



Session 1. Identifying Corporation Tax Avoidance

Moderator: Eric Toder

Tax Policy Center

Using IRS Data to Identify Income Shifting Lisa De Simone

Firms Stanford University

Income Shifting by U.S. Multinational

Corporations

Amy Dunbar

University of Connecticut

Corporations Onliversity of Confidential

The Economic Effects of Special Purpose Petro Lisowsky

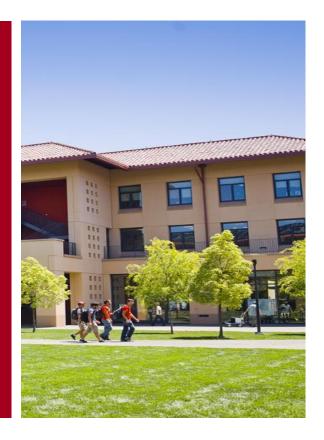
Entities on Corporate Tax Avoidance University of Illinois at Urbana-Champaign

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Eric Toder Tax Policy Center

Using IRS Data to Identify Income Shifting Firms

Lisa De Simone Stanford Graduate School of Business

June 21, 2017





Overview

What

 Measure the likelihood that a U.S. multinational entity (MNE) shifts income out of the U.S. using intercompany transactions with foreign subsidiaries

Why

- Increased international attention on income shifting: important to understand (i) magnitude, (ii) what types of firms shift income out of the U.S., and (iii) consequences
- Identifying income shifting is difficult
- Findings could inform potential cost/benefit of proposed tax reform that would alter income shifting incentives

How

- Measure net outbound intercompany transfers using Form 5471 Schedule M
- Develop a prediction model of net outbound shifting
- Examine audit outcomes of net outbound firms and firms that shift more out (or less in) than expected

Key Results

Likelihood of net outbound income shifting via intercompany transactions

- Positively associated with tax haven operations, high tech industry membership, tax incentives, R&D, and foreign profitability
- Negatively associated with high percentage of foreign sales, gross profits, size, and capital expenditures
- Holdout sample tests to validate model

"Aggressive" income shifting

- Defined as having a positive residual in a continuous OLS model: shift more out (or less in) than expected (i.e., exhibit a higher continuous net outbound amount than expected)
- Positively correlated with net outbound income shifting
- On average, 45% of sample firm-years shift more out (or less in) than predicted

Likelihood of audit

Net outbound or aggressive income shifters are not more likely to be audited

Predicting Net Outbound Income Shifting via Inter-Company Transactions

Factors	Why?	Proxies
Intangible intensity	Easy to migrate IP; Allows for royalties and other IC payments	R&D, Advertising, SG&A, Capitalized intangibles, Capex
Unique offerings	Greater latitude in setting prices for IC transactions	GP%, High-tech industry membership
Global footprint	Support for presence of economic activity abroad	% Foreign sales, Dom/for ROS, Dom/For growth, Haven operations
Tax incentives	Incentive to shift to lower-tax jurisdictions	Foreign effective tax rate differential (lagged)
Debt	Alternative tax shields	Leverage, Interest
Tax Planning	Ability to expend resources to shift effectively	Size, BigN auditor

Sample

Firm-years in IRS Business Returns Transaction File 2005-2014	351,843
Less:	
Observations without a matched Form 5471M	(306,379)
Observations with zero or missing Compustat SALE	(32,624)
Observations in a financial industry	(125)
Observations missing required data for estimation	(5,633)
Observations where FTR outside [-1,1]	(500)
Sample used for net outbound income shifting likelihood model	6,582
Less:	
Unable to match to IRS Audit Information Management System	(2,260)
Sample used for audit likelihood model	4,322

Descriptive Statistics by Net Outbound Income Shifting

		OutShif (n = 2,		OutShift (n = 4,		Differ	ences	
	<u>Variable</u>	<u>Mean</u>	<u>Median</u>	<u>Mean</u>	<u>Median</u>	<u>Mean</u>	<u>Median</u>	Net inbound shifting on
	OutShift	0.074	0.033	(0.057)	(0.026)	0.131 ***	0.059 ***	average
	Sales	6,078	964.9	5,623	1,533	455.3	(568.1) ***	average
	Pre-tax Income	576.7	52.65	630.5	98.88	(53.84)	(46.22) ***	
	ETR	0.215	0.238	0.249	0.281	(0.034)***	(0.043) ***	
	R&D	0.075	0.024	0.051	0.014	0.024 ***	0.011 ***	
	AD	0.009	-	0.012	-	(0.003) ***	0.000	More R&D
Intangible -	SG&A	0.293	0.236	0.271	0.225	0.022 ***	0.011	Less capex
Intensity	Intangibles	0.033	-	0.031	-	0.003	0.000	
·	Capex	0.041	0.026	0.052	0.028	(0.012) ***	(0.001) **	
Unique Offerings	GP%	0.418	0.380	0.422	0.389	(0.005)	(0.009)	
Ornque Oriennigo	HighTech	0.401	-	0.255	-	0.146 ***	0.000	
	ForeignSales%	0.402	0.380	0.424	0.417	(0.023) ***	(0.037) ***	Less foreign sales %,
	FROS	0.034	0.021	0.037	0.028	(0.003)*	(0.006) ***	return on sales
Global Footprint	DROS	0.009	0.024	0.022	0.035	(0.013) ***	(0.012) ***	return on sales
Global i ootpillit	FSalesGrowth	0.196	0.095	0.165	0.086	0.031 ***	0.009 **	
	DSalesGrowth	0.106	0.067	0.090	0.058	0.016 ***	0.009 ***	More growth
	HasHaven	0.809	1.000	0.807	1.000	0.002	0.000	
Tax	Lag_FTR	0.144	0.146	0.115	0.108	0.029 ***	0.038 ***	More tax incentives to
Incentive _{bt} =	Leverage	0.190	0.052	0.232	0.102	(0.043) ***	(0.050) ***	shift
Debt	Interest	0.014	0.005	0.016	0.008	(0.003) ***	(0.003) ***	
Tax Planning	Size	20.75	20.69	21.10	21.15	(0.348)***	(0.463) ***	Less debt, smaller
Tax Flatilling	BigN	0.840	1.000	0.867	1.000	(0.028)***	0.000	•

Work in Process: Additional Descriptive Analysis

By 5471M Line Pair

 Net inbound shifting at mean (median) except for CSAs, Services, Commissions (IP, CSAs, Commissions, Insurance)

By Size

 Smallest asset quartile (~< \$250M) is only quartile with net outbound income shifting on average

By Industry

 High Tech and Fama-French 12 "Other" (Mines, Construction, Building Materials, Transportation, Hotels, Business Services, Entertainment) are only industries with net outbound income shifting on average

By Year

Net shifting increasing over time; net outbound at mean in 2012 and 2014

Prediction Model: Likelihood of OutShifter = 1 (Net Outbound Shifting)

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			<u>(1)</u>	<u>(2)</u>	(3): OLS
	RD	+	2.5464***	2.2769***	-0.0549***
Intongible	AD	+	-2.2679	-2.6301	-0.0602
Intensity	Intangibles	+	0.4484	0.4984*	0.0076
L	SGA	+	-0.3623	-0.4342	0.0286**
Intangible Intensity	-0.0193				
}	GP%	+	-1.166***	-1.0589***	-0.0819***
	HighTech	+	0.6274***	0.6276***	0.0382***
Global Footprint	ForeignSales%	+	-0.9729***	-0.9083***	-0.0881***
	Lag_FROS	+	1.9507***	1.5106	0.1130***
L	Lag_DROS	?	0.0357	0.044	-0.0357***
	FSalesGrowth	+	0.0545	0.0541	0.0017
Debt -	DSalesGrowth	?	0.1329	0.1969	0.0035
Tay Planning	HasHaven	+	0.2964**	0.3109**	0.0139***
Tax Flaming 1	LagFTR	+	0.4042***	0.4869**	0.0078
	Leverage	?	-0.2335	-0.1807	-0.0067
	Interest	?	1.0400	0.2744	0.0249
Stanford Graduate School	Size	?	-0.1016***	-0.1022**	-0.0035***
	Big5	+	0.0648	0.1042	-0.0092**

