

## **R&R Memo for manuscript #ISQ-2022-11-0579 entitled "Planes, Trains, and Armored Mobiles: Introducing a Dataset of the Global Distribution of Military Capabilities (rDMC)"**

Please find my memo indicating changes made to address editor and reviewer comments. I am grateful to the editorial team and reviewer for the opportunity to revise and resubmit the manuscript, as well as for the thoughtful and enlightening feedback they have provided. I am very happy with ways the paper has improved and hope that you all feel the same way.

The memo is organized by reviewer comment:

1. Each suggested change is provided first *verbatim and italicized*
2. My response is indented
3. Where appropriate, I have replicated old and new figures/tables and also pasted the new verbatim text from the manuscript/appendix highlighted in gray.

I hope this eases the process of re-reviewing the paper by minimizing back and forth between the memo and the revised manuscript. Thank you all, again.

### **Editorial Team review**

*Editorial Team: "Make clear that this is not new data, but a database/dataset that makes currently available data more accessible to a wider community. This point is made by both reviewers."*

This has been clarified in the abstract, introduction, and section on "scope and data generating process". The conclusion also now clarifies that future versions of the data will diversify the sources used.

*Editorial Team: "Discuss in greater detail the reliability issues raised by Reviewer 2. Be clear that the currently available data have biases. This is also a shared concern of both reviewers."*

A new paragraph has been added to the end of the "scope and data generating process" section to discuss this in greater detail. It is replicated in full below under Reviewer 2's concern.

*Editorial Team: "Reviewer 2 (point 5) asks for an explicit finding that runs counter to our current understanding in the literature. We recognize that this submission is a research note and, therefore, not intended to make a larger empirical contribution per se. However, a demonstration of this argument would increase the impact of the article. It may be enough to provide a face-valid example using the database/dataset. If the author chooses not to include*

*such a demonstration, it would be better to delete the oblique references “calling into question several findings in the literature.””*

I have added a new subsection “6.3 Validation and Application: US and China Polarity” that makes an empirical contribution by showing how rDMC can contribute to our understanding of the US-China military capabilities gap in debates about China’s rise by replicating and extending Brooks and Wohlforth (2016). Given the current length of the paper and the research note format, I felt a replication and validation of descriptive data rather than a full statistical model was more appropriate. Nonetheless, I feel this has improved the paper and increased the potential impact of the article and also provided face validity about the quality and utility of the dataset. The other sections of the paper that called into question several findings in the literature have been rephrased to show how the data could be used to test these theories/hypothesis, rather than this research note doing that testing itself.

*Editorial Team: “Figure 4 is showcasing the variety of categories included in the database/dataset (which other figures also do), but it is overwhelming. The two subplots with “Na” raise questions. Without a specific legend regarding gradation, the reader does not know how widespread any particular weapon system is.”*

Figure 4 and its accompanying in-text references have been removed. The subsection “Technological trends across time” still discusses the research on diffusion and provides an example of how rDMC could be used for research on diffusion and military effectiveness with the original example and figure of aircraft.

*Editorial Team: “Reviewer 2’s comment of changing the year regarding Figure 7 seems reasonable.”*

This change has been made. Figure 6 (which used to Figure 7) now uses 1994 instead of 2004. The accompanying text has been updated to reflect the countries that are most similar to one another in 1994, with a particular emphasis on great powers.

### **Reviewer 1**

*Reviewer 1: “This is not a new source of data as all of it is drawn from the existing Military Balance. Thus, the utility of this project is the way it allows researchers to manipulate military capabilities data in a variety of ways to examine change over time and comparisons between states.”*

The abstract, introduction, and data collection section now clarify that the data comes from the existing Military Balance, and what is novel is it is now (1) organized in a structured and machine-readable format (.csv files) and (2) a new

ontology of consistent weapons categories has been produced (rDMC long and rDMC wide), making the sample consistent across countries and years.

Reviewer 1: *"The manuscript is publishable as a research note in its current form with the exception of what appears to be a comment from a previous reviewer on page 7 just below table 2."*

That comment has been removed.

## **Reviewer 2**

Reviewer 2 Important Change: *"1. Ultimately, the author(s) are really introducing a new dataset, not new data. This fact is not really apparent until p6. Since we use the word "dataset" somewhat ambiguously in political science, I think, in the interests of candor, it is especially important the author(s) make it clear that all of these data are already available to scholars, just not in a very useful form. The author(s) have done us all an important service in categorizing the data and making it accessible, including by introducing their tech categories and sorting everything into them. I think this contribution should be clear in both the intro and abstract."*

The abstract, introduction, and data collection section now clarify that the data comes from the existing Military Balance, and what is novel is that it is now (1) organized in a structured and machine-readable format (.csv files) and (2) a new ontology of consistent weapons categories has been produced (rDMC long and rDMC wide), making the sample consistent across countries and years. The words "new" and "novel" are no longer used to describe the dataset

Reviewer 2 Important Change: *"2. To justify the need for this dataset, the author(s) state "most current research uses measures only of the size of state militaries." This description matches my impression of the literature. However, the author(s) cite almost none of this current research, so far as can tell. This omission is odd, because they are otherwise quite thorough in canvassing existing scholarship. (The author(s) do cite several scholars who have criticized/attempted to improve various measures, e.g. Lebovic (1999).) I imagine the author(s) did not want to "name names" and risk offending anyone, since the gist of their point is that relying on such coarse measures is a mistake. I believe it is vital, though, to cite some of the research they are criticizing."*

Cites have been added to the section on existing measures of military capabilities. While there are many relevant works using aggregate measures of military spending, these cites were chosen based on (1) an interest in citing both foundational and recent work, (2) using aggregate military spending measures as

a dependent variable in some cases, and independent variable in others, (3) honoring ISQ's commitment to citing works by underrepresented groups, and (4) being high-quality and influential works. The section is reproduced below and new citations listed:

"Despite the importance of disaggregated military capabilities and a recognition of the shortcomings of aggregate measures for explaining concepts of interest, most current research uses measures only of the size of state militaries (Ward and Davis 1992; Sample 2002; Nordhaus, Oneal, and Russett 2012; Cappella Zielinski, Fordham, and Schilde 2017; Alley and Fuhrmann 2021)."

Michael D. Ward and David R. Davis, "Sizing up the Peace Dividend: Economic Growth and Military Spending in the United States, 1948–1996," *American Political Science Review* 86, no. 3 (September 1992): 748–55, <https://doi.org/10.2307/1964136>.

Susan G. Sample, "The Outcomes of Military Buildups: Minor States vs. Major Powers," *Journal of Peace Research* 39, no. 6 (November 1, 2002): 669–91, <https://doi.org/10.1177/0022343302039006002>.

William Nordhaus, John R. Oneal, and Bruce Russett, "The Effects of the International Security Environment on National Military Expenditures: A Multicountry Study," *International Organization* 66, no. 3 (July 2012): 491–513, <https://doi.org/10.1017/S0020818312000173>.

Rosella Cappella Zielinski, Benjamin O. Fordham, and Kaija E. Schilde, "What Goes up, Must Come down? The Asymmetric Effects of Economic Growth and International Threat on Military Spending," *Journal of Peace Research* 54, no. 6 (November 1, 2017): 791–805, <https://doi.org/10.1177/0022343317715301>.

Joshua Alley and Matthew Fuhrmann, "Budget Breaker? The Financial Cost of US Military Alliances," *Security Studies* 30, no. 5 (October 20, 2021): 661–90, <https://doi.org/10.1080/09636412.2021.2021280>.

*Reviewer 2 Important Change: "3. As I said above, I am persuaded that these data are largely reliable and accurate, and more so than e.g. data from the IMF. Nonetheless, the author(s) too blithely dismiss objections to its completeness. Yes, "it is widely recognized as the best unclassified source..." —but there's the rub. Obviously, many important capabilities will not be unclassified. There are biases built into this dataset, and it will probably undercount important platforms, especially among the most powerful states. It's important to acknowledge that limitation. (And no, I'm not persuaded that checking its accuracy against government reports, when available, from countries like New Zealand proves its overall accuracy. New Zealand is not China, nor America for that matter.)"*

I thank the reviewer for this important point. Believing rDMC and the Military Balance to be the existing state of the art does not mean they are free from bias. I have added a paragraph to the “Scope and Data Generating Process” section that describes what some of these biases may be (temporal, cross-national, and platform-specific), and identifies an avenue of future research in identifying these biases with examples from autocratic GDP data and Russian losses in the war in Ukraine. The new paragraph is pasted below in full:

However, being the most comprehensive data on military capabilities and from a highly reputable source does not guarantee accuracy. States themselves are uncertain about the military capabilities of others, and more pronounced uncertainty likely exists in the open source data used here (Kaplow and Gartzke 2021). Earlier years may have less accurate coverage as both private and government intelligence and surveillance efforts were less sophisticated, making it harder to know the capabilities of states that were far away (Lin-Greenberg and Milonopoulos 2021). New Zealand's transparency in capability reports is an anomaly, as many countries – particularly those of interest to security scholars like Russia and China – try to deceive their foes (or even friends) by misrepresenting their capabilities by exaggerated or underplaying their strength in particular platforms (Mastro 2016; Mawdsley 2016). Weapons systems also differ in whether open source intelligence is likely to identify them correctly, as capabilities whose effectiveness relies on secrecy and deception may not be known to the rest of the world until they are used (Green and Long 2020). It is important that scholars acknowledge this limitation and, where possible, take steps to address it. For example, biases in GDP estimates have been identified using night-time-light data gathered from satellites, allowing scholars to quantify the extent to which autocratic regimes exaggerate their economic growth (Martinez 2022). In the future, similar measures could be applied to rDMC by, as has been done with Russia in the invasion of Ukraine, combining reports about military use and loss during combat to pre-war inventory estimates (Zhukov, n.d.).

