## **Appendices – On-line Publication Only under Supplementary Materials**

## **Appendix 1 Classifications of Retail Chains (Anchors)**

Panel A provides classification descriptions in Vitorino (2012) and Gould, Pashigian and Prendergast (2005) and classifications of competitors based on the Gale and Hoover business reports. Zhou and Clapp (2015, 2016) develop a list of multi-line anchor stores from the Directory of Major Malls, and sort these stores into three price/quality categories: high, mid, and low following Vitorino (2012). They begin with the specific stores identified by Vitorino for each category, and then drawing on the Gale business report manually classified the remaining stores on their list into the same three categories based on stores that are identified as direct competitors. For example, Wal-Mart is classified as a low-price anchor while Nordstrom is classified as a high-price anchor. In Vitorino, "upscale department stores" (such as Dillard's, Macy's, and Nordstrom) sell designer merchandise at a premium price and provide checkout service and customer assistance in each department. In our application, we define this type of stores as "high-price." Similarly, our "mid-price" ("low-price") anchors resemble "midscale department stores" ("discount department stores") in Vitorino. However, Vitorino's criteria are difficult to apply to some of the anchors in our study. For example, according to Vitorino, TJ Maxx and Marshalls could be classified as either "midscale" or "discount." In this case, we check Gould, Pashigian and Prendergast (2005) and classify them into "midscale" because both have national reputations. Results are robust to categorizing TJ Maxx and Marshalls as low-price.

Finally, as a robustness check, we develop a new categorization of competing multi-line anchors using the CONCOR (CONvergence of iterated CORrelations) algorithm. CONCOR was developed in social network theory and centers on the application of algebraic concepts to examine structural equivalence of actors. A pair of actors is said to exhibit structural equivalence if they are tied to the same set of other actors. In the first step, the relations among subjects are coded as a binary variable in a symmetric matrix based on the Gale (Hoover) business report, resulting in a 49x49 binary matrix based on competition among chains (retail companies) defined in Gale (Hoover). Next, correlations between each pair (row-column) are calculated. The CONCOR algorithm repeatedly correlates the row and column vectors until all entries become either +1 or -1. As a result, this algorithm splits a set of actors into exactly two subsets and can be applied repeatedly to generate sub-sets (in bipartitions) of the actors. Because we have limited number of chains, we only divide anchors into two groups.

Panel B presents the complete classification of all stores based on (1) price-types created by Zhou and Clapp (2015, 2016) following Vitorino (2012) and Gould, Pashigian and Prendergast (2005); (2) CONCOR classification based on competition among retail chains following the Gale Business Report and (3) CONCOR classification based on competition among retail companies following Hoover. The first classification is used throughout the paper. The second two classifications are used in a robustness test that is shown in Panel C.

#### Panel A Classification Descriptions and Criteria

## Classification in Vitorino (2012) pp.177

"Upscale department stores: These stores generally sell designer merchandise above an average price level. When their items are on sale, their prices resemble those of average priced items at a lower-scale department store. Upscale department stores typically provide checkout service and customer assistance in each department. Examples include Dillard's, Macy's, and Nordstrom.

Midscale department stores: These stores sell brand names and non-brand names but do not sell upscale brand names. Compared with upscale department stores, midscale stores usually do not have perfumes and beauty supplies at the main entrance and do not have cosmetic specialists. Examples include JCPenney, Mervyn's, and Kohl's.

Discount department stores: These stores encompass retail establishments selling a variety of merchandise for less than conventional prices. Target, Sears, Wal-Mart, and Kmart are examples. Most discount department stores offer wide assortments of goods; others specialize in merchandise such as jewelry, electronic equipment, or electrical appliances. Discount stores are not dollar stores, which sell goods at a dollar or less. Discount stores differ because they sell branded goods, and prices vary widely among products. Compared with midscale department stores, discounters sell fewer major brand names and offer a wider variety of products. Stores in the discount department store category typically have fewer sales workers, relying more on self-service features, and have centrally located cashiers."

# Classification in Gould, Pashigian and Prendergast (2005) pp.414

"Type 1: Prestige/fashion department stores. These stores usually operated in only one or a few markets until recently, when they expanded into more regional and national markets, often by entering into existing malls.

Type 2: High- to moderate-quality department stores with national reputations since the 1950s and 1960s. These stores were usually in the mall right from the beginning and were eagerly recruited by developers to establish the mall.

