China's Energy Diplomacy:

Does Chinese Foreign Policy Favor Oil-Producing Countries?



Lee, Chia-yi (2019)

Foreign Policy Analysis, 15(4), 570–588

https://doi.org/10.1093/fpa/orz011



Replication Study
Applied Statistical Analysis II

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April 04, 2024

Original Study: Overview

- 'Due to its rapid economic growth and increasing demand for energy, China has engaged in numerous efforts to sustain its energy supplies and enhance its energy security'.
- How does China's energy concern affect its foreign policy preferences (esp. in Africa)?
- Dependent variables: three foreign policy instruments:
 - o **partnerships** (p., strategic p., comprehensive strategic p.);
 - o **foreign aid** (\$ (millions, logged), 2000-2013);
 - o leadership visits (1998-2013).
- Independent variables:
 - o country's annual **oil production,** barrels (thousands);
 - o country's oil reserves, barrels (thousands)

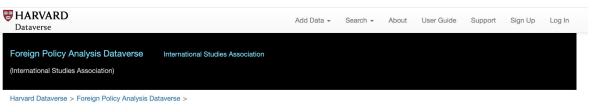






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Original Study: Data & Replication



Replication data for "China's Energy Diplomacy: Does Chinese Foreign Policy Favor Oil Producing Countries?"



Lee, Chia-yi, 2023, "Replication data for "China's Energy Diplomacy: Does Chinese Foreign Policy Favor Oil Producing Countries?", https://doi.org/10.7910/DVN/7E3Q5P, Harvard Dataverse, V1, UNF:6:x+exuF4WVfkngWChUWz4MQ= [fileUNF]

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51 obs. of 15 variables	African states (51)
159 obs. of 21 variables	UN states (159), -OECD states (129)
3332 obs. of 17 variables	UN states (162)
160 obs. of 18 variables	UN states (160)
	159 obs. of 21 variables 3332 obs. of 17 variables

Original Study: Main Findings

- 1. H: China is more likely to build partnerships with countries abundant in energy resources. ✓
- 2. H: China is more likely **to give aid** to countries abundant in energy resources. ✓
- 3. H: Top Chinese leaders are more likely to pay leadership visits to countries abundant in energy resources. ✓

Table 2. Energy production and Chinese foreign policy preferences

Dependent variable	Partnership Model 1	Strategic partnership Model 2	Chinese ai	d to Africa Model 4	Leadership visits Model 5
Oil production	0.131	0.184	0.050	0.316	0.074
	(0.058)**	(0.069)***	(0.080)	(0.151)**	(0.025)***
GDP per capita	-0.335	-0.525	-0.108	-0.673	-0.232
	(0.270)	(0.322)	(0.356)	(0.683)	(0.110)**
GDP growth	-0.033	-0.003	0.041	0.068	0.009
	(0.089)	(0.088)	(0.035)	(0.170)	(0.034)
FDI inflows	0.261	0.305	0.129	0.090	0.121
	(0.081)***	(0.096)***	(0.235)	(0.134)	(0.031)***
Trade importance	1.302	1.825	3.186	2.582	1.194
	(0.826)	(0.879)**	(2.096)	(5.063)	(0.249)***
Level of democracy	-0.007	-0.046	0.145	0.298	-0.025
	(0.051)	(0.059)	(0.066)**	(0.108)***	(0.021)
Domestic conflict	0.201	0.166	0.111	0.222	-0.030
	(0.138)	(0.159)	(0.064)*	(0.327)	(0.061)
US ally	-0.921	-0.505	0.734	-1.132	-0.099
	(0.777)	(0.891)	(2.704)	(3.435)	(0.315)
Lagged Chinese aid			0.203		
			(0.040)***		
Number of observations	125	125	603	49	125
Number of countries	125	125	49	49	125
Log likelihood	-55.55	-44.35	-1778.55	-123.23	-191.22
AIC	129.09	106.70	3583.10	266.46	400.44
BIC	154.55	132.16	3640.33	285.38	428.72

Table 3. Energy production and Chinese foreign policy preferences: robustness checks

