

# Technology and Real Options: Evidence from Patent Text

Davidson Heath, USC Marshall



# Motivation

- ▶ Technology is the central fact of long-run growth at the aggregate and the firm level
- ▶ Hard to measure, often as a residual (TFP)

# This Paper

- ▶ Measure firms' technology using patent text
  - ▶ Open up the black box, exploit vast textual data

Research Question:

- ▶ How do firms' technology positions affect risk and returns?

# Patent Text

- ▶ Body text of every patent granted by USPTO from 1926-2010
  - ▶ online at the Google Patents Project
- ▶ 6.2 million patents, 240GB text (~134,000,000 pages)
- ▶ Select one, two and three word phrases (“ngrams”)
  - ▶ English nouns
  - ▶ Standard text screens

# Ngrams

Year first appearing:		
1950	1970	1990
control circuitry	acid sequence	email address
clock cycle	interferon	notebook computer
substrate material	nucleotide	remote memory storage
epoxy resins	bus interface	jpeg image
computer networks	programming language	terabyte
command signal	interface card	picture experts
remote computer	plasma display	picture experts group
enantiomers	cholesterol	email addresses
breast cancer	mosfet	sound card
polymer matrix	cpu controls	multiple servers

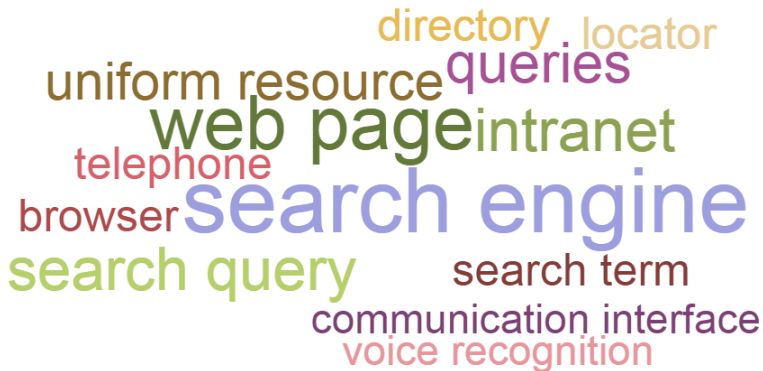
# Patent Data

- ▶ Each patent:
  - ▶ Evaluate presence (=1) or absence (=0) of each ngram
  - ▶ Boolean vector with 64,000 entries
  - ▶ Normalize to unit length
- ▶ Maps each patent to a point in 'word space'
- ▶ Text-based patent measures predict patent renewals

## Firm-Years

- ▶ 18-58% (mean=29%) of Compustat firms in each year 1961-2008 have at least one patent to their name
- ▶ Sum the vectors of all patents for each firm-year
  - ▶ Normalize to unit length
  - ▶ Maps firm-years in 'patent word space'

## GOOG in 2000



A word cloud of terms associated with Google in 2000. The words are arranged in a cluster, with 'search engine' being the largest and most central. Other prominent words include 'web page', 'queries', 'directory', 'locator', 'uniform resource', 'intranet', 'telephone', 'browser', 'search query', 'search term', 'communication interface', and 'voice recognition'. The words are in various colors including blue, green, purple, orange, and red.

search engine  
web page  
queries  
directory  
locator  
uniform resource  
intranet  
telephone  
browser  
search query  
search term  
communication interface  
voice recognition



## INTL in 2000

A word cloud of computer architecture terms from the year 2000. The words are arranged in a cluster, with 'clock cycle' being the largest and most central. Other prominent words include 'memory controller', 'semiconductor', 'latency', 'system memory', 'pipeline', 'memory location', 'access memory dram', 'memory address', 'execution unit', 'control logic', 'oxide', 'state machine', 'bus interface', 'memory dram', and 'clock'. The words are color-coded: green for 'clock cycle', 'system memory', 'bus interface', and 'access memory dram'; red for 'memory location', 'oxide', 'state machine', and 'memory dram'; blue for 'execution unit'; orange for 'pipeline'; and purple for 'control logic'.