Type 3: Lower-quality department stores with mostly local or regional reputations. (30% of anchor stores.)

Type 4: Department stores that are members of very well-known national chains that have long operated in many markets. These stores were also usually in the mall right from the beginning and were eagerly recruited by developers to establish the mall. (41% of anchor stores.)"

## Classification based on Gale and Hoover Business Report

List of Retail Chains	Company	Competitor Chain (Gale)	Competitor Company (Hoover)
Barneys New York	Barneys New York	Saks, Bloomingdale's	Saks, Bloomingdale's
Bealls (Florida)	Bealls	Kohl's, Target, Wal-Mart	Kohl's, TJX, Ross
Bealls Outlet	Bealls	Sears, Wal-Mart	Kohl's, TJX, Ross
Belk	Belk	Saks Fifth Avenue, Dillard's, JCPenney, Macy's	Dillard's, Kohl's, Macy's
McRaes	Belk	Saks Fifth Avenue, Dillard's, JCPenney, Macy's	Dillard's, Kohl's, Macy's
BJ's	BJ's	Sam's club, Costco	Target, Costco
Bon-Ton	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Boston Store	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Carson Pirie Scott	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Elder-Beerman	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Parisian	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Younkers	Bon-ton Stores	Dillard's, Macy's, Boscov's	Dillard's, Macy's, Boscov's
Boscov's	Boscov's	JC Penney, Kohl's, Macy's	JC Penney, Kohl's, Macy's
Burlington Coat Factory	Burlington Coat Factory	Ross Stores, TJ Maxx, Marshalls, AJ Wright, Target	Ross, TJX, Target
Costco	Costco	Sam's club, BJ's	Target, Wal-Mart
Dillard's	Dillard's	Burlington Coat Factory, Kohl's, Macy's, Neiman Marcus, Nordstrom, Saks Fifth Avenue	JC Penney, Kohl's, Macy's
Lord & Taylor	Hudson's Bay Company	Neiman Marcus	Neiman Marcus
JCPenney	JCPenney	Kohl's, Sears, Target, Wal-Mart, Dillard's, Nordstrom, Bon-Ton Stores	Sears, Kohl's, Macy's
Kohl's	Kohl's	Wal-Mart, Target, JCPenney, Sears, Mervyn's, TJ Maxx, Marshalls, AJWright, Ross Stores, Macy's	Target, JC Penney, TJX
Bloomingdale's	Macy's	Neiman Marcus, Saks, Barneys New York	Neiman Marcus, Saks, Barneys New York
Filene's	Macy's	Saks Fifth Avenue, Nordstrom, Dillard's, JCPenney	Dillard's, JC Penney, Saks
Kaufmanns	Macy's	Saks Fifth Avenue, Nordstrom, Dillard's, JCPenney	Dillard's, JC Penney, Saks
L. S. Ayres	Macy's	Saks Fifth Avenue, Nordstrom, Dillard's, JCPenney	Dillard's, JC Penney, Saks
Macy's	Macy's	Saks Fifth Avenue, Nordstrom, Dillard's, JCPenney	Dillard's, JC Penney, Saks
Strawbridge's	Macy's	Saks Fifth Avenue, Nordstrom, Dillard's,	Dillard's, JC Penney, Saks

