mentioned in at least 5 or more documents. Checkmarks represent whether that event could be hand matched to any detail in the ICB corpus, ICBe dataset, or any of the other event datasets (SI Appendix 3.2 and 3.3).

“impose embargo, boycott, or sanctions” captures the start of waves of sanctions and from who but are effectively constant as the news coverage does not distinguish between subtle changes or additions. In sum, dictionary-based methods on news corpora tend to have high recall because they parse everything in the news, but for the same reason, their specificity for most event types is too low to back out individual chess-like sequencing that ICBe aims to record.

## 5 Conclusion

The scope and complexity of international politics should not discourage the identification of trends, patterns, and regularities. In undertaking event abstraction from narratives about key historical episodes in international relations, this paper has proposed a mapping between unstructured historical records and a structured ontology of these events with high coverage of concepts of interest. Multiple validity checks find the resulting codings have high internal validity (e.g. intercoder agreement) and external validity (i.e. matching source material in both micro-details at the sentence level and macro-details spanning full historical episodes). Further, these codings perform much better in terms of recall, precision, coverage, and overall coherence in capturing these historical episodes than existing event systems used in international relations.



Figure 7: The unit of analysis is the dyad-day. Top 10 most active dyads per category shown. Red text shows events from the synthetic narrative relative to that event category. Blue bars indicate an event recorded by ICEWs for that dyad on that day.

These data, along with the open-source code, documentation, and companion website provide several substantive and methodological contributions to the discipline. Substantively, these data are appropriate for statistical analysis of hard questions in the study of crises like interactions between means of warfare and the preconditions for conflict escalation (Gannon 2022). Methodologically, our mapping from codings to source text at the sentence level provide a new resource for natural language processing with access to coder-level disaggregation that furthers the study of uncertainty in the interpretation of international events and in the quantitative coding of historical events. Finally, we provide a companion website ([crisisevents.org](http://crisisevents.org)) that incorporates detailed visualizations of all the data introduced here as a new resource for the study of international crises in a scalable, yet detailed, manner.

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## 8 Author Contributions

Conceptualization: R.W.D., E.G., J.L.; Methodology: R.W.D., T.L.S.; Software: R.W.D.; Validation: R.W.D., T.L.S.; Formal Analysis: R.W.D., T.L.S.; Investigation: S.C., R.W.D., J.A.G., C.K., N.L., E.M., J.M.C.N., D.P., D.Q., J.W.; Data Curation: R.W.D., D.Q., T.L.S., J.W.; Writing - Original Draft: R.W.D., T.L.S.; Writing - Review & Editing: R.W.D., J.A.G., E.G., T.L.S.; Visualization: R.W.D., T.L.S.; Supervision: E.G.; Project Administration: S.C., R.W.D., J.A.G., D.Q., T.L.S., J.W.; Funding Acquisition: E.G., J.L.

## 9 Data Availability Statement

This article’s data, supplementary appendix, replication material, and visualizations of every historical episode are available on the GitHub repository ICBEEventData and through the companion website crisisevents.org.

## 10 Competing Interests Declaration

The authors declare that there are no competing interests.

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