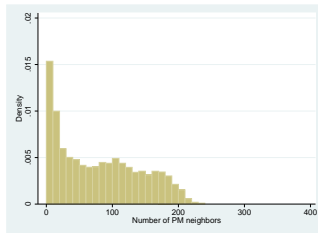
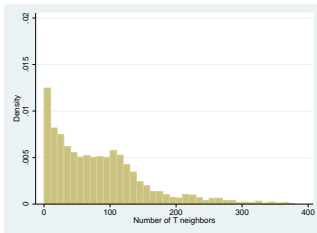
access memory dram  
memory location  
memory address pipeline  
execution unit control logic  
memory controller  
system memory  
clock cycle clock  
oxide semiconductor  
memory dram latency  
state machine  
bus interface

## KO in 2000



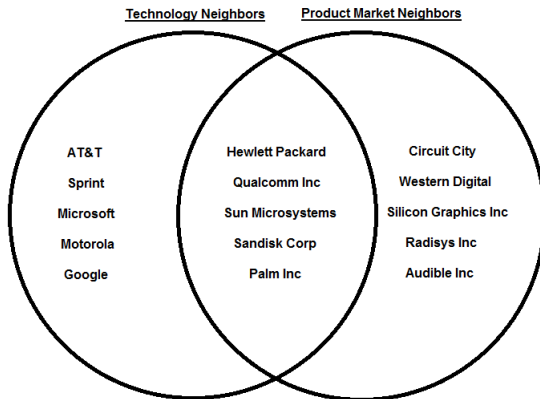
# Complementarity with TNIC

- ▶ Hoberg and Phillips TNIC uses 10-K product descriptions
  - ▶ “PROD” neighbors
- ▶ I assign firms to be neighbors if similarity > cutoff
  - ▶ “TECH” neighbors
- ▶ Both networks are ~3.1% saturated



# Complementarity with TNIC

- ▶  $Pr(TECH_t|PROD_t) = 32.7\%$ ,  $Pr(PROD_t|TECH_t) = 31.8\%$
- ▶ Apple in 2006: (pre-iPhone):



## Complementarity with TNIC

- Dummy for whether firms are PROD and TECH neighbors:

	(1) <i>PRODNEIGHBOR</i> <sub>t+1</sub>	(2) <i>TECHNEIGHBOR</i> <sub>t+1</sub>
<i>TECHNEIGHBOR</i> <sub>t</sub>	0.024*** (0.0075)	0.66*** (0.012)
<i>PRODNEIGHBOR</i> <sub>t</sub>	0.32*** (0.021)	-0.025*** (0.0018)
Observations	3,461,540	3,461,540
Fixed Effects	Year	Year
Pseudo R-squared	0.14	0.38

Robust standard errors in parentheses

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1

## Main Measure: *techdiff*

- ▶ For each firm-year, compute its cosine similarity with all other firms in the same year

- ▶ Define

$$techdiff_{it} = 1 - quantile_j^{99}(cos_{ijt})$$

- ▶ Of 1000 firms in that year, how close is the 10th closest firm

# Associated with $Q$

	(1) $\log(Q_t)$	(2) $\log(Q_t)$
$techdiff_t$	0.078*** (0.018)	0.17*** (0.040)
$\log(cites_t)$	0.014** (0.0059)	0.017 (0.011)
$HHI_{SIC,t}$	0.011 (0.0093)	-0.0064 (0.020)
$tnic3tsimm_t$		0.090*** (0.034)
$RDstock / AT_t$	-0.0072 (0.0068)	-0.059** (0.030)
$\log(AT_t)$	-0.23*** (0.023)	-0.35*** (0.049)
Observations	38,762	12,946
R-squared	0.041	0.078
No of firms	4,038	2,262
Fixed Effects	Firm + Year	Firm + Year