**JCPenney** Meijer, Inc. Meijer Wal-Mart, Target Target, Wal-Mart Barneys New York, Macy's, Bon-Ton Stores, Neiman Marcus Dillard's, Nordstrom, Saks Fifth Avenue, Von Saks, Barneys New York, Bloomingdale's Neiman Marcus Maur Neiman Marcus Neiman Marcus Nordstrom Rack, Saks Fifth Off last call Saks Fifth Avenue, Neiman Marcus, Dillard's, Nordstrom Nordstrom Neiman Marcus, Saks, Bloomingdale's Macy's, JCPenney, Sears Nordstrom Rack Nordstrom Neiman Marcus last call, Saks Fifth Off TJ Maxx, Marshalls, AJ Wright, Kohl's, Target, Ross Ross Stores Burlington Coat Factory, JC Penney, Stein Mart, Kohl's, TJX, Wal-Mart Goody's Barneys New York, Bloomingdale's, Neiman Neiman Marcus, Barneys New York, **Proffitts** Saks Marcus, Macy's, Dillard's, Nordstrom Bloomingdale's Barneys New York, Bloomingdale's, Neiman Saks Fifth Neiman Marcus, Barneys New York, Saks Avenue Marcus, Macy's, Dillard's, Nordstrom Bloomingdale's Off 5th Saks Nordstrom Rack, Neiman Marcus last call Fifth Avenue Saks Outlet Kmart Sears Wal-Mart, Target, JCPenney, Kohl's Wal-Mart, Macy's Wal-Mart, Target, JCPenney, Kohl's Wal-Mart, Macy's Sears Sears Pamida Shopko Wal-Mart, Target, Kmart Target, K-Mart, Wal-Mart Wal-Mart, Target, Kmart Shopko Shopko Target, K-Mart, Wal-Mart JC Penney, Sears, Ross Stage Stores JC Penney, Sears, Ross Goodys J.C. PENNEY, Sears, Ross Peebles Stage Stores JC Penney, Sears, Ross JC Penney, TJ Maxx, Marshalls, AJ Wright, Stein Mart Stein Mart TJX, Macy's, Belk Macy's Sun Capital Mervyn's Ross, Kohl's Macy's, Ross, Kohl's Partners Costco, Macy's, JCPenney, Kohl's, Sears, Wal-Target Target Wal-Mart, Sears Mart TJX AJ Wright Kohl's, JCPenney, Target, Macys Target, Ross, Kohl's Marshalls TJX Kohl's, JCPenney, Target, Macys Target, Ross, Kohl's

Kohl's, JCPenney, Target, Macys

Macy's, Nordstrom

Costco, Target, Kmart

BJ's, Costco

Target, Ross, Kohl's

Macy's, Nordstrom

Target, Costco

Target, Costco

TJ Max

Von Maur

Sam's club

Supercenters

Walmart

TJX

Von Maur

Wal-Mart

Wal-Mart

**Panel B Classification Details** 

List of Retail Chains	Company	Price-Type*	Gale Group**	Hoover Group**
Barneys New York	Barneys New York	High	1	1
Bealls (Florida)	Bealls	Mid	2	2
Bealls Outlet	Bealls	Low	2	2
Belk	Belk	High	1	2
McRaes	Belk	High	1	2
BJ's	BJ's	Low	2	1
Bon-Ton	Bon-ton Stores	High	1	2
Boston Store	Bon-ton Stores	High	1	2
Carson Pirie Scott	Bon-ton Stores	High	1	2
Elder-Beerman	Bon-ton Stores	High	1	2
Parisian	Bon-ton Stores	High	1	2
Younkers	Bon-ton Stores	High	1	2
Boscov's	Boscov's	High	1	2
Burlington Coat Factory	Burlington Coat Factory	Mid	2	2
Costco	Costco	Low	2	1
Dillard's	Dillard's	High	1	2
Lord & Taylor	Hudson's Bay Company	High	1	1
JCPenney	JCPenney	Mid	2	2
Kohl's	Kohl's	Mid	2	2
Bloomingdale's	Macy's	High	1	1
Filene's	Macy's	High	1	2
Kaufmanns	Macy's	High	1	2
L. S. Ayres	Macy's	High	1	2
Macy's	Macy's	High	1	1
Strawbridge's	Macy's	High	1	2
Meijer, Inc.	Meijer	Low	2	1
Neiman Marcus	Neiman Marcus	High	1	1
Neiman Marcus last call	Neiman Marcus	Mid	2	1
Nordstrom Nordstrom Rack	Nordstrom	High Mid	1	1
Ross	Nordstrom Ross Stores	Mid	2 2	1 2
Proffitts	Saks	High	1	1
Saks Fifth Avenue	Saks	High	1	1
Off 5th Saks Fifth Avenue Outlet	Saks	Mid	2	1
Kmart	Sears	Low	2	2
Sears	Sears	Low	2	2
Pamida	Shopko	Mid	2	1
Shopko	Shopko	Low	2	1
Goodys	Stage Stores	Mid	2	2
Peebles	Stage Stores	Mid	2	2
Stein Mart	Stein Mart	Mid	2	2
Mervyn's	Sun Capital Partners	Mid	2	2
Target	Target	Low	2	1
AJ Wright	TJX	Mid	2	2
Marshalls	TJX	Mid	2	2
TJ Max	TJX	Mid	2	2
Von Maur	Von Maur	High	1	2
Sam's club	Wal-Mart	Low	2	1
Walmart Supercenters	Wal-Mart	Low	2	1