*Reviewer 2 Important Change: “4. I think the dataset’s modularity will be especially valuable to researchers in different areas of security studies. It should be mentioned in the abstract.”*

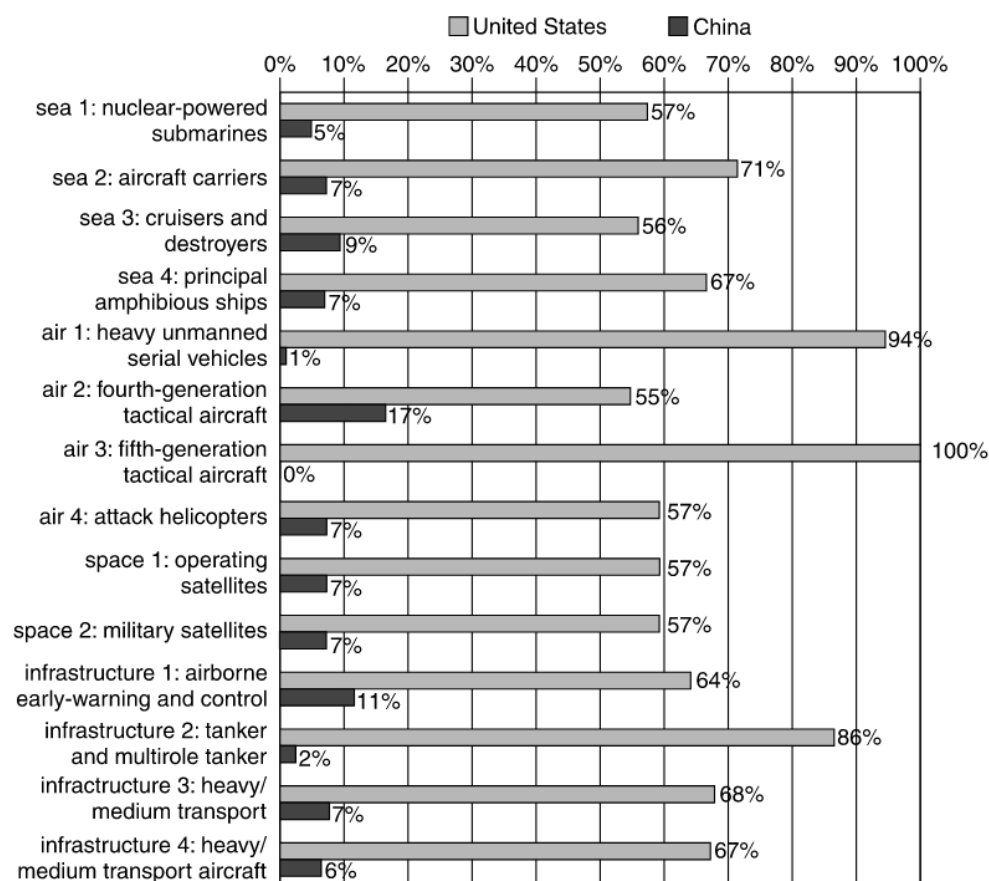
Modularity is now mentioned in the abstract.

*Reviewer 2 Important Change: “5. The author(s) hint, and come close to saying, that the data call into question several findings in the literature. They dance around this especially on p12. I think they should explicitly identify a finding, then carefully show how their new dataset should*

change our minds about it. Of all the revisions I'm suggesting, this will be the most time-intensive."

I have added a new subsection "6.3 Validation and Application: US and China Polarity" that makes an empirical contribution by showing how rDMC can contribute to our understanding of the US-China military capabilities gap in debates about China's rise by replicating and extending Brooks and Wohlforth (2016). For reference, here is the figure in Brooks and Wohlforth that is replicated:

Figure 2. Command of the Commons, the United States and China as Percentages of Six Major Powers



NOTE: Calculated from data in Stephen G. Brooks and William C. Wohlforth, "Moving Beyond Unipolarity? China's Rise and the Fate of America's Global Position," *International Security*, Vol. 40, No. 3 (Winter 2015/16), p. 20, fig. 1.

Given the current structure of the paper and the research note format and word limit, I felt a replication and validation of descriptive data rather than a full statistical model was more appropriate. I feel this has improved the paper and

increased the potential impact of the article and also provided face validity about the quality and utility of the dataset.

The other sections of the paper that called into question several findings in the literature have been rephrased to show how the data could be used to test these theories/hypothesis, rather than this research note doing that testing itself.

The author is currently working on other papers that use rDMC to change our minds about existing finding in more robust and theoretically-driven way. The goal is this piece is to introduce the dataset, make it publicly available as soon as possible, and identify some ways in which it can be used by other scholars. My hope is that the current structure accomplishes those ends within the smaller word limit that exists for a research note.

*Reviewer 2 Other Comments: "1. I think the title is too cute. It's clever, but it also seems unserious. I'd leave it up to the author(s), but I think a straightforward title like "Introducing a Dataset of Global, Disaggregated Military Capabilities (rDMC)" would be better."*

The reviewer's point is well taken, but I have decided to keep the title as is since it is known colloquially to some relevant scholars as the "Planes, Trains, and Armored Mobiles" dataset.

*Reviewer 2 Other Comments: "2. The author(s) should cite the growing literature on multi-domain warfare, e.g. Perkins (2017). After all, the very premise of such warfare is that different domains are not interchangeable, which is the justification for this dataset."*

The section on "The case for disaggregating defense" now cites some of the research on multi-domain warfare. A complete excerpt of the new text is below:

More recently, the increasing interest in multi-domain warfare demonstrates that among practitioners and academics alike, there is a recognition that the various tools of warfare - and the domains in which they operate - serve different ends and come with distinct advantages and disadvantages(Tan 2017; Perkins 2017; Kreps and Schneider 2019; Lindsay and Gartzke 2019)

*Reviewer 2 Other Comments: "3. While it is obvious from figure 1, I think it would also be worthwhile to say, on p5, that the data has almost 100% coverage of the great powers."*

Clarification added in the sentence with missingness percentages.

*Reviewer 2 Other Comments: "4. On p7, the manuscript seems to include a passage from a previous review. I assume the author(s) forgot to delete this."*

Passage omitted.

*Reviewer 2 Other Comments: "5. Footnote 7 says that standardizing unit-level names will take some time. Why can't this be automated?"*

As this is the main improvement I want to do for rDMC 2.0, I am currently trying to figure out ways to standardize the names based on existing military sources (NATO names) and existing public sources (wikipedia has pages for many of the units and each page has a standardized unique numeric key). Part of the impetus for including the main battle tanks figure in this manuscript was as a test case in standardizing the unit names, which I was unable to automate and had to do by hand for that figure. Having also now done that for 4<sup>th</sup> and 5<sup>th</sup> generation fighter aircraft in the new replication/extension, I am even more confident standardizing the names will take some time (and Ras). A few things have made the automation process challenging, although I still think it will be possible and plan to do it:

1. Names differ across countries – the Soviet MiG-21 has the NATO reporting name Fishbed. When the Soviet Union sold it to China, China renamed it the Chengdu J-7 which NATO calls the Fishcan. Even if we assume China made some changes so the Chengdu-7 is similar, but not identical to/synonymous with the Mig-21, one part of standardizing the names is deciding whether to standardize by giving every unit its NATO reporting name, producing a new dataframe of "equivalent to" that identifies all the other names for the Fishbed-21, or coming up with a new standardized name since NATO does not have names for all Soviet aircraft (nonetheless all equipment in the world).
2. Fuzzy string matching – the Military Balance pdf's cannot be scraped and were instead processed by hand. Coders were instructed to enter the unit names exactly as they were written in the PDF since not all were experts in military capabilities. As a result, there are entries for "F35", "F-35", "F 35", "F-35 Lightning II", etc. Many, but not all, of these differences concern delimiters (spaces, hyphens, etc) that can partially be matched automatically, but the use of full names or cases where the NATO name or full name are in parentheses complicates it.
3. Variants – some times, data is provided separately for the F-35A, F-35B, and F-35C. Other times, data is just provided for F-35 (no variant specified). Since the delimiters here are applied inconsistently, it's hard to know how to draw those



out. It's also unclear where the differences across the variants are significant enough to be of interest to scholars and how that should be handled systematically.

4. Combined variants/units – Related to the point above, other times the data is aggregated across some (but not all) variants of a same unit (F-35A/B) but yet other times counts are provided for an aggregation of multiple units (in 2011, Armenia had 8 T-54/T-55). Since other country-years have counts for the T-54 and T-55 separately, it's hard to know how to deal with the aggregated counts.

*Reviewer 2 Other Comments: “6. Footnote 10 is a testament to the care the author(s) and their coders have taken in constructing the dataset.”*

Thank you for the kind words.

*Reviewer 2 Other Comments: “7. I do not find figure 4 helpful. But perhaps other reviewers might.”*

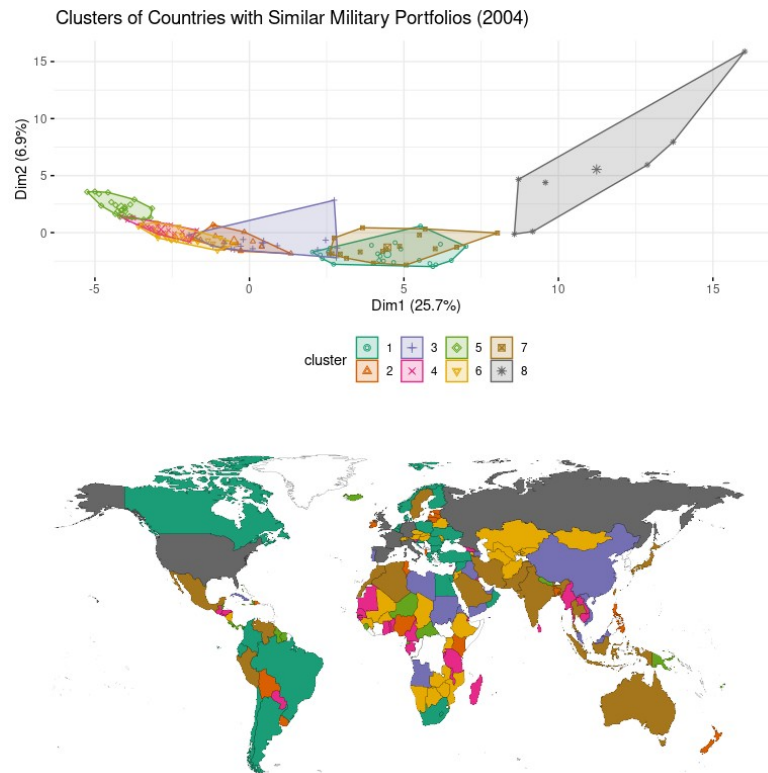
Figure 4 and its accompanying in-text references have been removed. The subsection “Technological trends across time” still discusses the research on diffusion and provides an example of how rDMC could be used for research on diffusion and military effectiveness with the original example and figure of aircraft.