Robust standard errors in parentheses

\*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$

# Associated with *ROA*, *CAPEX*

	(1) <i>ROA<sub>t</sub></i>	(2) <i>ROA<sub>t</sub></i>	(3) <i>CAPX/AT<sub>t</sub></i>	(4) <i>CAPX/AT<sub>t</sub></i>
<i>techdiff<sub>t</sub></i>	0.030*** (0.0042)	0.038*** (0.0079)	0.0042*** (0.00094)	0.0053*** (0.0015)
<i>log(cites<sub>t</sub>)</i>	-0.00079 (0.0016)	-0.0010 (0.0031)	0.00027 (0.00044)	-0.00044 (0.00067)
<i>HHI<sub>SIC,t</sub></i>	-0.0032 (0.0021)	-0.011* (0.0055)	0.00094 (0.00089)	0.000045 (0.0014)
<i>tnic3tsimm<sub>t</sub></i>		-0.0072 (0.0058)		-0.0023* (0.0014)
<i>RD/sales<sub>t</sub></i>	-0.061*** (0.0052)	-0.073*** (0.010)	0.00035 (0.00053)	-0.00034 (0.00077)
<i>log(AT<sub>t</sub>)</i>	0.095*** (0.012)	0.23*** (0.021)	0.00065 (0.0018)	-0.0011 (0.0028)
<i>Mkt/Book<sub>t</sub></i>	0.016*** (0.0037)	0.023*** (0.0062)	0.0030*** (0.00056)	0.0018*** (0.00042)
Observations	39,957	13,069	39,542	12,994
R-squared	0.134	0.177	0.006	0.005
No of firms	4,080	2,289	4,057	2,278
Fixed Effects	Firm + Year	Firm + Year	Firm + Year	Firm + Year

Robust standard errors in parentheses

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1



# Associated with TFP

	(1) $TFP_t$	(2) $TFP_t$
$techdiff_t$	0.098*** (0.011)	0.094*** (0.025)
$\log(cites_t)$	0.00094 (0.0032)	0.0072 (0.0052)
$HHI_{SIC,t}$	-0.012* (0.0068)	-0.0085 (0.019)
$trnic3tsimm_t$		0.010 (0.025)
$RDstock / AT_t$	-1.09*** (0.12)	-1.48*** (0.24)
$\log(AT_t)$	0.27*** (0.020)	0.44*** (0.046)
$Mkt / Book_t$	0.17*** (0.012)	0.16*** (0.020)
Observations	30,603	9,041
R-squared	0.182	0.180
No of firms	2,974	1,561
Fixed Effects	Firm + Year	Firm + Year

Robust standard errors in parentheses

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1

## Research Question

How do firms' technology positions affect their expected returns?

- ▶ High  $Q$  firms have low returns (value premium)
- ▶ Tuzel and Imhoroglu (2013): high TFP firms have low returns
  - ▶ Over and above the value premium

# TFP, Q, and Expected Returns

Two big streams:

- ▶ Adjustment Costs (Novy-Marx, Tuzel and Imohoroglu)
  - ▶ Endogenous heterogeneity in position on production function
  - ▶ No optionality, NPV rule holds
- ▶ Growth Options (Berk Green Naik, Carlson Fisher Giammarino, Garleanu Panageas Yu)
  - ▶ Uncertainty + irreversibility
  - ▶ Firms have heterogeneous investment *options*

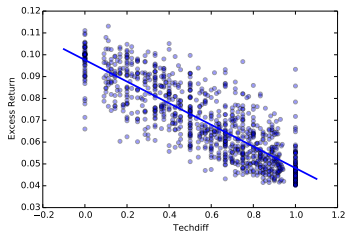
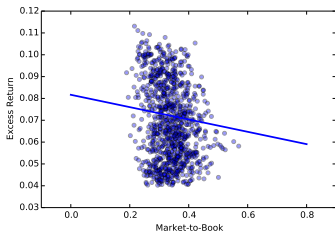
# Model

- ▶ Firms have growth options on differentiated products
- ▶ Invest  $I$  to get a product that yields stochastic cash flows
- ▶ More differentiated products have lower risk, lower returns (Hou and Robinson)

## Model

- ▶ Firms wait until  $NPV \geq D^* > I$ , so NPV rule does not hold
- ▶ Once exercised the product is on the balance sheet

# Predictions



- ▶ Market-to-book ratio reflects product market position
  - ▶ Mechanical value premium
- ▶ Technology position is correlated but separate
  - ▶ Tech differentiation also predicts lower returns