<sup>\*</sup>Following Zhou and Clapp (2015, 2016), Vitorino (2012), Gould, Pashigian and Prendergast (2005); \*\* Based on CONCOR algorithm

## Panel C: Robustness Test – Alternative Classification of Anchor Type

Probit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. Anchors are classified based on CONCOR algorithm. In Panel A, competitor chains are identified for each anchor based on Gale Business Report summarized in Appendix 1B. Based on CONCOR algorithm, Group 1 anchors include Barneys New York, Proffitts, Neiman Marcus, Belk, McRaes, Macy's, Strawbridge's, Bloomingdale's, Saks Fifth Avenue, Kaufmanns, LS.Ayres, Filene's, Lord & Taylor, Nordstrom, Bon-Ton, Younkers, Dillard's, Carson Pirie Scott, Boscov's, Boston Store, Elder-Beerman, Parisian and Von Maur. Group 2 anchors include JCPenney, Kohl's, Goodys, Peebles, Bealls (Florida) Mervyn's, Sears, Ross, Burlington Coat Factory, Bealls Outlet, TJ Max, Meijer, Marshalls, Pamida, Shopko, Target, Stein Mart, Walmart Supercenters, AJ Wright, Kmart, BJ's, Nordstrom Rack, Costco, Neiman Marcus last call, Sam's club, Off 5th Saks Fifth Avenue Outlet. Group 1-Open equals 1 if there is any openings of Group 1 anchors within the county and 0 otherwise. Group 1-Close, Group 2-Open and Group 2-Close are defined in a similar way. Group 1/Group 2 is the number of Group 1/Group 2 anchors pre-existing within the county at the beginning of the year of the opening/closing. In Panel B, competitor chains are identified for each anchor based on Hoover's business report summarized in Appendix 1B. Based on CONCOR algorithm, Group 1 company include Neiman Marcus, Barneys New York, Nordstrom, Saks, Macy's, Lord & Taylor, Meijer, Target, Costco, Shopko, BJ's, Wal-Mart. Group 2 company include TJX, Burlington Coat Factory, Stage Stores, Ross Stores, Bealls, Belk, Boscov's, Sears, Bon-ton Stores, Dillard's, Stein Mart, JCPenney, Kohl's and Von Maur. Group 1-Open equals 1 if there is any openings of Group 1 anchors within the county and 0 otherwise. Group 1-Close, Group 2-Open and Group 2-Close are defined in a similar way. Group 1/Group 2 is the number of Group 1/Group 2 anchors pre-existing within the county at the beginning of the year of the opening/closing. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for t-statistics > 2.58; \*\* for t-statistics > 1.96; and \* for t-statistics > 1.65.

#### Gale Classification

	Group 1		Group 2	
	Open	Close	Open	Close
P(Open)	0.019	0.039	0.287	0.054
	BC3p	BC3p	BC3p	BC3p
Coefficients				
Group 1	-0.706***	0.850***	-0.070	-0.209
	(0.235)	(0.253)	(0.207)	(0.251)
Group 2	-0.074	0.011	-0.176**	-0.001
	(0.071)	(0.053)	(0.073)	(0.087)
Marginal Effects				
Group 1	-0.017***	0.037***	-0.003	-0.005
	(0.006)	(0.010)	(0.007)	(0.006)
Group 2	-0.002	0.001	-0.007***	0.000
	(0.002)	(0.002)	(0.003)	(0.002)
% Change				
Group 1	-88%	95%	-1%	-9%
Group 2	-10%	3%	-3%	0%
Observations	875	875	875	875
County FE	Y	Y	Y	Y
Year FE	Y	Y	Y	Y

# Hoover Classification

	Group 1		Group 2	
	Open	Close	Open	Close
P(Open)	0.195	0.007	0.183	0.085
	BC3p	BC3p	BC3p	BC3p
Coefficients				
Low	-0.259***	N.A.	0.022	N.A.
	(0.066)		(0.080)	
Mid	0.050	N.A.	-0.415***	N.A.
	(0.065)		(0.082)	
Marginal Effects				
Low	-0.047***	N.A.	0.003	N.A.
	(0.011)		(0.010)	
Mid	0.009	N.A.	-0.057***	N.A.
	(0.011)		(0.010)	
% Change				
Low	-24%	N.A.	2%	N.A.
Mid	5%	N.A.	-31%	N.A.
Observations	875	875	875	875
County FE	Y	Y	Y	Y
Year FE	Y	Y	Y	Y