*Reviewer 2 Other Comments: “8. Figure 7 looks at a single year, 2004, to show how states do/don't cluster. I assume the author(s) chose 2004 because it's a decade before their data end. But it's also a year when the US is at war alongside a variety of countries. Presumably that affects the composition of their arsenals and their similarity. Would it be possible to look at a year where we can get a better feel for peacetime arming? 1994 would allow the author(s) to stay with their same ten-year intervals. I think, when constructing this figure, “boring is better” when choosing a year to depict.”*

The new figure and accompanying text now depict 1994. Both are reproduced below:

There are eight distinct clusters of countries that share significant commonality in the military capabilities they possess. Some great powers are similar to one another (the US, Russia, France, and UK), but others appear more distinct (China and India). In some cases the US is similar to allies like Germany and Italy, but not with others like Poland and Canada. There is similarity across states in some

geographic regions like Central Asia, but significant dissimilarity across states in other regions like the Pacific Rim.



Reviewer 2 Other Comments: “9. Figure 7 seems to suggest that the Great Powers are playing one game, and everyone else is playing another. Contrary to the implication at the top of p16, I’m therefore not surprised that the US looks more like France than Poland, since the former still tries to play in the big leagues, while the latter does not.”

The countries listed as being similar vs different from the US has been updated given the change in year displayed. Interestingly, China is no longer in the same cluster as the US and Russia, now instead possessing similar weapons (albeit, in different quantities and more advanced variants) as Cuba, Iraq, North Korea, Libya, and Yemen. There may be something interesting to be said (later) about autocratic inclination towards certain weapons, maybe given arms sales patterns, coup risk, recent history of civil conflict, industrialization patterns, etc. Food for thought!

# **Planes, Trains, and Armored Mobiles: Introducing a Dataset of the Global Distribution of Military Capabilities (rDMC)\***

Anonymized author

March 14, 2023

This article introduces a dataset on disaggregated national military capabilities from 1970 – 2014 drawn from the International Institute for Strategic Studies Military Balance. While practitioners have long recognized the importance of what weapons states own, scholars have largely examined surrounding questions in piecemeal fashion due to data limitations. The Distribution of Military Capabilities (rDMC) dataset identifies the weapons portfolios of states over the past half century at various levels of aggregation suitable to a wide variety of research questions. This paper begins by explaining the value of data on disaggregated national military capabilities, the data’s scope, and the data collection process including the creation of a new modular ontology of weapons categories consistent across time and space. I then identify some initial trends about changes in the distribution of military capabilities and its implications for China’s military rise. These data allows scholars to better investigate important questions concerning power projection, military innovation, conflict outcomes, use of force decisions, and interest group lobbying.

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\*Acknowledgements omitted for blinded peer review.

# 1 Introduction

The military capabilities states possess are an important instrument of military power, and consequently national power. Yet existing work on how states exercise military power are limited in empirical identification to coarse measures like military spending or military personnel. Not all soldiers are created equal, and much has been said about the problems of measuring military power using military spending figures (Perlo-Freeman 2017) or aggregate measures like the Composite Index of National Capabilities (CINC) (Beckley 2018; Carroll and Kenkel 2019). I define military capabilities as the technologies and weapons that states can use in combat operations. While military capabilities are but one of only many components of military power, its effects are significant, if hotly debated (Lieber 2005; R. A. Brooks and Stanley 2007). Inconsistent findings about the role of capabilities in conflict stem not from the fact that they do not matter, but rather they have been improperly identified and coarsely measured.

This paper contributes to ongoing research about the technological dimension of military power by producing the first comprehensive and structured dataset of the distribution of military capabilities across all states from 1970 – 2014 by cleaning and structuring the annual IISS Military Balance in a machine-readable format and creating a new ontology of military capabilities that makes comparison across time and space possible. Disaggregating military power into its component parts is an important, yet underdeveloped, enterprise. While aggregate military spending may help differentiate large and globally capably militaries from smaller ones, it risks conflating differences in the *composition* of nominally equivalently sized militaries. The composition of a state’s military may influence its power projection and warfighting capabilities in some conflicts, but not others (Gartzke and Lindsay 2020), and the relationship between the military capabilities a country could acquire, actually possesses, and subsequently uses in a contest could shed light on contrasting findings about the impact of military capabilities on

international affairs (Douglass and Gannon 2019).

This paper proceeds as follows. Section 2 identifies the role that military capabilities play as both a dependent and independent variable in the study of international politics. Section 3 outlines existing empirical work on the military capabilities states possess. Section 4 outlines the scope of the Distribution of Military Capabilities (rDMC) dataset and Section 5 briefly describes the data collection process, which consistent of structuring, cleaning, and re-categorizing the existing Military Balance data to make both quantitative and qualitative comparison and analysis possible. Section 6 identifies some initial trends in variation in the distribution of military capabilities across time and space and replicates and extends existing findings on the US-China balance of military capabilities. Section 7 concludes.

## 2 The case for disaggregating defense

One of the most political aspects of international security for policymakers concerns what military capabilities a state should possess. What a states produces and omits in its defense portfolio reflects its political priorities in ways that economic considerations alone cannot explain (Caverley 2007). As then-Senator Joseph Biden (2008) remarked regarding domestic economic policy during his vice presidential campaign, “[d]on’t tell me what you value. Show me your budget, and I’ll tell you what you value.” This also holds true in the military context, where disconnects concerning a state’s budget and alleged priorities are missed by scholars looking at top-line spending figures. Despite decades of concern about a Chinese invasion of Taiwan, it was not until 2017 that China began investing in the amphibious assault capabilities that are necessary for that threat to actually be carried out (Office of the Secretary of Defense 2018, 95–99). This important fact was missed by many who looked only at the size of China’s defense budget, rather than what capabilities they actually possessed.

The wide body of literature studying war has recognized that international affairs is influenced by *how* states fight (Gannon 2022). The United States' decision to adopt a more capital-intensive military concentrated in a few high-value weapons may explain its relatively infrequent uses of force, given its high military spending (Fordham 2004), while states with combined arms militaries are able to end civil conflicts more rapidly (Caverley and Sechser 2017). Looking at specific military domains, states with capable navies fight wars further from their borders (Crisher 2017) while those conducting aerial mobilizations fail at demonstrating resolve (Post 2019). Concerning intrastate conflict, access to armored fighting vehicles helped the Yemeni Army secure the capital during the 1962 coup (Albrecht and Eibl 2018, p 319) and well-resourced armies are better able to negotiate with their civilian counterparts in a way that shapes civil-military relations (Kadercan 2014). More recently, the increasing interest in multi-domain warfare demonstrates that among practitioners and academics alike, there is a recognition that the various tools of warfare - and the domains in which they operate - serve different ends and come with distinct advantages and disadvantages (Tan 2017; Perkins 2017; Kreps and Schneider 2019; Lindsay and Gartzke 2019).

Domestically, the portfolio of military capabilities also informs questions outside the traditional scope of international security. Tension between political budget constraints and military security desires gets to the heart of the role that military capabilities play in furthering our understanding of domestic politics (Cappella Zielinski 2016). Partisan politics is intricately tied to the composition of a state's military, as the Cold War witnessed a preference among Republican US presidents for funding strategic forces while Democratic presidents preferred funding conventional forces (Fordham 2002). In the 1960's, the US developed the F-111 fighter-bomber and C-5A jumbo transport aircraft to keep production lines for the four major aerospace corporations funded, demonstrating how patterns of weapons acquisition can tell us about the role of domestic industry in national security decisions (Kurth 1972).

### 3 Existing measures of military capabilities

The combination of capabilities that comprise a military's toolkit determine the operations it undertakes, the types of threats it can credibly make, and the consequences of resorting to force. Military means likely matter, but proving that they do (and if so, the degree and circumstance) is an impossible endeavor without empirical data on the distribution of those means across time and space. Despite the importance of disaggregated military capabilities and a recognition of the shortcomings of aggregate measures for explaining concepts of interest, most current research uses measures only of the *size* of state militaries (Ward and Davis 1992; Sample 2002; Nordhaus, Oneal, and Russett 2012; Cappella Zielinski, Fordham, and Schilde 2017; Alley and Fuhrmann 2021).<sup>1</sup> National militaries are primarily quantified and compared by military spending levels or, in rarer cases, military personnel counts. Yet, defense dollars don't tell us what capabilities a state can actually use during a conflict. Despite its frequent use, data on military spending also poses known problems for quantitative cross-national and temporal comparison (Perlo-Freeman 2017). Since there are no common definitions about what constitutes military spending, some states measure things like pension and R&D while others do not (Amara and Paskevics 2010). Exchange rates are often used to standardize all spending to the same currency, but with ill-information applications of domestic purchasing power (Fontanel 1996). Difficulty accounting for inflation and budget cycles also dramatically impacts inferences drawn from different data sources (Lebovic 1999). The IMF, UN Office for Disarmament Affairs, US Bureau of Arms Control, Verification, and Compliance, International Institute for Strategic Studies (IISS), and Stockholm International Peace Research Institute (SIPRI) all produce annual military spending estimates using different estimating procedures, primary sources, and preparation methods (Brzoska 1981), meaning findings from one data source are often not supported using comparable data from another source (Boniface 1995).

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<sup>1</sup>Of course, these two are not synonymous. Military capabilities are only one component of military spending and the cost of labor is a large component of military spending (Whitten and Williams 2011).

Since the cost of military platforms has increased over time, having data on the actual platforms helps get around the problems of accounting for military “purchasing power” over time.

Research on variation in the *composition* of militaries remains nascent. Scholars have noted “there is a lack of knowledge about variation between states in their behavior on armaments policy decisions” because of problems empirically identifying the military capabilities states possess (Mawdsley 2018). Recent works that disaggregate military capabilities have narrowed their focus to a small set of platforms like mechanized vehicles (Lyall and Wilson 2009; Sechser and Saunders 2010), strategic lift aircraft (Kupchan 1988), principal naval platforms (Crisher and Souva 2014), or fighter jets (Saunders and Souva 2020) or to a small set of countries like great powers (S. G. Brooks and Wohlforth 2016) or China and its regional rivals (Beckley 2017). Even the most detailed of these are selective in their disaggregation; Saunders and Souva (2020) do not include bombers, air transport, or support aircraft, Sechser and Saunders (2010) omit helicopters, and Souva (2022) aggregates all capabilities to five categories (nuclear weapons, naval tonnage, air power, armor, and ballistic missiles). The primary reason for this relatively limited use of fine-grained high-quality data is difficulty in converting the data to an easily-usable format and standardizing it across countries and years. rDMC aims to overcome all of these problems in being a massively expanded dataset that better enables systematic analysis across time, space, and military capability.