# Predictions

1. Higher *techdiff* predicts lower returns, over and above the value premium
2. Product market position does not
3. Pattern stronger in growth firms

# Returns

	(1) <i>rtn - rf</i>	(2) <i>rtn - rf</i>	(3) <i>rtn - rf</i>
<i>techdiff</i>	-0.032*** (0.0068)	-0.034*** (0.0067)	-0.061*** (0.016)
<i>log(cites)</i>		0.020*** (0.0068)	
<i>RD/sales</i>		0.15 (0.13)	-0.0100 (0.015)
<i>HHI<sub>SIC</sub></i>		-0.0033 (0.0043)	-0.0054 (0.011)
<i>TNIC3tsimm</i>			0.017 (0.023)
<i>log(Size)</i>	-0.063*** (0.013)	-0.064*** (0.013)	-0.087*** (0.029)
<i>Book / Mkt</i>	0.018** (0.0081)	0.017** (0.0085)	0.000060 (0.018)
<i>rtn<sub>t-1</sub></i>	-0.12*** (0.0087)	-0.12*** (0.0094)	-0.080*** (0.016)
<i>rtn<sub>t-12,t-1</sub></i>	0.036*** (0.011)	0.040*** (0.011)	-0.018 (0.022)
Observations	496,741	406,698	156,266
R-squared	0.060	0.085	0.075

Standard errors in parentheses

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1



# Returns, Interacted

	(1) <i>rtn - rf</i>	(2) <i>rtn - rf</i>
<i>techdiff</i>	-0.053*** (0.010)	-0.0035** (0.0015)
<i>techdiff</i> × <i>Book / Mkt</i> <sub>quintile</sub>	0.0070*** (0.0026)	
<i>Book / Mkt</i> <sub>quintile</sub>	0.0057 (0.0056)	
<i>techdiff</i> × <i>Size</i> <sub>quintile</sub>		0.00044 (0.00032)
<i>Size</i> <sub>quintile</sub>		-0.0029*** (0.00072)
<i>log(Size)</i>	-0.065*** (0.013)	
<i>Book / Mkt</i>		0.0019*** (0.00065)
<i>rtn</i> <sub><i>t</i>-1</sub>	-0.12*** (0.0086)	-0.0098*** (0.00073)
<i>rtn</i> <sub><i>t</i>-12,<i>t</i>-1</sub>	0.037*** (0.011)	0.0030*** (0.00089)
Observations	497,311	496,741
R-squared	0.064	0.062

Standard errors in parentheses

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1

# Portfolio Sort

		<i>techdiff</i> Portfolio					
		1	2	3	4	5	5 minus 1
Size Portfolio	1	0.0129 (0.0070)	0.0079 (0.0044)	0.0095 (0.0041)	0.0071 (0.0035)	0.0047 (0.0033)	-0.0082 (0.0054)
	2	0.0090 (0.0048)	0.0026 (0.0036)	0.0003 (0.0033)	-0.0011 (0.0032)	-0.0026 (0.0031)	-0.0116 (0.0034)
	3	0.0038 (0.0036)	0.0001 (0.0033)	-0.0024 (0.0028)	-0.0027 (0.0029)	-0.0035 (0.0029)	-0.0073 (0.0023)
	4	-0.0001 (0.0028)	-0.0027 (0.0027)	-0.0032 (0.0027)	-0.0032 (0.0028)	-0.0039 (0.0026)	-0.0038 (0.0016)
	5	-0.0041 (0.0022)	-0.0047 (0.0024)	-0.0051 (0.0022)	-0.0042 (0.0023)	-0.0049 (0.0023)	-0.0008 (0.0014)

# Conclusion

- ▶ Use patent text to map firms' technological position
  - ▶ Dynamic (year by year)
  - ▶ Distinct from product market position
  - ▶ Predicts product market position
- ▶ Measure of tech differentiation, *techdiff*
  - ▶ Associated with higher Q, ROA, Capex, TFP
  - ▶ Predicts lower returns
    - ▶ Stronger for growth firms
    - ▶ Fits with model of growth options on heterogeneous products