Likelihood of Net Outbound Income Shifting:

- + R&D, HighTech, foreign return on sales, has haven operations, foreign tax rate differential
- Capex, gross profit %, foreign sales %, size

Predictive Power

AUCs low but improving in revision (~0.70+)

Holdout sample analysis

Aggressive Shifters

If positive residual in OLS model, shifting more out (or less in) than predicted

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Additional Tests: Firms that Shift More Out (Less in) than Expected

Aggressive Shifters vs. Net Outbound Income Shifters

	OutShifter = 0	OutShifter = 1	<u>Total</u>
AggShifter = 0	3,130	385	3,515
AggShifter = 1	1,038	2,029	3,067
Total	4,168	2,414	6,582

Percent of sample years firms are Aggressive Shifters

Unique MNEs	<u>Mean</u>	<u>p25</u>	<u>p50</u>
1,526	0.448	0.000	0.333

Additional Tests: Audits by Year			utShifter		_	gShifter : n = 2,032	
Addits by Teal	<u>Year</u>	<u>N</u>	Audit = 1	Audit Rate	<u>N</u>	Audit = 1	Audit Rate
Even ignoring most recent	2005	36	20	55.56%	67	36	53.73%
years (for which audits may not be initiated), audit rate is	2006	152	91	59.87%	195	135	69.23%
declining.	2007	171	97	56.73%	233	144	61.80%
	2008	166	102	61.45%	219	135	61.64%
In future work, we would like	2009	148	74	50.00%	229	129	56.33%
to compare to average audit rates for all firms.	2010	226	98	43.36%	317	147	46.37%
	2011	286	104	36.36%	336	123	36.61%
	2012	264	63	23.86%	292	76	26.03%
	2013	100	27	27.00%	122	32	26.23%
	2014	19	7	36.84%	22	11	50.00%
Stanford Graduate School of Business	Total	1,568	683	43.56%	2,032	971	47.79%

Additional Tests: Likelihood that Audit = 1

Here and in revision, consistent evidence of no different audit likelihood for net outbound income shifters or firms that shift more out (or less in) than expected.

	<u>Shift</u> Variable =	<u>(1)</u> OutShifter	<u>(2)</u> AggShifter	(<u>3)</u> OutShifter	<u>(4)</u> AggShifter
Intercept	?	-0.0363	-0.1364	-11.5536***	-11.6247***
Shift	+	-0.2228**	0.0419	-0.0873	-0.0608
Size	+			0.5493***	0.5523***
Big5	+			0.0638	0.0668
ROS	+			1.0636***	1.0752***
NOL	-			0.1949	0.1899
Leverage	-			-0.5523***	-0.5398***
Observations		4,322	4,322	4,322	4,322
AUC		0.526	0.505	0.738	0.738

Conclusions

Ongoing work

- Continuing to improve predictive power of the model
- Increased descriptive analysis to understand who the shifters are, how they vary over time

Identification of firms with net outbound intercompany transactions

- Net inbound on average
- Net outbound for High Tech industry and IP transactions
- Increasing in R&D, High Tech industry membership, foreign return on sales, foreign effective tax rate differential
- Decreasing in capex, gross profit, foreign sales, size

Audit likelihoods

 Net outbound shifters and firms that shift more out (or less in) than expected have no different likelihood of audit

Thank you!



Income Shifting by U.S. Multinational Corporations

7th Annual IRS-TPC Joint Research Conference on Tax Administration

Ted Black (IRS RAAS), Amy Dunbar (UConn), Andrew Duxbury (James Madison), and Tom Schultz (Western Michigan)

June 21, 2017

Research Question

What is the question?

- Can we use new foreign tax information reporting (FTIR) data that includes payments of interest and royalties (FDAP income) to U.S. MNCs to estimate *tax compliant* income shifting?
 - We assume that the FDAP "footprint" from the FTIR data provides an empirical proxy measure for compliant income shifting.

• Why important?

- OECD suggests that the FTIR data can be used to improve compliance
- Similar to the effect of a 1099 on reporting behavior

How do we answer the question?

Approaches to estimating income shifting

Prior research

- *magnitude* of income shifting
 - Rousslang (1997) and Christian and Schultz (2005) use a theoretical approach that assumes equal marginal after-tax rate of returns across jurisdictions.
- *tax-motivated* income shifting
 - Hines and Rice (1994) use a production function with tax rates
 - Collins, Kemsley and Lang (1998) and Klassen and LaPlante (2012) use foreign return on sales.

Our approach: estimate compliant income shifting

1. Estimate magnitude of income shifting

2. Match FTIR and IRS data

3. Estimate compliant income shifting using FTIRC measures

Our approach: estimate compliant income shifting

1. Estimate magnitude of income shifting

• Extend prior two-state model to an N-state: CFC /country by country basis

N-State Income Shifting Model

$$\frac{[Y_{\underline{d}} + IS_{\underline{T}}] * (1-t_{\underline{d}})}{K_{\underline{d}}} = \frac{[Y_{\underline{1}} - IS_{\underline{1}}] * (1-t_{\underline{1}})}{K_{\underline{1}}} = \frac{[Y_{\underline{2}} - IS_{\underline{2}}] * (1-t_{\underline{2}})}{K_{\underline{2}}}$$

- Y = pretax income
- K= capital stock
- t = tax rate (T/Y) where T = level of taxes
- $IS_T = total income shift [IS_T = IS_1 + IS_2]$
- 1. Compute ATROC for each MNC and CFC;
- 2. Adjust for the income shift (IS) to equalize the ATROCs;
- 3. Aggregate the IS across all firms within N jurisdictions within year

Our approach: estimate compliant income shifting

1. Estimate magnitude of income shifting

- Extend prior two-state model to an N-state: CFC /country by country basis
- Obtain measures of capital and income from IRS Forms 1120 and 5471

Income Shifting Dataset: MNCs/CFCs

<u>Initial</u> #MNCs		<u>2007</u>	<u>2008</u>	<u>2009</u>	<u>2010</u>	<u>2011</u>	<u>2012</u>	<u>2013</u>	<u>2014</u>	<u>2015</u>
18,009		6,147	6,560	7,009	7,799	8,356	8,777	9,418	10,085	10,626
	#CFCs									
	668,523	67,114	69,897	70,042	71,333	74,653	76,419	77,971	79,283	81,811

Dropped 5471s that report zero income or capital
Using the reduced dataset still resulted in HUGE estimates of income shifting
- small denominator led to ATROCs that exceeded 100%.

Income Shifting Dataset: MNCs/CFCs

<u>Initial</u>		2007	2008	2009	2010	<u>2011</u>	2012	<u>2013</u>	2014	<u>2015</u>
#MNC	S									
18,009	9	6,147	6,560	7,009	7,799	8,356	8,777	9,418	10,085	10,626
	#CFCs									
	668,523	67,114	69,897	70,042	71,333	74,653	76,419	77,971	79,283	81,811
<u>Final</u>										
9,70	1	3,345	3,407	3,571	3,848	4,088	4,187	4,311	4,480	4,547
	214,049	22,304	21,521	21,642	22,890	24,436	25,230	25,533	25,315	25,178

More reasonable ATROCs but magnitude of shifting substantially reduced.