## **4 The rDMC dataset: scope and data generating process**

This paper uses the annual International Institute for Strategic Studies (IISS) Military Balance reports (n.d.) to create a structured and machine-readable dataset of disaggregated military capabilities measured as the annual count of military capabilities owned by 184 countries from 1970 – 2014. By military capabilities, I mean the technologies, weapons, and equipment that



states can use in combat operations, excluding small arms like handguns and rifles and also excluding munitions and bombs for which annual data is unreliable.<sup>2</sup> The data were coded by a team of 39 undergraduate coders from four universities over five years, resulting in 354,405 unique observations. Figure 1 shows the percent of years coded for each country. 54% of countries have no missing data across the entire time span (including the majority of the great powers), with the median country having data available for 95% of its years and the mean country having 88% coverage.<sup>3</sup> The data and all code used to process, clean, and analyze the data are publicly available at (website redacted).

### Temporal Coverage of the Data by Country

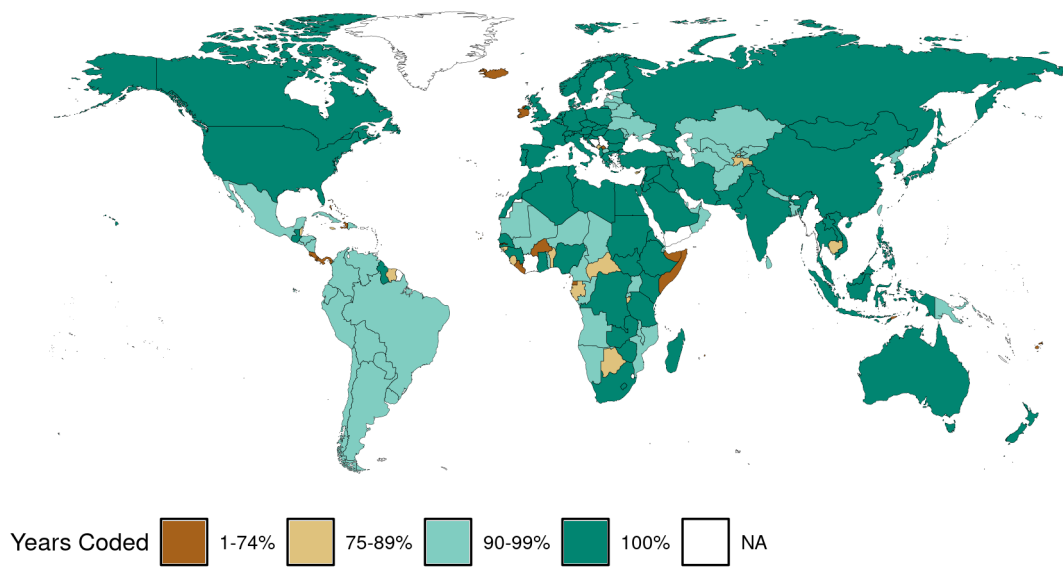


Figure 1: Missingness map illustrating the extent of available data on state military capabilities. Almost all states have data for at least 90% of the years they were recognized state entities.

Policymakers frequently use the IISS Military Balance to guide their decisions. The former

<sup>2</sup>See the section “Defining military capabilities” in the codebook for a more detailed definition of military capabilities.

<sup>3</sup>I use percent of years since not all countries exist across the entire duration of the data. 100% of Slovenia is covered, for example, because data exists for all 22 years since it established independence in 1991.

Supreme Allied Commander of NATO, Admiral James Stavridis, said “throughout my career, I have relied extensively on The Military Balance produced so expertly by the IISS. It is the “go to” source for serious analysts and warriors facing real world challenges.” Former US Army General Petraeus describes it as “the go-to source of unclassified, independent information on defense capabilities around the world”. Similarly, former US Secretary of Defense Robert Gates noted it “provides essential facts and analysis for decision-makers and for better informed public debate” and former US Secretary of Defense Leon Panetta remarked that it is “widely recognised as the best unclassified source of defense information on personnel, equipment and budgets for every country.”<sup>4</sup>

Practitioner praise notwithstanding, as with much data in international affairs, concerns about data quality and accuracy remain salient. The extensive use of the IISS Military Balance by policymakers as well as by other scholars gives cursory confidence in its accuracy and allows for the data codings to be validated. Most of this academic work has used IISS data on military spending (Wohlforth 1999; Greenhill and Major 2007) or personnel (Stanton 2013; Gaibullov et al. 2015). Recent works by Lyall and Wilson (2009), Sechser and Saunders (2010), S. G. Brooks and Wohlforth (2016), and Caverley and Sechser (2017) use detailed information from the IISS Military Balance on capabilities like mechanized vehicles and power projection platforms. The accuracy of the data has also been double checked in instances where there are reliable primary source data from government reports. New Zealand, for example, publishes annual reports on the military’s performance targets that describe the resources at the military’s disposal (Alexander, King, and Robert 2002). Although such data is not available for all countries nor for all years, checking the data where possible provides face validity about its accuracy.

However, being the most comprehensive data on military capabilities and from a highly reputable source does not guarantee accuracy. States themselves are uncertain about the military

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<sup>4</sup>Quotes come from [IISS Testimonials](#).

capabilities of others, and more pronounced uncertainty likely exists in the open source data used here (Kaplow and Gartzke 2021). Earlier years may have less accurate coverage as both private and government intelligence and surveillance efforts were less sophisticated, making it harder to know the capabilities of states that were far away (Lin-Greenberg and Milonopoulos 2021). New Zealand’s transparency in capability reports is an anomaly, as many countries – particularly those salient to contemporary international security like Russia, China, North Korea, and Iran – try to deceive their foes (or even friends) in misrepresenting their capabilities by exaggerating or underplaying their strength in particular platforms (Mastro 2016; Mawdsley 2016). Weapons systems also differ in whether open source intelligence is likely to identify them correctly, as capabilities whose effectiveness relies on secrecy and deception may not be known to the rest of the world (or rest of the government) until used in combat (Green and Long 2020). It is important that scholars acknowledge this limitation and, where possible, take steps to address it. For example, biases in GDP estimates have been identified using night-time-light data gathered from satellites, allowing scholars to quantify the extent to which autocratic regimes exaggerate their economic growth (Martínez 2022). In the future, similar measures could be applied to rDMC by, as has been done with Russia in the invasion of Ukraine, combining reports about military use and loss during combat to pre-war inventory estimates (Zhukov n.d.).

## 5 Data Collection and Formats

The first step of the data collection process creates a consistent typology of military equipment types, equipment names, and unit names. I create three versions of the data. The first, *rDMC raw*, organizes military equipment true to the original IISS categorizations. The second and third, *rDMC long* and *rDMC wide* produce a new more aggregated classification of military capabilities. The *rDMC* codebook defines the military capabilities included. Table 1 shows

the unique values that exist at each nested level, and how that is aggregated in the final classification. A detailed description of the three different data versions follows.

Table 1: Description of unit of analysis and variables in the different versions of the *rDMC* dataset.

Data Version	Unit of Analysis	N	Variables
Raw	Country-year-military unit	354,405	Service, disaggregated category, count
Long	Country-year-military technology	423,486	Aggregate category, count
Wide	Country-year	6,535	Technologies

## 5.1 *rDMC raw*

*rDMC raw* provides a version of the original data without subjective coding decisions or equipment categorizations, altered only by unambiguous data cleaning and process. IISS categorizes military capabilities in five descending levels: equipment types, equipment subtypes, equipment names, equipment subnames, and unit names.<sup>5</sup> An example is provided in Table 2. Equipment type involves the most aggregate categorizations (“aircraft” or “armored fighting vehicles”). Equipment names are the next primary IISS categorization (“transport” or “fighter” aircraft). Subtype and subname are auxiliary classifications that exist for some, but not all technologies (designations of “light”, “medium”, and “heavy” transport aircraft or “FFM” frigates with surface to air missiles). Lastly, the unit names identifies specific models (“M1A1 Abrams” tank). *rDMC raw* is the only version of the data that provides this unit-level information.<sup>6</sup> In this way, scholars can use the disaggregated data to identify, for example, how many countries have Aim-9 Sidewinder missiles in each year. Comparisons across country and year are possible, provided the research is careful in noting different naming conventions at

<sup>5</sup>While these 5 classifications levels are (somewhat) produced by IISS, their labels are the author’s.

<sup>6</sup>Standardizing string variables across unit names is a challenging endeavor that will be advanced in future iterations of the data. For more information about using the unit-level information in *rDMC raw*, see the appendix.

the unit level.<sup>7</sup>

Table 2: Equipment categories example from *rDMC raw*. Non-equipped category columns omitted.