## Appendix 2 Robustness Test - Adding Population

This table is supplementary to Table 2 and 3 with addition control of population. Probit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. In Panel A, Columns of Low/Mid/High include results using dependent variables based on low/mid/high-price anchors. For example, Low-Open corresponds to dependent variable that equals 1 if there is any openings of low-price within the county and 0 otherwise. Low-Close corresponds to dependent variable that equals 1 if there is any closings of low-price within the county and 0 otherwise. Low/mid/high is the number of low/mid/high-price anchors pre-existing within the county at the beginning of the year of the opening. In Panel B, Columns of Small/Big include results using dependent variables based on small/big-scale anchors. For example, Small-Open corresponds to dependent variable that equals 1 if there is any openings of small-scale within the county and 0 otherwise. Small-Close corresponds to dependent variable that equals 1 if there is any closings of small-scale within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. Population is log of population. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for *t*-statistics > 1.96; and \* for *t*-statistics > 1.65.

Panel A: By Price

	Low		Mid		High	
	Open	Close	Open	Close	Open	Close
P(Open)	0.198	0.026	0.173	0.026	0.025	0.042
	BC3p	BC3p	BC3p	BC3p	BC3p	BC3p
Coefficients						
Low	-0.402***	-0.057	0.066	N.A.	-0.021	-0.161
	(0.093)	(0.227)	(0.089)		(0.129)	(0.116)
Mid	0.164**	-0.332*	-0.609***	N.A.	-0.049	0.134
	(0.075)	(0.192)	(0.115)		(0.131)	(0.127)
High	-0.096	-0.236	0.058	N.A.	-0.522***	0.836***
	(0.135)	(0.254)	(0.154)		(0.200)	(0.244)
Population	-2.289	2.557	0.879	N.A.	0.636	8.556
	(1.654)	(2.781)	(1.698)		(7.405)	(5.553)
Marginal Effects						
Low	-0.070***	-0.002	0.010	N.A.	-0.001	-0.007
	(0.014)	(0.007)	(0.012)		(0.004)	(0.005)
Mid	0.029**	-0.012*	-0.093***	N.A.	-0.002	0.006
	(0.012)	(0.006)	(0.015)		(0.004)	(0.005)
High	-0.017	-0.008	0.009	N.A.	-0.017***	0.038***
	(0.021)	(0.008)	(0.021)		(0.006)	(0.010)
Population	-0.400	0.919	0.135	N.A.	0.021	0.384
	(0.264)	(0.641)	(0.237)		(0.213)	(0.239)
% Change						
Low	-35%	-8%	6%	N.A.	-4%	-17%
Mid	15%	-46%	-54%	N.A.	-8%	14%
High	-9%	-31%	5%	N.A.	-68%	90%
Observations	875	875	875	875	875	875
County FE	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y

Panel B: By Scale

	Small		Big	
	Open	Close	Open	Close
P(Open)	0.101	0.027	0.257	0.066
	BC3p	ВС3р	BC3p	BC3p
Coefficients				
Small	-0.516***	N.A.	-0.141	-0.153
	(0.153)		(0.107)	(0.111)
Big	0.098	N.A.	-0.222***	0.081
	(0.065)		(0.059)	(0.057)
Population	4.837	N.A.	0.124	11.53
	(4.095)		(1.662)	(14.434)
Marginal Effects				
Small	-0.055***	N.A.	-0.029	-0.012
	(0.014)		(0.020)	(0.008)
Big	0.010*	N.A.	-0.046***	0.006
	(0.006)		(0.011)	(0.005)
Population	0.514	N.A.	0.025	0.921
	(0.390)		(0.307)	(1.306)
% Change				
Small	-54%	N.A.	-11%	-18%
Big	10%	N.A.	-18%	9%
Observations	875	875	875	875
County FE	Y	Y	Y	Y
Year FE	Y	Y	Y	Y

## Appendix 3 County-Level Openings - Fixed Effect Logit Regressions with Bias Correction

This table is supplementary to Table 2 of county-level openings. Logit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. In Panel A, Open\_Low/Mid/High equals 1 if there is any low/mid/high-price openings within the county and 0 otherwise. Low/mid/high is the number of low/mid/high-price anchors pre-existing within the county at the beginning of the year of the opening. In Panel B, Open\_Small/Big equals 1 if there is any small/big-scale openings within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. All the regressions are controlled for county and year fixed effect. FE denotes uncorrected fixed effects estimator. BC3 denotes the bias-corrected estimator proposed by Fernandez-Val (2009). BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. \*\*\* for t-statistics > 2.58; \*\* for t-statistics > 1.96; and \* for t-statistics > 1.65.