Income Shifting Estimates – Positive Outbound Shifts (ISp) (\$M)

	All (<u>CFCs</u>
•••	N-CFCs	<u>ISp</u>
2007	22,304	\$144,844
2008	21,521	161,884
2009	21,642	140,042
2010	22,890	181,310
2011	24,436	188,616
2012	25,230	165,990
2013	25,533	177,618
2014	25,315	179,714
2015	25,178	161,417

Income Shifting Estimates – Positive Outbound Shifts (ISp)

(\$M)

	All CFCs		1 CFCs FTIR Countries		Non-FTIR	<u>Countries</u>	<u>Unidentified</u> <u>Country</u>		
	N-CFCs	<u>ISp</u>	<u>N</u>	<u>ISp</u>	<u>N</u>	<u>ISp</u>	<u>N</u>	<u>ISp</u>	
2007	22,304	\$144,844	14,813	\$75,783	7,350	\$67,462	141	\$1,599	
2008	21,521	161,884	14,145	79,088	7,277	76,986	99	5,811	
2009	21,642	140,042	14,276	75,221	7,313	63,351	53	1,470	
2010	22,890	181,310	15,121	90,049	7,725	89,200	44	2,062	
2011	24,436	188,616	16,010	91,494	8,374	93,995	52	3,127	
2012	25,230	165,990	16,601	79,829	8,569	82,828	60	3,333	
2013	25,533	177,618	16,678	83,896	8,798	91,745	57	1,978	
2014	25,315	179,714	16,386	81,565	8,877	97,363	52	786	
2015	25,178	161,417	16,437	74,847	8,692	86,245	49	325	
		· · · · · · · · · · · · · · · · · · ·							

Normalized ISp to CFCs Controlled by a Compliance Assurance Process (CAP)-MNC vs NonCAP-MNCs

	CAP					NonCAP				
	N						N			
	MNCs	CFCs	<u>ISp</u>	<u>Yd</u>	NISp	MNC	s CFC	<u>ISp</u>		
2007	88	1,529	\$21,080			11	8 2,923	\$18,981		
2008	70	998	15,095			9	9 2,544	18,455		
2009	68	1,084	12,707			9:	2,278	3 14,954		
2010	84	1,356	19,803			10	3 2,716	24,440		
2011	77	1,248	19,363			10	4 2,884	27,769		
2012	80	1,388	14,583			102	2 3,049	27,145		
2013	88	1,510	18,366			10	1 3,124	24,875		
2014	87	1,468	18,961			9.	5 2,831	26,469		
2015	88	1,395	14,394			8	8 2,630	20,689		

Normalized ISp to CFCs Controlled by a Compliance Assurance Process (CAP)-MNC vs NonCAP-MNCs

	CAP			NonCAP						
	N					N				
	MNCs	<u>CFCs</u>	<u>ISp</u>	<u>Yd</u>	<u>NISp</u>	MNCs	<u>CFCs</u>	<u>ISp</u>	<u>Yd</u>	<u>NISp</u>
2007	88	1,529	\$21,080	118,739	0.151	118	2,923	\$18,981	113,091	0.144
2008	70	998	15,095	80,579	0.158	99	2,544	18,455	94,908	0.163
2009	68	1,084	12,707	78,034	0.140	92	2,278	14,954	78,352	0.160
2010	84	1,356	19,803	105,422	0.158	103	2,716	24,440	98,139	0.199
2011	77	1,248	19,363	74,710	0.206	104	2,884	27,769	96,320	0.224
2012	80	1,388	14,583	85,602	0.146	102	3,049	27,145	104,134	0.207
2013	88	1,510	18,366	118,856	0.134	101	3,124	24,875	111,970	0.182
2014	87	1,468	18,961	124,384	0.132	95	2,831	26,469	131,880	0.167
2015	88	1,395	14,394	127,650	0.101	88	2,630	20,689	128,475	0.139

Our approach: estimate compliant income shifting

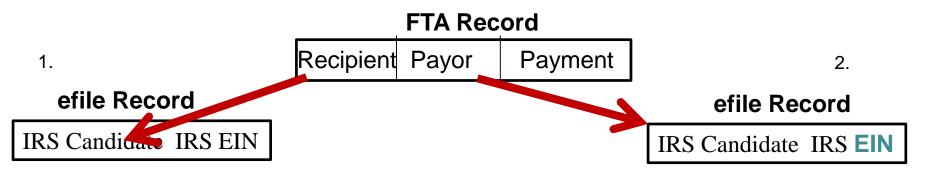
1. Estimate magnitude of income shifting

2. Match FTIR and IRS data

- Match U.S. MNC-related *payors* in the FTIR data to their Form 5471 data.
- Use computational linguistics for matching process because no FTIR TIN.

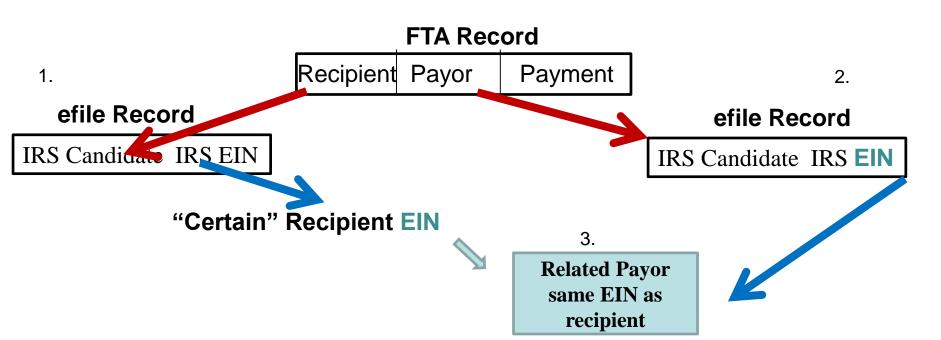
Identifying Related Payors that Report to FTA and IRS

- 1. Match FTA recipient to IRS corporate efile identify "certain" match
- 2. Match FTA payor to IRS corporate efile
- 3. Determine if matched payor EIN matches a "certain" match recipient EIN



Identifying Related Payors that Report to FTA and IRS

- 1. Match FTA recipient to IRS corporate efile identify "certain" match
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Foreign Tax Information Reporting Data: 2007-2012 (\$M)

	Payments N	Percent	
	Corporations	Any Recip	Corp FDAP
FDAP Payments			
Dividends	175,745	284,712	36.0%
Interest	62,598	153,852	19.4%
Royalties	68,489	114,287	14.4%
Capital Gains	1,881	4,127	0.5%
Non-FDAP Payments			
Other income	<u>77,609</u>	234,547	
	\$386,322	\$791,525	

Recipients Matching: Certain/Uncertain

N (unique)	<u>Certain</u> 63,921	<u>Uncertain</u> 159,311	<u>Total</u> 223,232
\$ M Total % of	\$265,649	\$117,322	\$382,971 \$386,322
recipients	28.6%	71.4%	
% of income	69.4%	30.6%	

The recipient matching process found multiple potential matches for 223,232 unique recipients, but there were recipients that could not be matched at all. For example, the name on the data record was simply random characters. These certain and uncertain matches account for \$382,971/\$386,322 (99%) of the payments to U.S. corporate recipients.

Recipients Matching: Certain/Uncertain Payors: Related/Unrelated

N (unique)	<u>Certain</u> 63,921	<u>Uncertain</u> 159,311	<u>Total</u> 223,232
\$ M Total	\$265,649	\$117,322	\$382,971 \$386,322
% of recipients % of income	28.6% 69.4%	71.4% 30.6%	· ,

Certain and uncertain matches account for 99% of the payments to U.S. corporate recipients.