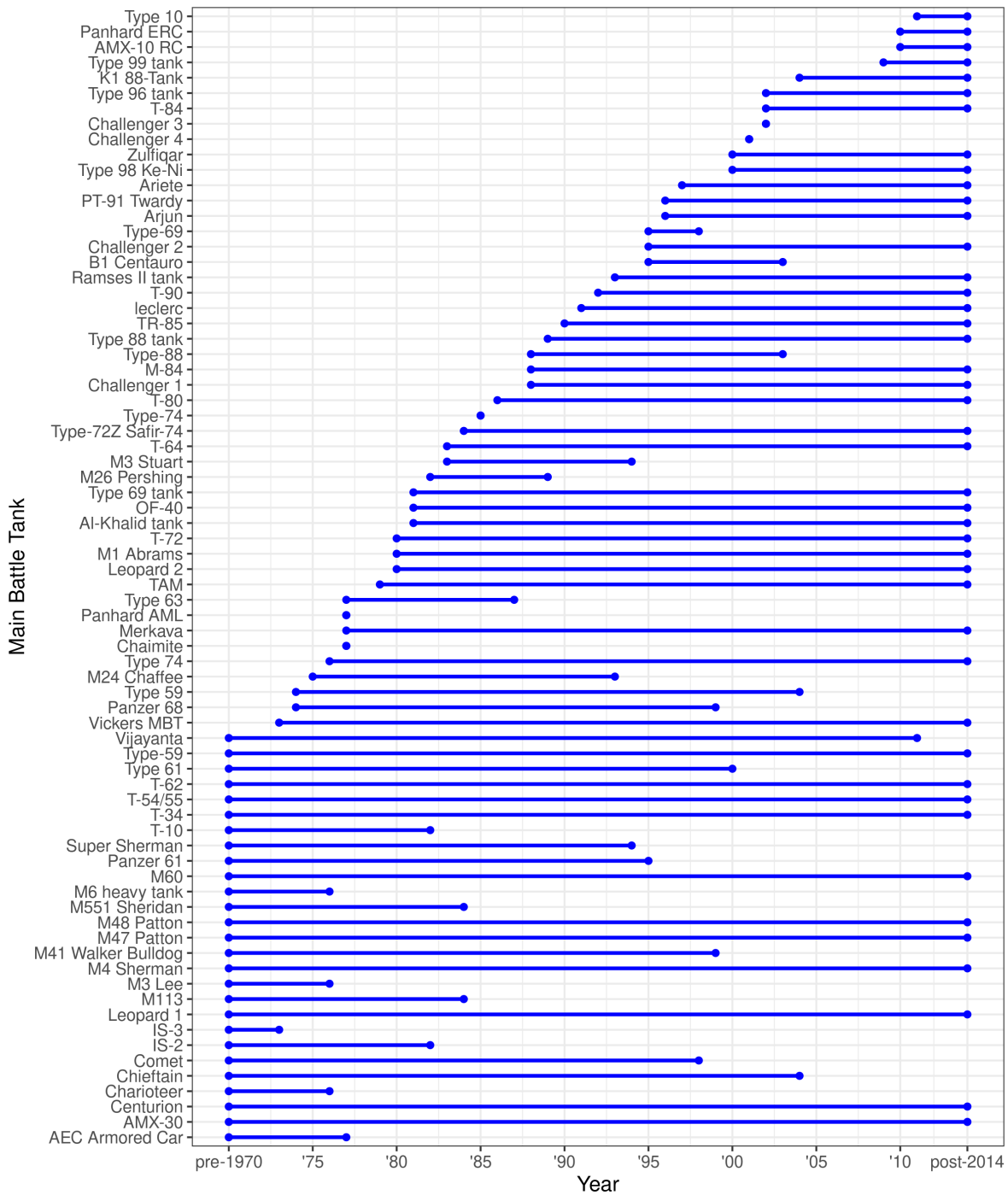
Type	Subtype	Name	Subname	Unit name
Aircraft	NA	TPT	light	Cessna 208B
Principal surface combatants	frigates	FFM	NA	Mourad Rais
Armoured fighting vehicles	NA	MBT	NA	M1A1 Abrams

*rDMC raw* can be used to examine questions concerning topics like combat effectiveness, arms sales, and interest group lobbying. Scholars have identified the challenges of identifying observable ex-ante proxies for the military effectiveness of particular platforms (R. A. Brooks and Stanley 2007). Biddle (2005, 135) uses the age of a military capability as a proxy for its effectiveness. During Operation Desert Storm, the average date of introduction for US weapons was 1974 while Iraq’s was 1962, with the assumption being that newer capabilities were more advanced. But until now, broader measures of the age of different components of a state’s military portfolio have remained un-examined. Figure 2 shows a broader comparison of the age of introduction for each main battle tank in the data as well as its last recorded year in service. The year of introduction is identified as the first year in which at least one state possessed that type of tank.<sup>8</sup> Scholars could broaden Biddle’s analysis beyond the Gulf War by identifying differences in predicted battlefield performance across generations of all types of technologies. This exercise also serves as a method of data validation, as the data are consistent with historical accounts concerning the development of many of these tanks like the M1 Abrams entering service in 1980 and the Japanese Type 10 being developed in 2012 (Ludeke 2018).

<sup>7</sup>For example, entries exist for the “AIM-9”, “AIM-9 Sidewinder”, “Aim-9”, and “Aim-9 Sidewinder”, which all refer to the same capability but with differences in capitalization and punctuation. *rDMC 2.0* will standardize the unit-level names, but the time-intensive nature of that effort means it will not be completed for some time.

<sup>8</sup>Since the data starts in 1970, tanks first deployed in 1970 most likely represent models developed before then. A more thorough analysis would track down the actual deployment date for each tank model, rather than relying on their deployment date as done here. Importantly, the data does distinguish between different

Main Battle Tanks: Years of Operation (1970-2014)



Source: IISS

Figure 2: The first year in which each type of main battle tank was deployed by any state. The figure is organized chronologically, with the newest main battle tanks at the top.

## 5.2 rDMC long and rDMC wide

*rDMC long* and *rDMC wide* provide a new, aggregated typology of national military capabilities.<sup>9</sup> The 69 categories that comprise the technologies are shown in Figure 3. These categories were chosen because they represent weapons categories commonly recognized and used by states in arms reduction agreements like the Treaty on Conventional Armed Forces in Europe (CFE). As a result, national records are most consistent and accurate at this level of analysis since those records were used during international negotiations. The resulting data provides a count of, for example, “aircraft – transport” for every country with a value that is the sum of all the rows in *rDMC raw* that had the original 5-level categorizations that match the new “aircraft - transport” category.

Empirically, this new typology is helpful because the technology categories are definitionally uniform across the data sample. Unlike *rDMC raw*, the new typology is standardized so it is consistent across country and year, thus simplifying time series-cross sectional comparisons. Where inconsistencies arise, transparent coding decisions were made with reference to external sources documented in the code repository. This ensure that, for example, the C-130H Hercules is always listed with an equipment type coding of ‘aircraft’ and an equipment name coding of ‘transport (TPT)’.<sup>10</sup> Aggregating the technology categories also reduces the sparseness of a dataset that is already zero-inflated. While most country-years possess armored fighting vehicles, not all possess every kind of armored fighting vehicle (main battle tanks, armored personnel carriers, armored infantry fighting vehicles, and reconnaissance vehicles) let alone

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platform upgrades, differentiating an F-16A from an F-16B in most cases.

<sup>9</sup>*rDMC long* and *rDMC wide* are identical in terms of scope and values and differ only in the unit of analysis. In *rDMC long* the unit of analysis is the country-year-technology while in *rDMC wide* the unit of analysis is the country-year. They are both provided to simplify the process of merging with existing datasets much like the Alliance Treaty Obligations and Provisions (ATOP) provides alliance data at various units of analysis (Leeds et al. 2002).

<sup>10</sup>There are cases where an equipment’s category changes in ways the data maintains. For example, many aircraft and helicopters are phased out by being shifted to non-combat roles like training before they are fully retired. A country may thus experience a decrease in combat aircraft and an increase in training aircraft from one year to the next without the actual aircraft they possess changing.

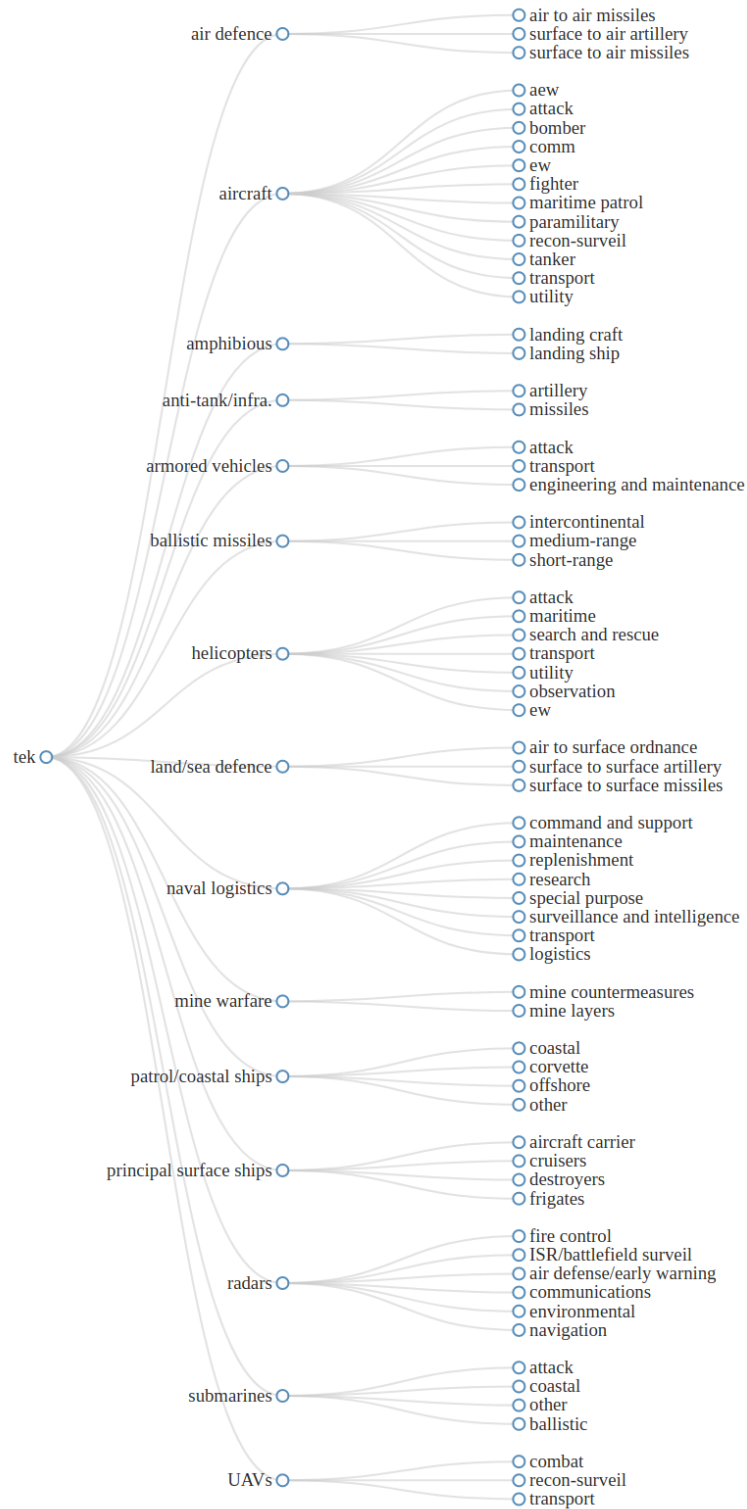


Figure 3: Description of the disaggregated technology categories used to compute the distribution of military capabilities.



each of the 1,759 distinct units categorized as “armored fighting vehicles”. That is not to say that every type or model is the same; but those distinctions present computational challenges for broad international relations questions. Scholars interested in making those distinctions are advised to use the *rDMC raw* version of the data.