Dependent variable	Partnership	Strategic Partnership partnership		Chinese aid to Africa		Number of days spent on leadership visits	
	Model 6	Model 7	Model 8	Model 9	Model 10	Model 11	Model 12
Oil reserves	0.106	0.137	0.016	0.273	0.055		0.051
	(0.050)**	(0.059)**	(0.072)	(0.119)**	(0.022)***		(0.028)*
Oil production						0.073 (0.033)**	
GDP per capita	-0.321	-0.462	-0.143	-0.615	-0.209	-0.196	-0.164
	(0.268)	(0.310)	(0.353)	(0.646)	(0.112)*	(0.145)	(0.146)
GDP growth	-0.027	0.004	0.050	0.067	0.014	0.012	0.019
	(0.092)	(0.092)	(0.036)	(0.168)	(0.035)	(0.046)	(0.046)
FDI inflows	0.261	0.306	0.149	0.131	0.122	0.149	0.149
	(0.080)***	(0.095)***	(0.240)	(0.136)	(0.031)***	(0.038)***	(0.039)***
Trade importance	1.327	1.830	3.574	3.122	1.235	1.558	1.645
•	(0.823)	(0.873)**	(2.112)*	(4.890)	(0.252)***	(0.392)***	(0.396)***
Level of democracy	-0.005	-0.050	0.125	0.301	-0.024	-0.020	-0.019
	(0.052)	(0.060)	(0.067)*	(0.106)***	(0.022)	(0.028)	(0.029)
Domestic conflict	0.198	0.179	0.104	0.206	-0.024	-0.019	-0.014
	(0.139)	(0.157)	(0.065)	(0.322)	(0.062)	(0.080)	(0.081)
US ally	-0.914	-0.475	0.579	-0.999	-0.127	-0.320	-0.352
*	(0.772)	(0.871)	(2.683)	(3.402)	(0.318)	(0.406)	(0.409)
Lagged Chinese aid			0.196				
			(0.040)***				
Number of observations	125	125	588	49	125	125	125
Number of countries	125	125	49	49	125	125	125
Log likelihood	-55.85	-45.33	-1736.43	-122.76	-192.34	-281.43	-282.20
AIC	129.69	108.66	3498.87	265.51	402.69	580.86	582.40
BIC	155.14	134.11	3555.76	284.43	430.97	609.14	610.69

Notes: Standard errors are in parentheses. *p < .1; **p < .05; ***p < .01.

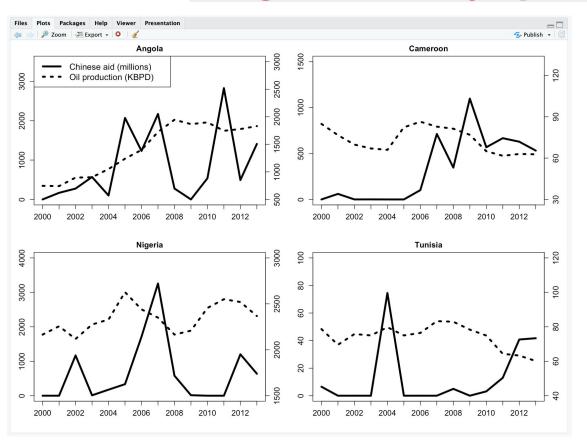
Notes: Standard errors are in parentheses. *p < .1; **p < .05; ***p < .01.

*Original Study: Main Findings (repl.)