	Related	<u>Unrelated</u>	Total
N (unique)	7,911	39,165	47,076
Dividends	\$99,910	\$75,835	\$175,745
Interest	36,388	26,210	62,598
Royalties	42,011	26,478	68,489
Capital Gains	182	1,699	1,881
Other	33,510	44,100	77,610
Total	\$212,001	\$174,321	\$386,322
% of payors	16.8%	83.2%	
% of payments	54.9%	45.1%	

The related payors make payments to US beneficial owners that average \$27M (\$212,000/7,911) over the 2007-2012. The unrelated payors remit an average \$4.5M to US beneficial owners.

Recipients Matching: Certain/Uncertain Payors: Related/Unrelated

N (unique)	<u>Certain</u> 63,921	<u>Uncertain</u> 159,311	<u>Total</u> 223,232
\$ M Total	\$265,649	\$117,322	\$382,971 \$386,322
% of recipients	28.6%	71.4%	
% of income	69.4%	30.6%	

Certain and uncertain matches account for 99% of the payments to U.S. corporate recipients.

N (unique)	Related 7,911	Unrelated 39,165	<u>Total</u> 47,076
Dividends Interest Royalties Capital Gains Other	\$99,910 36,388 42,011 182 33,510	\$75,835 26,210 26,478 1,699 44,100	62,598 68,489 1,881
Total % of payors % of payments	\$212,001 16.8% 54.9%	\$174,321 83.2% 45.1%	\$386,322

The related payors make payments to US beneficial owners that average \$27M (\$212,000/7,911) over the 2007-2012. The unrelated payors remit an average \$4.5M to US beneficial owners.

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2. Match FTIR and IRS data

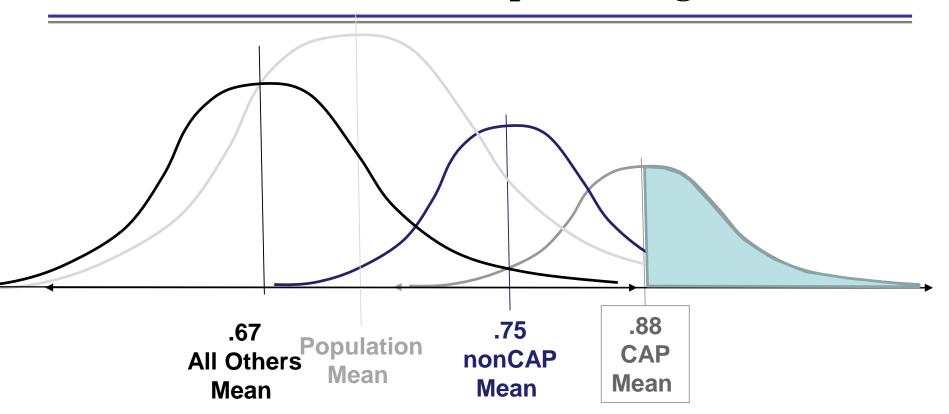
- 3. Estimate compliant income shifting using FTIRC measures
 - Compute two compliance measures

Foreign Tax Information Reporting Compliance (FTIRC) Measures

FTIRC1 FTIRC2

- the ratio of the matched recipients to all recipients for each related payor, weighted by the FDAP payments.
 - The percent of all payments reported to the FTA for recipients matched to the corporate efile.

FTIRC1 – Recipient Filing



Foreign Tax Information Reporting Compliance (FTIRC) Measures

FTIRC1

- the ratio of the matched recipients to all recipients for each related payor, weighted by the FDAP payments.
 - The percent of all payments reported to the FTA for recipients matched to the corporate efile.

FTIRC2

- the ratio of expense reported by the CFC to the FTA, relative to the expense reported by the CFC to the IRS.
 - FTA royalties/IRS royalties
 - FTA interest/IRS interest

FTIRC2 Compliance Adjustment Factor (CAF)

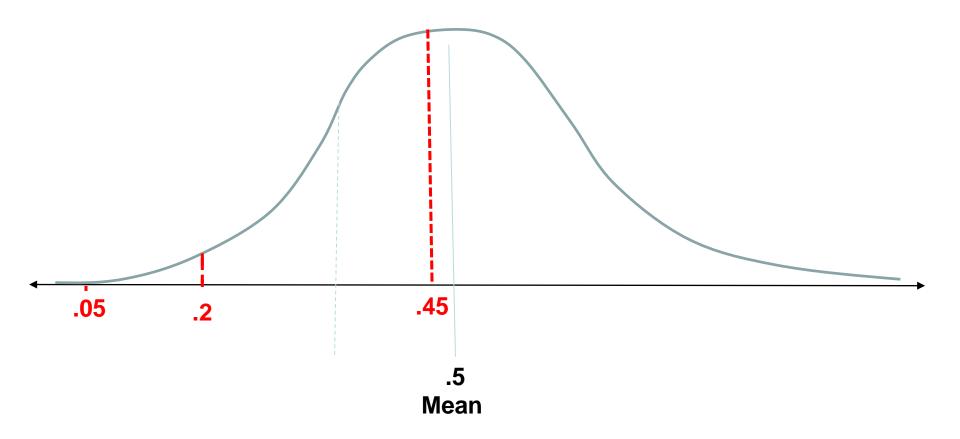
If a CFC's

■ FTIRC2 < mean FTIRC2 for the CFC's group,

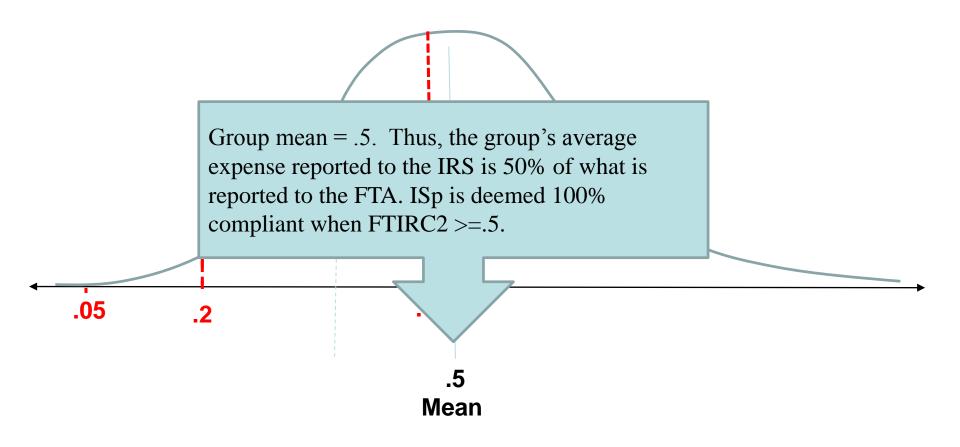
Then we compute

- CAF = FTIRC2/mean FTIRC2.
- Then compute a weighted average of the royalty and interest CAFs, where the weights are the relative share of (interest or royalty) expense