There are, of course, many ways to categorize technologies depending on the research question. Disagreement over the categories produced by *rDMC long* and *rDMC wide* are remedied by the existence of *rDMC raw* that can be used by scholars to produce their own categorizations. “Aircraft – transport” and “helicopters – transport” could be considered somewhat interchangeable to those interested in a state’s ability to move personnel and material. Alternatively, “helicopters – transport” and “helicopters – search and rescue” may be reasonably combined if studying arms sales or military base location given their similar physical make-up. In other cases, categories could be further *disaggregated*. The category “aircraft – maritime patrol”, for example, include anti-submarine warfare, anti-surface warfare, and maritime reconnaissance which all “patrol” different areas of the sea. Rather than try to create and justify a single definitive ontology of military technologies, the data are constructed so that all aggregations are transparent and modular. By simply selecting new aggregation categories, scholars can produce their own counts with different categories to produce a new dataset consistent with classifications that suit their research question.

## 6 Global Trends

This section identifies some descriptive trends in military capabilities across time and space to highlight ways scholars can use these data in their own research.

## 6.1 Technological Trends Across Time

*rDMC* can shed light on topics like military diffusion and military effectiveness by providing empirical data about how militaries change over time (Horowitz 2010; Gilli and Gilli 2014). Research on military diffusion has argued that the rate at which weapons technologies spread to other states around the globe influences the likelihood of war (Bas and Coe 2012) as well as its outcomes (Zarzecki 2002). For example, much has been written about the consequences of aerial bombing (Pape 1996; Allen and Martinez Machain 2017). Although this research has looked at cases where states used aerial bombing, it does not identify what states have the capacity to conduct aerial bombing or how that changes over time.<sup>11</sup> Figure 4 shows annual changes in the total number of military aircraft as well as changes in the annual average across all states. The end of the Cold War in 1991 is an important turning point in both respects. The number of total military aircraft in the world dropped from roughly 40,000 to just over 20,000 and the average military aircraft per country went from around 400 to just under 200. Why such a significant reduction in combat aircraft worldwide? Part of the explanation may be the Adapted Treaty on Conventional Armed Forces in Europe (CFE) which required NATO and the Warsaw Pact – referred to in the CFE as the “Groups of States” – to each maintain no more than 6,800 combat aircraft (Bolvin 2000). This resulted in the destruction of 69,000 military units designated as Treaty Limited Equipment, with the Warsaw Pact destroying over 30% of its arsenal and NATO destroying 5% (McCausland 2012).

## 6.2 Technological Trends Across States

Military capabilities differ in their purpose. Some capabilities are most salient for states projecting power while others are relevant for territorial defense; some are better for winning and

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<sup>11</sup>New data on combat aircraft has recently been produced by Saunders and Souva (2020).

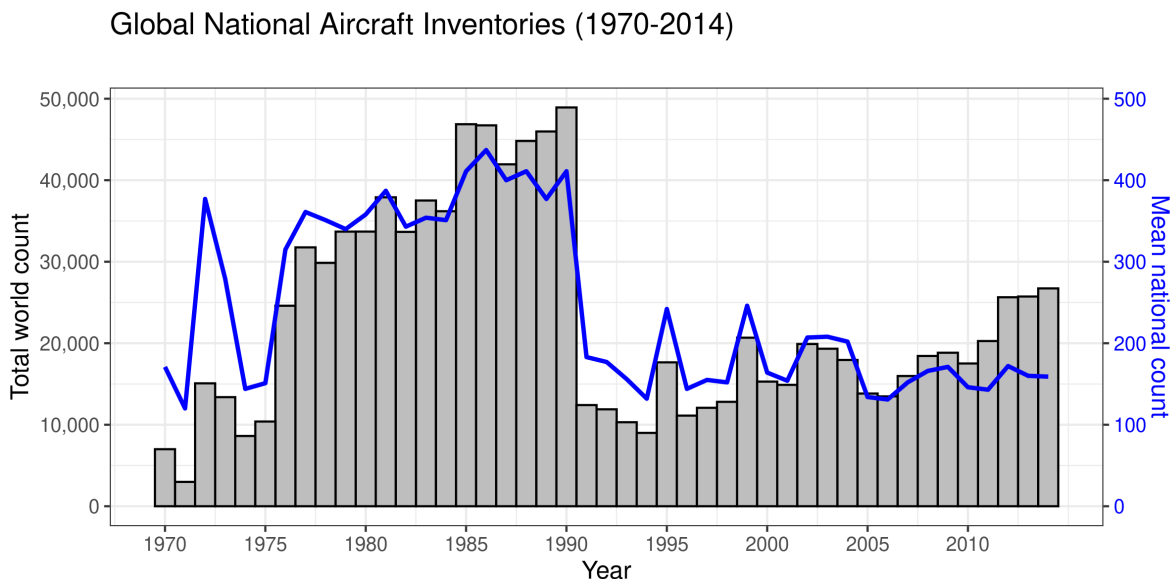
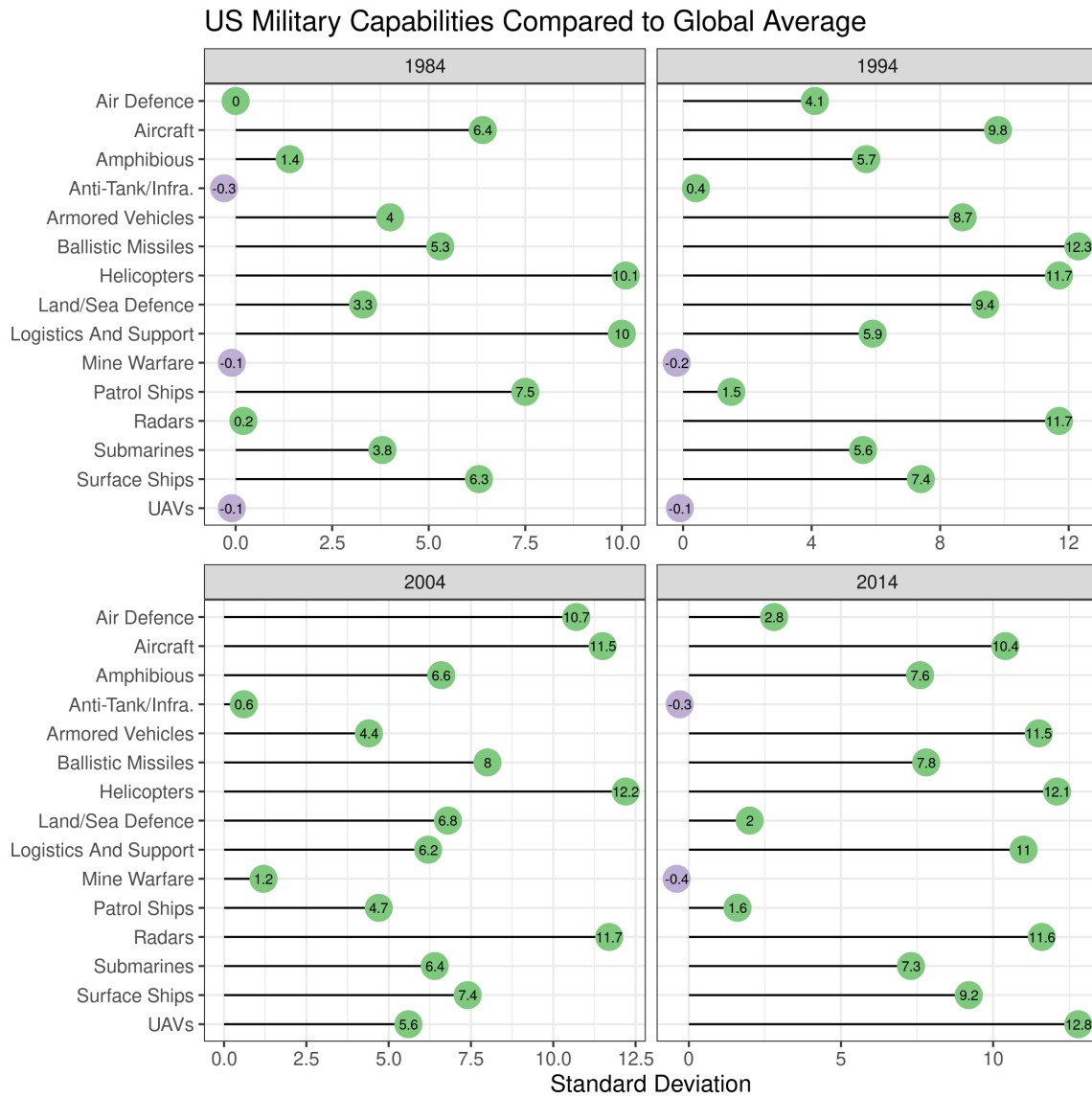


Figure 4: Bars (left y-axis) represent the total number of military aircraft in the world in each year. The blue line (right y-axis) represents the average number of aircraft owned by each national military in each year.

others better for warning (Lindsay and Gartzke 2022). All else being equal, what capabilities do states emphasize? Figure 5 shows the distribution of US military capabilities relative to the world average at decade intervals. Not surprisingly, US capabilities generally dwarf those of the rest of the international system. But this is not universal, nor is the degree of US dominance constant. For much of the past half century, the United States has had fewer anti-tank/anti-infrastructure capabilities as well as fewer mine warfare capabilities than the average state. The 2022 Russian invasion of Ukraine has made this fact especially salient, as transfers of one third of the US anti-tank/anti-infrastructure weapons stockpile to Ukraine has prompted concern about the sustainability of such support (Cancian 2022).

Variation in the US distribution of capabilities could be explained by myriad factors. Perhaps geography makes mine warfare less valuable since the US primarily fights its wars far from home. Military capabilities have substitutes and complements, and it's possible that anti-tank/anti-infrastructure needs are adequately addressed with bombing aircraft and land



Note: Bars represent standard deviations from the annual mean for each category.

Figure 5: Count of US military capabilities relative to other states in decade intervals. Each bar represents the standard deviation of the count of US capabilities relative to the world average. Colors represent positive and negative standard deviations.

defense missiles. Quantity is also not synonymous with quality, so it could be that the US is still more than sufficiently capable in these domains, but with fewer platforms. Whatever the theorized explanation, these new data now allows scholars to empirically examine claims about why states possess the distribution of military capabilities that they do.