			Table 2	lent variable:	
	partner	partner2	Inaid	lnaid2	visit
		logistic	linear	OLS	negative
			mixed-effects		binomial
	(1)	(2)	(3)	(4)	(5)
prod2	0.131**	0.184***		0.316**	0.074***
		(0.069)		(0.151)	(0.025)
GDPpc2	-0.335	-0.525		-0.673	-0.232**
	(0.270)	(0.322)		(0.683)	(0.110)
growth2	-0.033	-0.003		0.068	0.009
	(0.089)	(0.088)		(0.170)	(0.034)
FDI2	0.261***	0.305***		0.090	0.121***
	(0.081)	(0.096)		(0.134)	(0.031)
trade.de2	1.302	1.825**		2.582	1.194***
uuc.dez	(0.826)	(0.879)		(5.063)	(0.249)
polity2	-0.007	-0.046		0.298***	-0.025
ponty2	(0.051)			(0.108)	(0.021)
dom2	0.201	0.166		0.222	-0.030
dom2	(0.138)	(0.159)		(0.327)	(0.061)
usallv2	-0.921	-0.505		-1.132	-0.099
usany2		(0.891)		(3.435)	(0.315)
lnaid.1	(01111)	(0.051)	0.203***	(0.100)	(0.0.10)
maid.1			(0.040)		
Inprod1			0.050		
inprour			(0.080)		
InGDPpc.1			-0.108		
mobi pc.i			(0.356)		
growth.1			0.041		
gio main			(0.035)		
I(lnFDI.1/10)			0.129		
.,			(0.235)		
trade.de.1			3.186		
			(2.096)		
polity.1			0.145**		
. ,			(0.066)		
Indom.1			0.111*		
indom:			(0.064)		
usally.1			0.734		
usury.r			(2.704)		
Constant	4.416**	-4.817*	5.289**	13.537***	-0.810
Constant	(2.226)	(2.611)	(2.380)	(4.767)	(0.939)
Observations	125	125	603	49	125
R ²				0.314	
Adjusted R ²		44.252	1 220 551	0.176	101.010
Log Likelihood	-35.546	-44.352	-1,778.551		-191.219
theta	120.001	106 701	3,583.102		2.152*** (0.81) 400.438
Akaike Inf. Crit. Bavesian Inf. Crit.	129.091	106.704	3,583.102		400,438
Residual Std. Error			5,040.527	3.312 (df = 40)	
F Statistic				2.284** (df = 8; 40)	
Note:				*p<0.1; **r	•

				Table 3			
				Dependent	variable:		
	partner	partner2	Inaid	lnaid2	visit	d	ay
	logistic	logistic	linear	OLS	negative		ative
	(1)		mixed-effect.		binomial		mial
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
reserve2	0.106**			0.273**	0.055**		0.051*
	(0.050)	(0.059)		(0.119)	(0.022)		(0.028)
prod2						0.073**	
						(0.033)	
GDPpc2	-0.321	-0.462		-0.615	-0.209*	-0.196	-0.164
	(0.268)	(0.310)		(0.646)	(0.112)	(0.145)	(0.146)
growth2	-0.027	0.004		0.067	0.014	0.012	0.019
		(0.092)		(0.168)	(0.035)	(0.046)	(0.046)
FDI2		0.306***		0.131	0.122***	0.149***	0.149***
	(0.080)	(0.095)		(0.136)	(0.031)	(0.038)	(0.039)
trade.de2	1.327	1.830**		3.122	1.235***	1.558***	1.645***
	(0.823)	(0.873)		(4.890)	(0.252)	(0.392)	(0.396)
polity2	-0.005	-0.050		0.301***	-0.024	-0.020	-0.019
	(0.052)	(0.060)		(0.106)	(0.022)	(0.028)	(0.029)
dom2	0.198	0.179		0.206	-0.024	-0.019	-0.014
	(0.139)	(0.157)		(0.322)	(0.062)	(0.080)	(0.081)
usally2	-0.914	-0.475		-0.999	-0.127	-0.320	-0.352
	(0.772)	(0.871)		(3.402)	(0.318)	(0.406)	(0.409)
lnaid.1			0.196***				
			(0.040)				
Inreserve1			0.016				
			(0.072)				
lnGDPpc.1			-0.143				
			(0.353)				
growth.1			(0.050				
Id-PDI 100							
I(lnFDI.1/10)			(0.240)				
trade.de.1							
trade.de.1			3.574* (2.112)				
polity.1			0.125*				
			(0.067)				
lndom.l			(0.065)				
usally.1			0.579				
usany.r			(2.683)				
Constant	4.520**	-5.229**	5.678**	12.359***	-0.975	-0.727	-0.939
Constitut		(2.571)	(2.359)	(4.441)	(0.944)	(1.193)	(1.196)
Observations	125	125	588	49	125	125	125
Observations R ²	123	143	200	0.327	123	143	123
Adjusted R ²				0.192			
Adjusted R ^a Log Likelihood	-55 845	-45.328	-1,736.433	0.192	-192.343	-281.428	-282.201
theta	-33.043	43.320	-1,730,433			-281.428 0.564*** (0.114)	
Akaike Inf. Crit.	129.689	108.656	3,498.866		402.687	580.856	582,402
Bayesian Inf. Crit.	-27.007	_ 50.050	3,555.764		1021007	500,050	5021102
Residual Std. Error				3.280 (df = 40)			
F Statistic				2.427** (df = 8; 40)			