FTIRC2 – Compliance Adjustment Factor



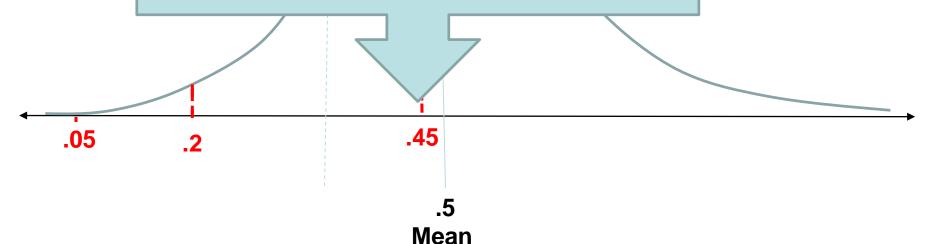
FTIRC2 – CAF



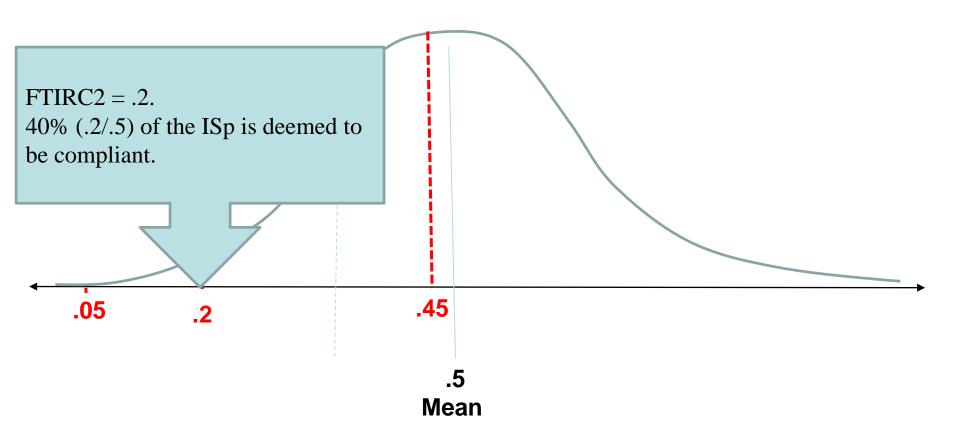
FTIRC2 – CAF



The CFC's royalty expense reported to the IRS is 45% of what is reported in the source country. Because this CFC's FTIRC2 is below the mean, 90% (.45/.5) of the ISp is deemed to be compliant.



FTIRC2 – CAF



Compliant Income Shifting Estimates

							Compliance Adjustment Factor			
		ISp		C	ompliant l	ISp	(Percent Compliant)			
_	CAP	nonCAP	others	CAP	nonCAP	others	CAP	nonCAP	others	
2007	9,518	13,958	51,039	4,135	6,150	22,523	0.434	0.441	0.441	
2008	6,559	12,771	56,683	3,058	5,970	25,077	0.466	0.467	0.442	
2009	12,928	14,705	48,533	6,177	6,663	21,940	0.478	0.453	0.452	
2010	10,633	20,167	58,241	5,035	9,435	26,362	0.474	0.468	0.453	
2011	12,212	18,254	61,226	5,591	8,170	26,965	0.458	0.448	0.440	
2012	10,049	16,185	53,660	4,800	7,280	23,699	0.478	0.450	0.442	

Conclusion

- FTIRC1 showed expected differences between CAP, nonCAP and all other filers.
- FTIRC2 did not support the conclusion that CAP taxpayers are more compliant.
 - Further study is needed to confirm the attributes used to create the groups.

Income Shifting by U.S. Multinational Corporations by Black, Dunbar, Duxbury, and Schultz

and

Using IRS Data to Identify Income Shifting Firms by Simone, Mills, and Stomberg

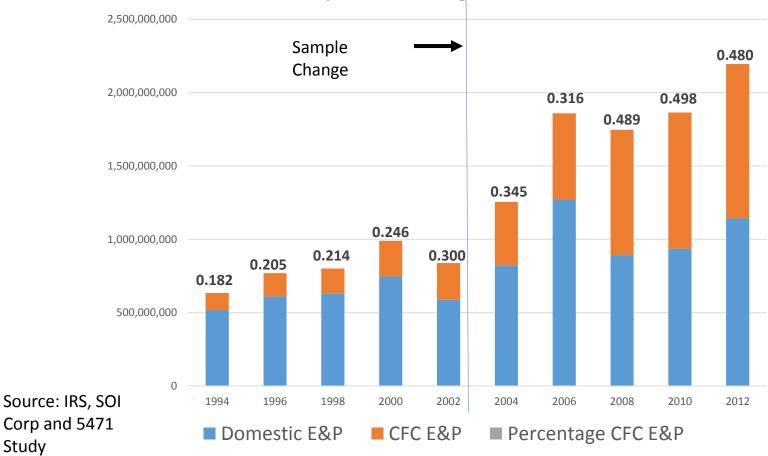
Discussion by

Tim Dowd, Joint Committee on Taxation

Profit Shifting

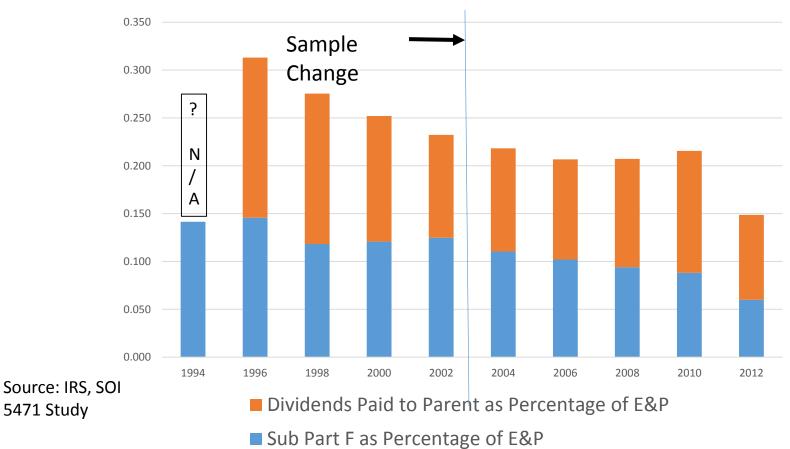
- Both papers look at the profit shifting of US corporations.
 - Black, et al. objective is to quantify the amount of US base erosion from Shifting and understand foreign CFC compliance.
 - Simone, et al. objective is to use third party publicly available data to identify profit shifting corporations.
- These are important questions for policy makers, investors, and the IRS.
- Why? Because there is a lot of money out there, and it is in low tax jurisdictions.

Corporate Earnings and Profits



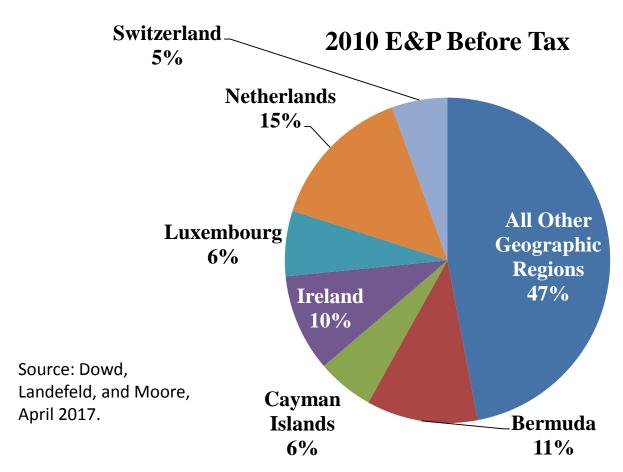
Study

Distributions to Parent as a Percent of CFC E&P



5471 Study

Figure 7a: 2010 Foreign Earnings and Profits before tax for US multi-national corporations



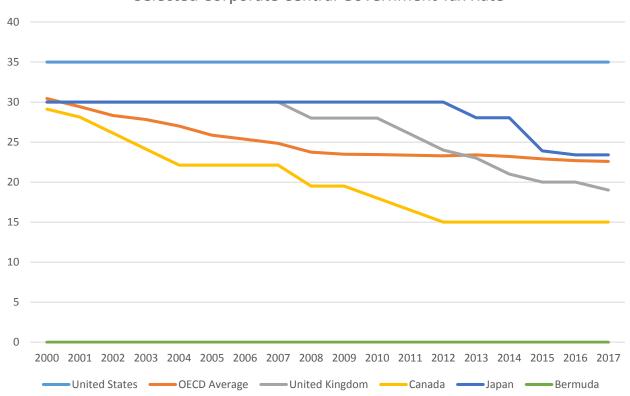
Economic and Population Growth

 Increased foreign E&P could be driven by more rapid foreign population and economic growth:

	U.S.	ROW
Pop Growth 1994-2015	20%	27%
GDP Growth 1994-2015	50%	67%

Corporate Tax Rates

Selected Corporate Central Government Tax Rate



Income Shifting by US Multinationals

- Premise: Variation in after-tax rates of return on investment are driven by profit shifting. In the absence of profit shifting
 - ATROI_i = ATROI_i = ATROI_{worldwide} = ATROI_{US}
- can calculate the amount of shifted income in order to equalize the observed after tax rates of return.
- Authors sum up all positive shifted income in each CFC-country.
 - Find that over the period 2007-2015, there was positive average income shifted of \$166 billion a year.