When do states' military portfolios most closely mirror those of others? Some scholars argued socialization and competition under anarchy results in convergence, whereby states emulate the capabilities of the most powerful states in the international system (Waltz 1979, p 127; Resende-Santos 1996, p 196) while others have noted functional differentiation across allies who produce distinct, complementary military assets (Gannon 2021). Although there are instances of Cold War military planners falling prey to the homeopathy heuristic by embracing “missile matching” vis-à-vis their rivals (Kanwisher 1989), systematic evidence of similarity or difference across states' militaries has remained empirically untested. This claim is significant for scholars of international politics. If states mirror the military capabilities of others, then “new and proven military methods, even if they are truly revolutionary, will have no lasting affect on the balance of international influence” (Goldman and Andres 1999, p 83). This also has implications for what we know about security cooperation, with scholars cautioning that states imitating the military practices of their peers “rarely resort to alliances for their security” (Parent and Rosato 2015, p 52).

*rDMC* allows the similarity of militaries around the world to be identified and quantitatively measured. Figure 6 shows the results of a k-means clustering analysis for all states in 1994 using binary values indicating whether a state possessed each of the 69 military technologies.<sup>12</sup> A form of community detection, k-means clustering is an un-supervised learning algorithm that partitions the data into clusters of most-similar states using a gap statistic identifying the

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<sup>12</sup>I exclude states with a population below 750,000 as done by Eyre and Suchman (1996) and Sechser and Saunders (2010). Other states are missing data because they were not yet sovereign states (South Sudan) or because data on military capabilities was not collected for states just after or still in a serious civil conflict (Congo).

optimal number of groups (Tibshirani, Walther, and Hastie 2001).<sup>13</sup> There are eight distinct clusters of countries that share significant commonality in the military capabilities they possess. Some great powers are similar to one another (the US, Russia, France, and UK), but others appear more distinct (China and India). In some cases the US is similar to allies like Germany and Italy, but not with others like Poland and Canada. There is similarity across states in some geographic regions like Central Asia, but significant dissimilarity across states in other regions like the Pacific Rim.

Cautioning that Figure 6 is a simplified example (a single year and using binary rather than count values), the results nonetheless demonstrate the utility of disaggregating military capabilities for identifying similarity in national defense portfolios and generating testable hypotheses concerning the relationship between military composition and factors like technological capacity, geography, threat environment, and conflict history. The fact there are differences in how even similarly-sized states arm themselves is *prima facie* evidence of the non-fungibility of material military power.

### 6.3 Validation and Application: US and China Polarity

To validate the accuracy of the data as well as demonstrate how these data can advance existing scholarship, I use rDMC to replicate and temporally expand some of the key findings in S. G. Brooks and Wohlforth (2016) concerning the US-China balance of power. S. G. Brooks and Wohlforth (2016, 8) argue that the distribution of military capabilities provides evidence that “the United States will long remain the only state with the capability to be a superpower” and China’s military advancements have been and will remain slow, ensuring the United States retains a greater military advantage than superpowers have had in previous power transitions. This has remained a topic of great debate, with many scholars agreeing that China is unlikely

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<sup>13</sup>This clustering method has been used for community detection of economic industries (Delgado, Porter, and Stern 2016), democratization patterns (Gleditsch and Ward 2006), and arms sales (Akerman and Seim 2014).

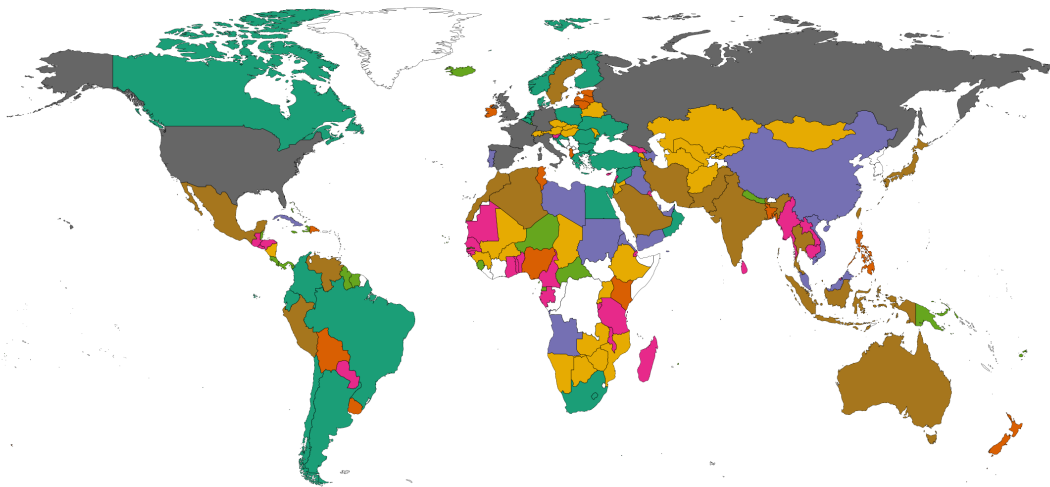
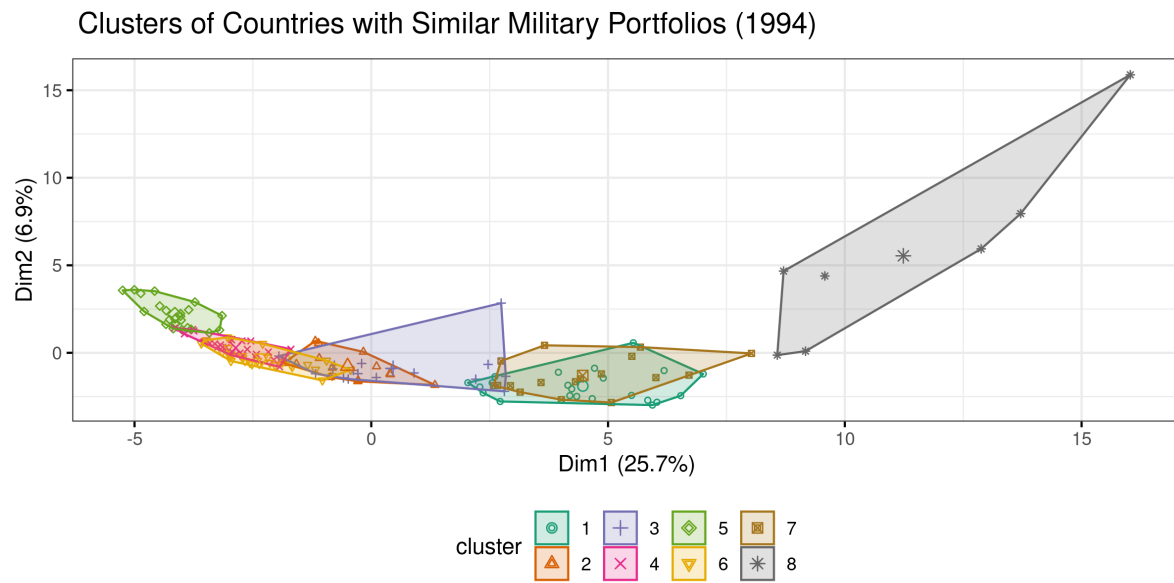


Figure 6: K-means clustering showing the similarity of military technologies portfolios.

to militarily match the United States (Cheung 2013; Christensen 2015; Beckley 2017; Gilli and Gilli 2019) and others taking a more pessimistic view that China’s military rise is rapid and threatens US interests (Yunzhuang 2012; Johnston 2013; Cliff 2015; Mastanduno 2019).

In measuring the distribution of power and identifying what makes a state a superpower, S. G. Brooks and Wohlforth (2016) argue that the three material components are military, economic, and technological capacity. In measuring military capacity, they recognize the shortfalls of aggregate measures like military spending and thus disaggregate Posen (2003)’s four categories for “command of the commons” and situate the main elements of power projection-capable military capacity in command of the sea (nuclear powered submarines, aircraft carriers, cruisers and destroyers, and principal amphibious ships), command of space (operating satellites and military satellites), command of the air (heavy unpiloted aerial vehicles, 4th generation tactical aircraft, 5th generation tactical aircraft, and attack helicopters) and infrastructure of command (airborne early warning and control, tanker and tanker/transport aircraft, heavy and medium transport helicopters, and heavy and medium transport aircraft).

To validate our data, Figure 7 shows, as a share of six great powers, how much of each of the command of the commons capabilities the US and China each possess.<sup>14</sup> The results are virtually identical to Figure 2 in S. G. Brooks and Wohlforth (2016), finding that the United States has a sizable lead over China in virtually every category. The only difference between our data and that in S. G. Brooks and Wohlforth (2016) beyond a rounding error concerns the relative share of 4th generation fighter aircraft where they find the US to have 55% of the great powers’ share whereas rDMC calculates it at 21%.<sup>15</sup>

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<sup>14</sup>The six great powers they consider are the United States, China, France, Russia, Great Britain, and India. The percentages for each military capability in the figures that follow are thus the share the US and China respectively possess relative to the total possessed by these six states.

<sup>15</sup>The discrepancy here seems to stem from the count of Russia’s 4th generation aircraft in 2013. rDMC seems to record a higher number than Brooks and Wohlforth, although their count for the other four great powers is not provided so this is conjecture. rDMC’s counts of great power 4th generation fighter aircraft were double checked relative to the original IISS data and the coding of the aircraft considered 4th generation was verified using Jane’s Yearbook and in consultation with uniformed personnel from the office of the under secretary of the Air Force.