Panel A: By Price

		Open_Low			Open_Mid			Open_High	
P(Open)		0.198			0.173			0.025	
	FE	BC3	BC3p	FE	BC3	BC3p	FE	BC3	BC3p
Coefficients									
Low	-0.854***	-0.667***	-0.481***	0.117	0.091	0.084	-0.087	N.A.	N.A.
	(0.127)	(0.109)	(0.115)	(0.076)	(0.072)	(0.084)	(0.303)		
Mid	0.267***	0.210**	0.181**	-1.066***	-0.839***	-0.643***	-0.668*	N.A.	N.A.
	(0.094)	(0.087)	(0.079)	(0.144)	(0.124)	(0.128)	(0.390)		
High	-0.271	-0.216	-0.153	0.158	0.123	0.141	-2.267***	N.A.	N.A.
	(0.176)	(0.161)	(0.148)	(0.175)	(0.161)	(0.164)	(0.673)		
Marginal Effec	ts								
Low	-0.122***	-0.115***	-0.085***	0.015	0.014	0.013	-0.002	N.A.	N.A.
	(0.016)	(0.015)	(0.017)	(0.010)	(0.010)	(0.012)	(0.006)		
Mid	0.038***	0.036***	0.032**	-0.137***	-0.131***	-0.103***	-0.013	N.A.	N.A.
	(0.013)	(0.013)	(0.013)	(0.015)	(0.015)	(0.017)	(0.008)		
High	-0.039	-0.037	-0.027	0.020	0.019	0.023	-0.045***	N.A.	N.A.
	(0.025)	(0.025)	(0.024)	(0.023)	(0.022)	(0.024)	(0.011)		
% Change									
Low	-62%	-58%	-43%	9%	8%	8%	-8%	N.A.	N.A.
Mid	19%	18%	16%	-79%	-76%	-60%	-52%	N.A.	N.A.
High	-20%	-19%	-14%	12%	11%	13%	-180%	N.A.	N.A.
Observations	875	875	875	875	875	875	875	875	875
County FE	Y	Y	Y	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y	Y	Y	Y

Panel B: By Scale

		Open_Smal	1		Open_Big			
P(Open)		0.101			0.257			
	FE	BC3	BC3p	FE	BC3	BC3p		
Coefficients								
Small	-0.884***	-0.726***	-0.522***	-0.120	-0.098	-0.132		
	(0.173)	(0.155)	(0.188)	(0.132)	(0.120)	(0.105)		
Big	0.079	0.066	0.065	-0.541***	-0.417***	-0.250***		
	(0.062)	(0.060)	(0.061)	(0.087)	(0.076)	(0.063)		
Marginal Effects								
Small	-0.080***	-0.078***	-0.057***	-0.021	-0.020	-0.028		
	(0.014)	(0.013)	(0.017)	(0.023)	(0.022)	(0.021)		
Big	0.007	0.007	0.007	-0.093***	-0.086***	-0.054***		
	(0.006)	(0.006)	(0.006)	(0.014)	(0.013)	(0.012)		
% Change								
Small	-79%	-77%	-56%	-8%	-8%	-11%		
Big	7%	7%	7%	-36%	-33%	-21%		
Observations	875	875	875	875	875	875		
County FE	Y	Y	Y	Y	Y	Y		
Year FE	Y	Y	Y	Y	Y	Y		

## Appendix 4 County-Level Closings - Fixed Effect Logit Regressions with Bias Correction

This table is supplementary to Table 3. Logit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. In Panel A, Close\_Low/Mid/High equals 1 if there is any low/mid/high-price closings within the county and 0 otherwise. Low/mid/high is the number of low/mid/high-price anchors pre-existing within the county at the beginning of the year of the opening. In Panel B, Close\_Small/Big equals 1 if there is any small/big-scale closings within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for t-statistics > 2.6; \*\* for t-statistics > 2.3; and \* for t-statistics > 1.96.