Income Shifting by US Multinationals ISM Information Reporting and CAP

- Find that much more shifting occurring in non-participant countries.
- Also much more shifting occurring with non-cap corporations.
- But, there are inference problems here:
 - Information reporting country results:
 - Are these generally higher tax countries?
 - Is this where there are negative shifting out occurring?
 - Etc..
 - CAP Program results:
 - Rollout of new compliance program might → choose the easy to understand returns.
 - Invitation only → sample selection problems.
 - Application only later with requirement that the corporation "display a general willingness to be transparent and cooperative the IRS."

Income Shifting by US Multinationals ISM Suggestions

- The authors make two related assumptions:
 - 1. all positive income shifted to CFCs is shifted out of the U.S.
 - 2. ignore negative income shifted out of CFCs.
- These assumptions are wrong because they assume shifting out of CFCs into other CFCs is from the US.
- Corporations engage in considerable foreign to foreign income shifting.

Income Shifting by US Multinationals ISM Suggestions

	US Parent	German CFC	Cayman CFC	Sum of IS
True Income Shifting	-100	-50	+150	0
Assume Positive CFC Only	-150	0	+150	0

Using IRS Data to Identify Income Shifting Firms

- Premise:
 - Net Outbound Payments to Related Parties Signify Shifting Activity.
 - Authors find that 37% of their sample are shifters.
- Run logit model to predict who is a shifter using publicly available data for the independent variables.
- Test model on out of sample data.
- Test aggressiveness of shifters by looking at the intensity of shifting.
- Test IRS Audit propensity against both the likelihood model and the intensity model

Using IRS Data to Identify Income Shifting Firms Results

- Firms that engage in RD and High Tech are likely to be net outbound shifters.
- Firms that have large capital expenditures are less likely to be shifters.
- Havens increase likelihood.
- The lagged tax gap between the US and foreign increases the likelihood.
- Find that "Aggressive income shifters also have greater R&D expense, capital expenditures, gross profit percentages, foreign sales intensities, domestic sales growth and worldwide sales (Size) at the mean and median relative to other MNEs."
- And that net related party shifting is not correlated with the probability of an audit.

Using IRS Data to Identify Income Shifting Firms More Data/Heterogeneity of Corporations

- Net Shifters shifted on average 7% of gross sales.
 - What does this look like at the 75th percentile? 90th percentile?
- Net related party payments can be decomposed.
 - Net payments for stock in trade.
 - Net payments for tangible property.
 - Net payments for Intangible income.
 - Net dividends.
 - Net interest.

Using IRS Data to Identify Income Shifting Firms Likelihood Model

- Authors use lagged values of foreign taxes to avoid endogeneity.
 - Do we care about endogeneity here?

- Authors use a logit model and try an out of sample prediction.
 - Maybe a linear probability model will give better predictions.

Using IRS Data to Identify Income Shifting Firms Shifting Intensity and Audits

- Zero effect of shifters on audit probability. Was IRS getting it wrong?
 - Not clear.
 - Audit regressions with change in liability indicator as the dependent variable would answer whether the net outbound related party payment measure is the right measure of income shifting.
 - In other words, is the premise of the paper correct?
- Effect of size on shifting (-) and audits (+).
 - \rightarrow size shouldn't be used as a determinant of audits.
 - However, historically it has been a major determinant. So, maybe it has been effective. Can this be teased out with firms crossing the audit size threshold?



The economic effects of special purpose entities on corporate tax avoidance

Paul Demeré
Univ. of Georgia

Michael Donohoe

Univ. of Illinois

Petro Lisowsky
Univ. of Illinois, NoCeT



ILLINOIS

Overview

Goal: Examine how SPEs facilitate corporate tax avoidance

Special Purpose Entities (SPEs)

 Separate legal entities created by a sponsor-firm to perform narrow, predefined business activities or series of transactions (Feng Gramlich Gupta 2009)

Research questions

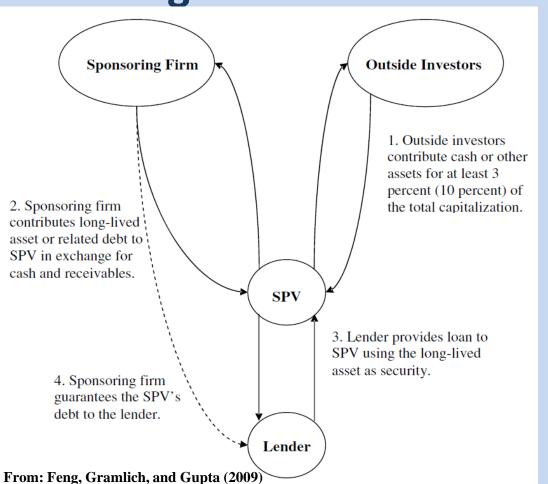
- To what extent are transactions used within SPEs for tax avoidance?
- How large are the total corporate tax savings facilitated by SPEs?
- For which transactions do SPEs enhance relative tax savings?

Overview

Motivation

- Many papers and reports study different tax advantaged transactions
- Few studies on the organizational structures facilitating transactions
 - Mostly on multinational firms seeking to shift income
- Most evidence on tax revenue losses is anecdotal
 - But, SPEs have many non-tax applications
- Corporate use of SPEs is large (50% of S&P 500) and growing (600%)
- Prior research focuses on determinants of SPEs, but not effects.
- What are the corporate tax effects of these structures?

Background on SPEs



Tax Advantages of SPEs

Facilitate tax avoidance

- Enable sponsors to conduct a greater level of tax-advantaged transactions
- Enhance the tax efficiency (i.e., relative tax savings) of such transactions, holding level constant

 $Total\ Tax\ Savings = f(Level, Efficiency)$

Disadvantages of SPEs

Non-tax and tax costs of SPEs are potentially large

Reduce information qualityFeng et al. (2009)

Increase regulatory scrutiny
IRB 2011-39; Inland Revenue (2013)

Enhance public pressureDyreng et al. (2016)

Result in large tax penalties Wilson (2009)

Result in higher taxesWittendorff (2010)

- Key drivers of SPE use (Feng et al. 2009)
 - Financial reporting pressures, governance, and others
 - Tax avoidance not necessarily a major objective for SPEs
- Is tax avoidance via SPEs economically significant?