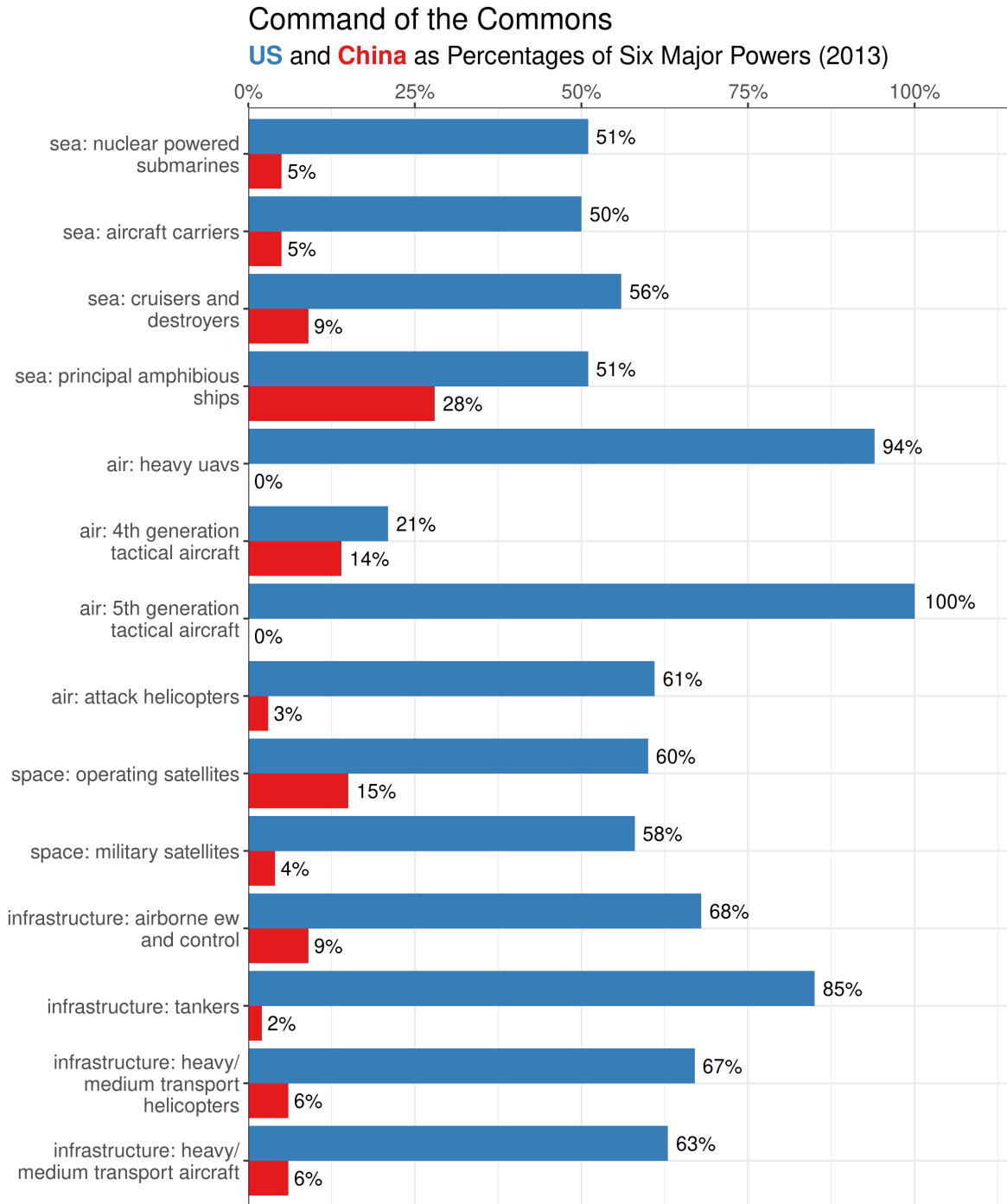


Figure 7: Replication of Figure 2 in Brooks and Wohlforth (2016) as data validation. rDMC produces an almost identical comparison between US and China military capabilities, with the exception of 4th generation fighter aircraft.

However, the claim that military gap between the US and China is “unprecedented in modern international relations” cannot be established from a single-year snapshot (S. G. Brooks and Wohlforth 2016, 22). Figure 8 uses rDMC to trace the temporal changes in US and China command of the commons capabilities from 1991-2014. Contrary to the claim that China’s military advancements have been slow, changes in the capability gap in some cases appear fast, in others slow, and in others reverses. While the US lead in infrastructure of command and command of the sea is as high as at any point in the post-Cold War era, the US lead in space has narrowed significantly and the US lead in air capabilities appears to be declining due to China’s recent heavy UAV and 4th generation fighter aircraft developments. Furthermore, China’s J-20 5th generation fighter aircraft did not hit low-rate initial production (LRIP) until 2015, so the dip in US dominance in command of the air precipitated by Chinese heavy UAV development is likely to continue rather than reverse course (Heath, Gunness, and Cooper III 2016; Mizokami 2016).<sup>16</sup>

This exercise serves to validate the accuracy of the rDMC data relative to another reputable source and to demonstrate how it can contribute to important debates in international relations. Far from the last word on China’s rise or the balance of capabilities between the US and other great powers, the data nonetheless suggest that the military capabilities gap between the United States and China may be narrowing, not widening, when the temporal lens is zoomed out beyond a single year and that the gap varies across both commons and capabilities. Whether this gap matters is another question, but one whose answer is better informed by proper measurement of its extent. Chinese innovation in satellites, radar, and air defense (Biddle and Oelrich 2016; Erickson et al. 2017) has different implications for the US and regional stability than their advancement in naval capabilities (Goldstein 2017; Cunningham 2020) or in drones (Zegart 2018; Calcara et al. 2022; Lin-Greenberg 2022). Using rDMC, these differences can be identified, explained, and analyzed by scholars and practitioners alike.

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<sup>16</sup>Although rDMC 1.0 only goes up to 2014, an auxiliary supplement extending the data to 2021 for China, Japan, Taiwan, South Korea, and Australia confirms this trend continues. See (redacted).

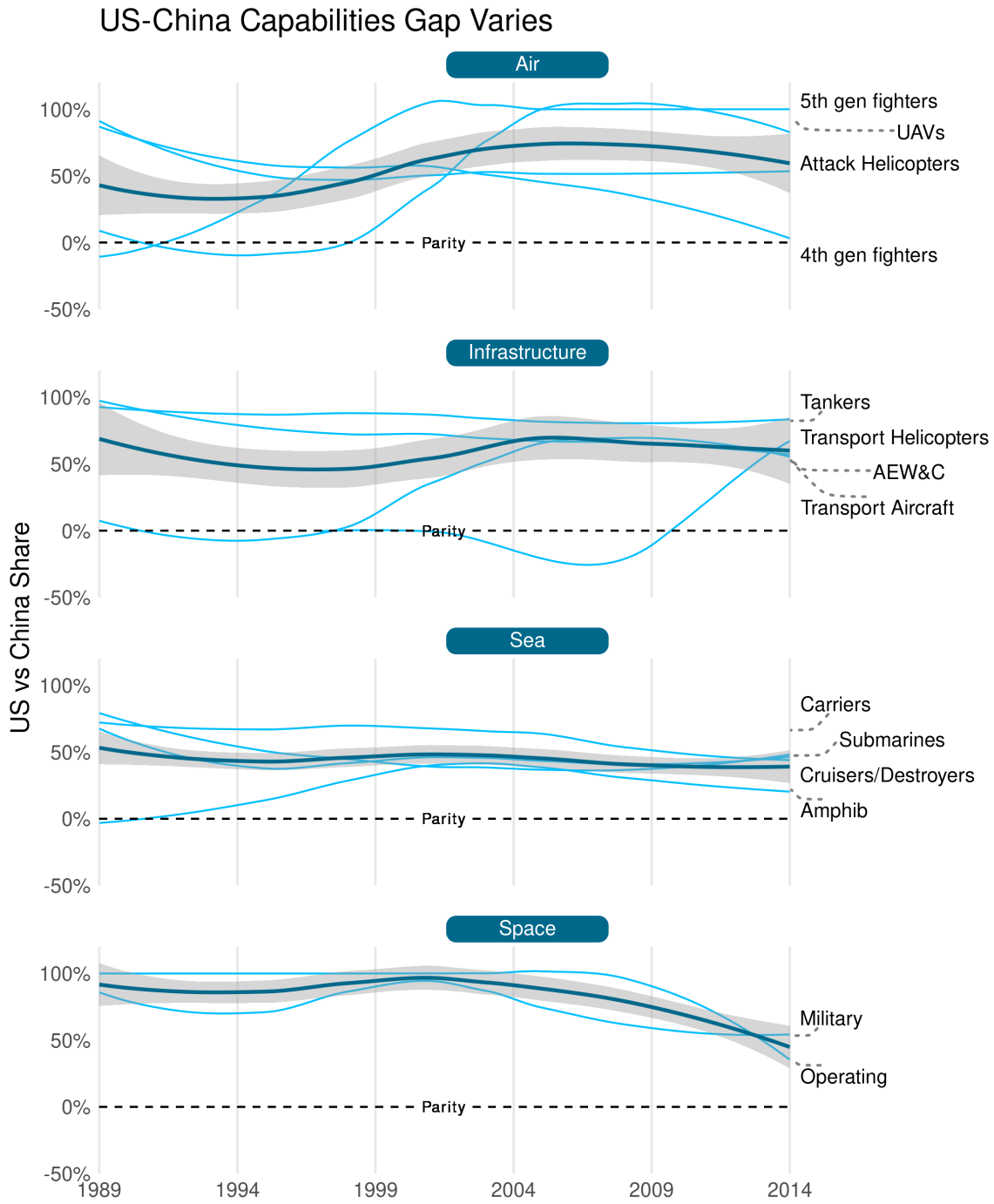


Figure 8: Positive (negative) y-axis values mean the US (China) has a higher share as a percent of six great powers. The dark lines represent the average for that common.

## 7 Conclusion

This paper introduced version 1.0 of a dataset that has demonstrated value in helping us understand how states fight; efforts are already underway to expand the data temporally – both beyond 2014 and further back than 1970, increase the ease of using *raw rDMC* by standardizing unit names across space and time, and diversify original source material beyond the IISS Military Balance. Even still, while much ink has been spilled debating the degree to which military technology matters, *rDMC* can help the discussion productively shift to *what* military technologies matter and *how*. Understanding differences in the composition of military capabilities is vital to understanding military power precisely because these components are not homogeneous. Identifying the dimensions of this heterogeneity is a necessary precondition for explaining why states develop the weapons portfolio that they do as well as the consequences of armament decisions. The composition of military assets, rather than the defense dollars spent, are what truly matter for how well states deal with threats to their security because military spending itself does not create military power. Rather, that money must be translated into capabilities that allow for the exercise of power through a variety of distinct means.

To date, explanations of the causes or effects of variation in the composition of a state’s military assets has been empirically limited because that data has not existed in a form conducive to systematic analysis. As an easier to use and more structured disaggregated dataset on the military capabilities states possessed over the past half century, *rDMC* can help the broader scholarly community better explain how states arm, why, and to what effect.

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