Original Study: (Tiny) Twist 1

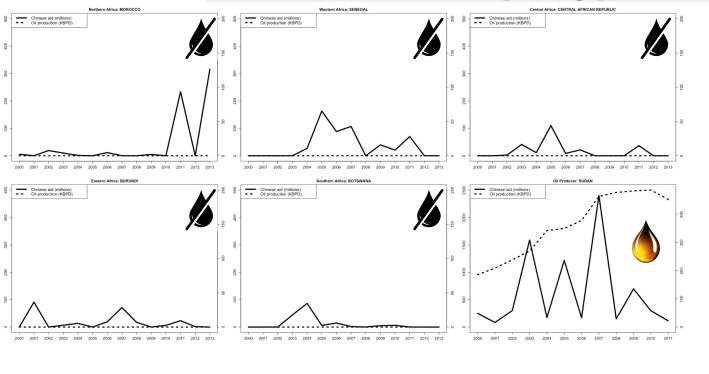


'I select four African oil-producing countries that have received Chinese aid, [...] from 2000 to 2013 [...]. **Chinese aid** (denoted by the solid line) **fluctuates** from year to year, whereas oil production (denoted by the dashed line) is relatively stable over time'.

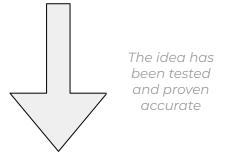
'In other words, if the time-serial variation is taken into account, the results may be confounded by **the volatility of Chinese aid.** Because the annual aid amount is subject to the Chinese government's yearly budget, it is less likely that it will clearly follow aid recipients' oil production every year'.

'When I only focus on the <u>variation across</u> <u>African countries</u>, however, it is observed that **African oil producers in general receive more Chinese aid than non-oil producers'.**

Contribution: (Tiny) Twist 1



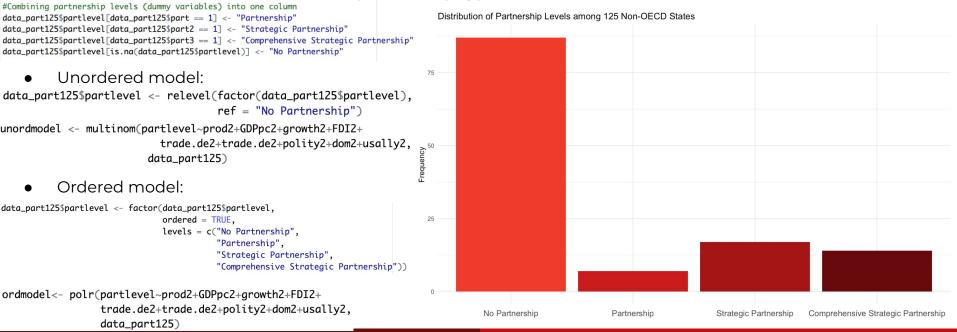
- 5 non-oil producing states from each of the 5 African regions: Morocco, Senegal, Central African Republic, Burundi, Botswana
- One additional oil-producer: Sudan



If focused on the <u>variation across African countries</u>, it is observed that **African non-oil producers in general receive less Chinese aid than oil producers**

Contribution: Ordered/Unordered Model for Partnerships

- Original Model 1: binomial ordered logit, DV: Partnership (0 for no partnership, 1 for any kind of partnership)
- $model1 <- glm(partner \sim prod2 + GDPpc2 + growth2 + FDI2 + trade.de2 + polity2 + dom2 + usally2, family = binomial, data = data_part125)$
 - But the dataset includes more detailed data on the types of partnerships...
 - ... Fit ordered/unordered models for partnership types. Is any of them more suitable and insightful than Model 1?