Research Design

- Measures of SPE use
 - Identification using Feng et al. (2009) approach
 - Python script: LLP, LLC, LP, and other pass-thru subs in Exhibit 21
 - Mitigates selection bias (mandatory disclosure)
 - SPETOT = log of (one plus) the total number of SPEs
 - Winsorize at top 1% to mitigate outliers
 - SPEBIN = indicator for firm-years with an SPE; 0 otherwise
- Measures of tax avoidance
 - Forward-looking ETRs estimated over three years (t to t+2)
 - GETR (GAAP ETR) = total tax expense / pre-tax book income
 - CETR (Cash ETR) = worldwide cash taxes paid / pre-tax book income

Research Design

Empirical model

- $ETR = \beta_{i0} + \beta_1 SPE_{it} + \sum_{j=2}^{11} \beta_j TAT_{jit} + \sum_{j=12}^{20} \beta_j CTRL_{jit} + \delta_{0t} + \epsilon_{it}$
 - TAT vector of variables capturing <u>Tax-Advantaged Transactions</u>
 - CTRL vector of control variables (for ETR regressions)
 - Also include other structures (haven, business segments)
 - Firm and year fixed-effects → generalized difference-in-differences
 - Adapt model to examine our research questions: Path and Moderation

Sample selection

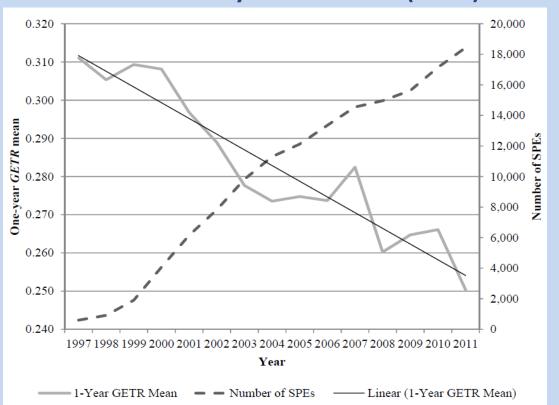
- Compustat [1997-2011]
 - Publicly traded; domestic; positive total assets
 - Drop negative three-year pre-tax income; regulated/financial firms
 - Require two future years of data for future ETRs
 - 25,533 observations from 4,566 unique firms

Descriptive statistics

Temporal distribution

	(1)	(2)	(2)	(4)	(5)	(6)
	— (1) Total	(2) SPE	(3) SPEs Per User	(4) Non-	(5) Total Obs.	(6) SPE Use
Maan						
Year	SPEs	Users	[(1)/(2)]	Users	[(2)+(4)]	[(2)/(5)]
1997	605	165	3.67	1,589	1,754	9.4%
1998	917	207	4.43	1,517	1,724	12.0%
1999	1,910	268	7.13	1,372	1,640	16.3%
2000	4,075	520	7.84	1,372	1,556	33.4%
2001	6,161	607	00/1078 12.25	1,010	1,617	37.5%
2002	7,827	726	0/1078	1,071	1,797	40.4%
2003	9,823	86	12.25	1,054	1,856	43.2%
2004	11,302	839	13.47	970	1,809	46.4%
2005	12,125	875	13.86	909	1,784	49.0%
2006	13,353	876	15.24	828	1,704	51.4%
2007	14,552	917	15.87	805	1,722	53.3%
2008	14,961	910	16.44	785	1,695	53.7%
2009	15,634	921	16.98	708	1,629	56.5%
2010	17,147	971	17.66	666	1,637	59.3%
2011	18,436	999	18.45	610	1,609	62.1%
Total	148,828	10,603	14.04	14,930	25,533	41.5%

- Descriptive statistics
 - Time trends in SPEs and one-year GAAP ETR (GETR)



Descriptive statistics

Industry distribution

·	(1)		(2))	(3))	(-	4)
	Total SPEs		SPE U	SPE Users		Non-Users		[(2+3)]
NAICS Industry	Obs.	%	Obs.	%	Obs.	%	Obs.	SPE Use
62: Health Care More	24,872	16.7	381	3.6	256	1.7	637	59.8%
71: Arts & Entertainment	2,274	1.5	112	1.1	78	0.5	190	58.9%
81: Other Services intangibles	1,173	0.8	83	0.8	62	0.4	145	57.2%
23: Construction	5,447	3.7	228	2.2	172	1.2	400	57.0%
56: Administrative & Support Service egal risk	5,351	3.6	366	3.5	335	2.2	701	52.2%
44: Consumer Retail	11,757	7.9	468	4.4	526	3.5	994	47.1%
72: Accommodation & Food Services	5,793	3.9	319	3.0	367	2.5	686	46.5%
54: Professional Services	4,605	3.1	569	5.4	660	4.4	1,229	46.3%
53: Real Estate	6,568	4.4	298	2.8	348	2.3	646	46.1%
45: Miscellaneous Retail	1,604	1.1	249	2.3	308	2.1	557	44.7%
21: Mining, Oil, & Gas Extraction	8,388	5.6	656	6.2	818	5.5	1,474	44.5%
32: Wood & Petroleum Products Manufacturing	10,942	7.4	1,451	13.7	1,888	12.6	3,339	43.5%
31: Food & Apparel Manufacturing	7,965	5.4	695	6.6	968	6.5	1,663	41.8%
51: Information	19,099	12.8	994	9.4	1,398	9.4	2,392	41.6%
49: Couriers & Warehousing	99	0.1	38	0.4	55	0.4	93	40.9%
42: Wholesale Trade	3,226	2.2	493	4.6	717	4.8	1,210	40.7%
61: Education	490	0.3	44	0.4	78	0.5	122	36.1%
48: Transportation	9,567	6.4	352	3.3	626	4.2	978	36.0%
33: Other Manufacturing Intangibles	19,225	12.9	2,742	25.9	5,121	34.3	7,863	34.9%
99: Other	246	0.2	35	0.3	75	0.5	110	31.8%
11: Agriculture, Forestry, & Fishing legal risk	137	0.1	30	0.3	74	0.5	104	28.8%
_Total	148,828	100.0	10,603	100.0	14,930	100.0	25,533	41.5%

Relation between SPEs and corporate tax avoidance

	(1)		(2)		(3)		(4)		
	GETR		GETR		CET	R	CETR		
	Coeff.	t-stat	Coeff.	t-stat	Coeff.	t-stat	Coeff.	t-stat	
Measures o	of SPE use (S	PE)							
SPETOT	-0.010 **	* -3.95			-0.008 **	* -2.68			
SPEBIN			-0.014 ***	-2.75			-0.012 **	-2.44	

- First large-sample evidence on the overall relation between SPEs and ETRs
 - SPEs facilitate tax avoidance above and beyond common tax-advantaged transactions (TAT) and controls (CTRL)
- Results serve as an important starting point
 - Overall Effects = Direct Effects + Indirect Effects
 - Path Analysis (RQ1 and RQ2)
 - Moderation Analysis (RQ3)

Path analysis diagram

Empirical mapp	ing —		
Tax-advantaged transaction	Vector TAT		
Increase leverage	LEV, DISS, CLTD		
Hold R&D activities	RDE		Corporate ta
Hold depreciable assets	PPE, INTANG	Direct effects	avoidance
Foreign transfer pricing	INTANG, FINC		ETR
Domestic transfer pricing	INTANG		EIK
Tax credit and loss duplication	NOL		
Basis shift to tax preferred assets	PPE, $M&A$		
Foreign tax deferral	FINC, HAVEN		
Unmeasured	tax-advantaged transactions	Special Purpose Entities	
		SPE	

- With SPE in the model, coefficients for TAT capture direct effect of measured transactions on ETRs, absent the use of SPEs (solid arrows)
- Path analysis steps
 - Map each tax-advantaged transaction to at least one TAT variable
 - Estimate model with and without SPE to obtain path coefficients

Level of tax-advantaged transactions used within SPEs (RQ1)

Panel A: GETR										Pan	el B:CETR			
	Tot	al	Dire	ect	Indirect (within)		Total		Direct		Indirect (within		in)	
	Coeff.	t-stat	Coeff.	t-stat	Coeff.	t-stat	%	Coeff.	t-stat	Coeff.	t-stat	Coeff.	t-stat	%
SPETOT			-0.037	-3.95						-0.027	-2.68			
Tax-advantaged	l transactio	ons (TAT)												
LEV	-0.030	-2.55	-0.029	-2.46	-0.001	-1.87	3.6	-0.053	-4.19	-0.052	-4.12	-0.001	-1.79	1.8
DISS	0.036	3.45	0.035	3.37	0.001	2.04	2.5	0.028	2.58	0.027	2.51	0.001	1.87	2.6
CLTD	-0.026	-2.60	-0.026	-2.58	0.000	0.89	1.2	-0.016	-1.52	-0.016	-1.50	0.000	0.65	1.1
RDE	-0.016	-1.59	-0.015	-1.49	-0.001	-2.61	6.6	-0.010	-0.97	-0.009	-0.89	-0.001	-2.12	8.7
PPE	-0.027	-1.85	-0.027	-1.88	0.000	0.70	1.4	0.028	1.94	0.028	1.92	0.000	0.79	1.3
INTANG	-0.052	-3.64	-0.050	-3.51	-0.002	-2.39	3.5	0.019	1.31	0.020	1.39	-0.001	-1.74	6.1
NOL	-0.066	-5.24	-0.065	-5.15	-0.001	-2.83	2.0	-0.025	-2.23	-0.024	-2.16	-0.001	-2.03	3.3
M& A	0.010	1.17	0.010	1.19	0.000	0.40	1.4	0.031	3.46	0.031	3.46	0.000	0.26	0.2
FINC	-0.029	-3.01	-0.028	-2.96	-0.001	-1.21	1.7	-0.034	-3.62	-0.034	-3.59	0.000	0.76	0.7
HAVEN	-0.017	-1.86	-0.011	-1.21	-0.006	-3.72	34.6	-0.001	-0.13	0.003	0.31	-0.004	-2.59	331.2

- Negative Indirect → SPEs result in more tax avoidance for given variable
 - **Example:** A one std. dev. increase in *LEV* results in a 0.030 std. dev. decrease in *GETR*, where 0.001 occurs from leverage *within* SPEs and 0.029 occurs from leverage *outside* of SPEs
 - Indirect% → 3.6% of total tax savings from LEV occurs within SPEs
- SPEs facilitate a greater level of specific transactions such that an economically large portion of the total cash tax savings occurs within SPEs
 - Lev (1.8%); NOL (3.3%); R&D (8.7%); intangibles (6.1%); haven (all)

- Total tax savings facilitated by SPEs (RQ2)
 - SPE users: GETR and CETR are 1.6 and 1.2% points lower than non-users
 - Firm-level: GAAP and cash tax savings of \$9.84M and \$7.77M per year
 - Sample-level: cash tax savings alone averages \$82B (as high as \$165B)
 - 1.9% (up to 3.7%) of total U.S. corporate tax revenues collected
- Comparisons:
 - Havens (\$1.3B); round-tripping (\$33.0B); shelters (\$12.4B); derivs (\$3.8B)

	Pa	anel A: GETR		Panel B: CETR				
	(1)	(2)	(2)/(1)	(1)	(2)	(2)/(1)		
	Main Effort	Interaction	Change	Main Effect	Interaction	Change		
	(t-stat)	(t-stat)	(%)	(t-stat)	(t-stat)	(%)		
SPETOT	-0.038***			-0.031***				
	(-3.30)			(-2.59)				
Tax-advantaged	папзаснов (1.	4 <i>T</i>)						
LEV	-0.034***	-0.002	0%	-0.046***	0.022**	47.8% ↓		
	(-3.17)	(-0.21)		(-4.09)	(2.19)			
DISS	0.032***	0.011	0%	0.020**	-0.001	0%		
	(3.44)	(1.59)		(2.10)	(-0.16)			
CLTD	-0.026***	0.014*	53.8% ↓	-0.015	0.007	0%		
	(204)	(1.72)		(-1.55)	(0.91)			
RDE	-0.027**	-0.025**	92.6% ↑	-0.009	-0.010	0%		
	(-2.57)	(-2.30)		(-0.96)	(-0.96)			
PPE	-0.029**	-0.001	0%	0.026**	0.007	0%		
	(-2.19)	(-0.06)		(2.05)	(0.65)			
INTANG	-0.051***	-0.037***	72.5% ↑	0.026*	-0.013	0%		
	(-3.61)	(-3.30)		(1.89)	(-1.07)			
NOL	-0.059***	0.028**	47.5% ↓	-0.028**	-0.001	0%		
	(-4.76)	(2.23)		(-2.50)	(-0.06)			
M& A	0.009	0.005	0%	0.030***	-0.016**	53.3% ↑		
	(1.21)	(0.60)		(3.54)	(-2.09)			
FINC	-0.031***	0.008	0%	-0.037***	0.012	0%		
	(-3.41)	(0.98)		(-4.17)	(1.43)			
HAVEN	-0.008	0.003	0%	0.004	0.006	0%		
	(-0.88)	(0.32)		(0.40)	(0.69)			
Intercept	-0.001*			-0.001				
	(-1.71)			(-0.96)				

- Debt (NOLs) within SPEs is 47.8%-53.8% (47.5%) *less* tax efficient
- R&D and intangibles-based trans. 92.6% and 72.5% more tax efficient

Example: R&D

Main

One s.d. increase in *SPETOT* results in a 0.038 s.d. decrease in *GETR* (at the mean of all variables).

One s.d. increase in *RDE* results in a 0.027 s.d. decrease in *GETR*.

Interaction

For one s.d. increase in *SPETOT*, effect of one s.d. increase in *RDE* is assoc. with further 0.025 s.d. decrease in *GETR* (for total of 0.052).

|Change| (%)

Incremental reduction in *GETR* due to increase in *SPETOT* (-0.025 / -0.027)

Other Tests

- Tax aggressiveness
 - SPEs facilitate some, but not overly, aggressive tax positions
- SPEs in domestic vs. foreign jurisdictions
 - GETR (but not CETR) results stronger for U.S. MNCs
 - Majority of tax savings from avoiding U.S. federal income taxes
- SPEs by industry
 - Despite high-tech/intangibles anecdotes, results pervasive across industries
- Endogenous choice to use SPEs
 - Heckman two-stage model, PSM, and entropy balancing
- Robust to:
 - Analyses relating to minority interest
 - Changes in Exhibit 21 disclosures ("disappearing subs") and disclosure regimes

Bottom Line

Results

- SPEs facilitate a greater level of specific transactions such that an economically large portion of the total cash tax savings occurs within SPEs
 - Lev (1.8%); NOL (3.3%); R&D (8.7%); intang (6.1%); haven (all)
- SPEs facilitate an economically large amount of total tax savings
 - \$82.4 billion (sample); 2% of U.S. corporate tax revenue
- SPEs enhance the tax efficiency of some transactions
 - R&D (92.6%) and intangibles (72.5%)

Contributions

- Differ from traditional tax avoidance research (the "what")
 - Organizational structures (the "how")
- First large-sample empirical estimates of tax savings facilitated by SPEs
- First to use path and moderation analysis to separate level from efficiency





Session 1

Identifying Corporation Tax Avoidance

Discussion of

 The Economic Effects of Special Purpose Entities on Corporate Tax Avoidance (Pete Lisowsky)

Eric Toder
(Tax Policy Center)

IRS-TPC Research Conference
June 21, 2017





Session 1. Identifying Corporation Tax Avoidance

Moderator: Eric Toder

Tax Policy Center

Using IRS Data to Identify Income Shifting Lisa De Simone

Firms Stanford University

Income Shifting by U.S. Multinational

Amy Dunbar

University of Connecticut

Corporations University of Connecticut

The Economic Effects of Special Purpose Petro Lisowsky

University of Illinois at Urbana Champa

Entities on Corporate Tax Avoidance
University of Illinois at Urbana-Champaign

Discussants:Tim Dowd Joint Committee on Taxation staff

Eric Toder Tax Policy Center