Contribution: Ordered/Unordered Model for Partnerships

Ordered Model (eyn)

Unordered Model (eyn)

Original Model 1 (eyn)

Origina	ii Model I (exp.)		Unordered Model	(exp.)		Ordered Model (exp.)	
	Dependent variable:		Dependen	t variable:			Dependent variable:
	Partnership		${\bf Comprehensive\ Strategic\ Partnership}$	Partnership	Strategic Partnership		Partnership Level
Oil production	1.139**		(1)	(2)	(3)	Oil production	1.124**
	(0.058)	Oil production	1.221**	1.013	1.197**		(0.057)
GDP per capita	0.715		(0.104)	(0.105)	(0.083)	GDP per capita	0.694
	(0.270)	GDP per capita	0.347**	1.004	0.756		(0.256)
GDP growth	0.968		(0.520)	(0.490)	(0.363)	GDP growth	0.958
GDI GIOWIII	(0.089)	GDP growth	1.003	0.864	0.980	de la comunicación de la comunic	(0.086)
	,	0-1 0-1	(0.157)	(0.227)	(0.103)		
FDI inflows	1.299***					FDI inflows	(1.324***
	(0.081)	FDI inflows	1.885**	1.133	1.266**		(0.078)
Trade importance	3.678		(0.284)	(0.148)	(0.097)	Trade importance	6.095**
Trade importance	(0.826)	Trade importance	10.218**	0.153	2.253	rrade importance	(0.745)
	(0.020)	Trade imperation	(1.189)	(4.021)	(1.083)		(0.113)
Level of democracy	0.993					Level of democracy	0.953
	(0.051)	Level of democracy	0.944	1.095	0.973		(0.050)
D	1.000		(0.087)	(0.106)	(0.067)	D at at	
Domestic conflict	1.223 (0.138)	Domestic conflict	0.957	1.308	1.337	Domestic conflict	1.155 (0.131)
	(0.138)	Domestic connec	(0.239)	(0.236)	(0.177)		(0.131)
US ally	0.398		()	()	()	US ally	0.841
300000	(0.777)	US ally	1.536	0.230	0.288		(0.678)
			(1.268)	(1.281)	(1.047)		, ,
Constant	0.012**	Constant	0.000	0.008	0.003**	Intercepts	
	(2.226)	Constant	(5.576)	(3.694)	(2.928)	No Partnership—Partnership	77.138**
-	122		(* * * * *)	()	()	Partnership—Strategic Partnership	118.238**
Observations Log Likelihood	125 -55.546	Observations	125			Strategic Partnership—Comprehensive Strategic Partnership	513.783**
Residual Deviance	-55.546 111.09	Log Likelihood	-83.263			Observations	125
AIC	129.091	Residual Deviance	166.526			Log Likelihood Residual Deviance	-90.19376 180.3875
BIC	154.5459	AIC	220.526			AIC	202.3875
Note:	*p<0.1; **p<0.05; ***p<0.01	BIC	296.89			BIC	233.499
	F 10.2, P 10.00, P 10.01	Note:		*p<0	1.1; **p<0.05; ***p<0.01	Note:	*p<0.1; **p<0.05; ***p<0.01
						1	

Contribution: Ordered/Unordered Model for Partnerships

	LogLik	Residual Deviance	AIC	BIC	Brant Test library(brant) brant(ordmodel)	Accuracy Score confusionMatrix (predictions, data\$y)
Original Model 1	-55.546	111.09	129.091	154.546		78.4%
Unordered Model	-83.263	166.526	220.526	296.89		74.4%
Ordered Model	-90.193	180.386	202.388	233.499	PRA holds	75.2%

- New models confirmed the initial findings that China is **more likely to build partnerships** with countries **abundant in energy resources** + it is also driven by **economic considerations** (significant coefficients for *Oil production*, *FDI inflows*, *Trade importance*).
- Both new models performed <u>reasonably efficiently</u> as compared to the initial binomial model, though the **ordered** model achieved a higher accuracy score and coefficients more similar in value.
- Parallel Regression Assumption (PRA) holds for the **ordered model,** which makes it a more adequate fit for the data.
- Interesting experiment overall, BUT: **the original binomial logit Model 1 is still the most accurate** → we might assume that being China's partner versus not being China's partner at all carries greater significance compared to being involved in a specific type of partnership.