Panel A: By Price

	Close_Low	Close_Mid	Close_High
P(Open)	0.026	0.026	0.042
	BC3p	BC3p	BC3p
Coefficients			
Low	N.A.	N.A.	-0.229*
			(0.120)
Mid	N.A.	N.A.	0.195
			(0.124)
High	N.A.	N.A.	0.913***
			(0.292)
Marginal Effects			
Low	N.A.	N.A.	-0.011**
			(0.005)
Mid	N.A.	N.A.	0.009*
			(0.005)
High	N.A.	N.A.	0.043***
			(0.012)
% Change			
Low	N.A.	N.A.	-26%
Mid	N.A.	N.A.	21%
High	N.A.	N.A.	102%
Observations	875	875	875
County FE	Y	Y	Y
Year FE	Y	Y	Y

Panel B: By Scale

	Open_Big	Wal-Mart	
P(Open)	0.027	0.066	
	BC3p	ВС3р	
Coefficients			
Small	N.A.	-0.036	
		(0.094)	
Big	N.A.	0.029	
		(0.055)	
Marginal Effects			
Small	N.A.	-0.003	
		(0.007)	
Big	N.A.	0.002	
		(0.004)	
% Change			
Small	N.A.	-5%	
Big	N.A.	3%	
Observations	875	875	
County FE	Y	Y	
Year FE	Y	Y	

# Appendix 5 Openings in Growth, Stable and Decline Markets - Fixed Effect Logit Regressions with Bias Correction

This table is supplementary to Table 4. Logit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. Counties are classified into "Growth", "Stable" and "Decline" based on average growth rate of employment in retail trades from 1995Q1 to 2005Q1. For example "Growth" counties have growth rates greater than 67<sup>th</sup> percentile. "Decline" counties have growth rates less than 33<sup>rd</sup> percentile. There are 125 counties, among which 42 counties are classified as "Growth", 42 counties are classified as "Decline" and the remaining 41 are classified as "Stable". In Panel A, Open\_Low/Mid/High equals 1 if there is any low/mid/high-price closings within the county and 0 otherwise. Each specification contains three subsamples, "Growth", "Stable" and "Decline". Low/mid/high is the number of low/mid/high-price anchors preexisting within the county at the beginning of the year of the opening. In Panel B, Open\_Small/big equals 1 if there is any small/big-scale closings within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for *t*-statistics > 2.58; \*\* for *t*-statistics > 1.96; and \* for *t*-statistics > 1.65.

Panel A: By Price

	Open_Low			Open_Mid			Open_High		
	Growth	Stable	Decline	Growth	Stable	Decline	Growth	Stable	Decline
P(Open)	0.241	0.213	0.139	0.235	0.202	0.082	0.031	0.017	0.027
	BC3p	BC3p	BC3p						
Coefficients									
Low	-0.795***	-0.410***	-0.366***	0.164	-0.065	0.262*	N.A.	N.A.	N.A.
	(0.159)	(0.188)	(0.165)	(0.139)	(0.159)	(0.145)			
Mid	0.166	0.092	0.360*	-1.022***	-0.580***	-0.483***	N.A.	N.A.	N.A.
	(0.159)	(0.124)	(0.161)	(0.191)	(0.188)	(0.180)			
High	-0.429	-0.147	0.122	0.332	0.173	-0.212	N.A.	N.A.	N.A.
	(0.306)	(0.219)	(0.321)	(0.291)	(0.238)	(0.286)			
Marginal Effec	ets								
Low	-0.150***	-0.074***	-0.053***	0.032	-0.011	0.022*	N.A.	N.A.	N.A.
	(0.027)	(0.028)	(0.020)	(0.025)	(0.025)	(0.012)			
Mid	0.031	0.017	0.052**	-0.197***	-0.101***	-0.041***	N.A.	N.A.	N.A.
	(0.028)	(0.020)	(0.021)	(0.029)	(0.028)	(0.014)			
High	-0.081	-0.026	0.018	0.064	0.030	-0.018	N.A.	N.A.	N.A.
	(0.053)	(0.036)	(0.043)	(0.051)	(0.038)	(0.023)			
% Change									
Low	-62%	-35%	-38%	14%	-5%	27%	N.A.	N.A.	N.A.
Mid	13%	8%	37%	-84%	-50%	-50%	N.A.	N.A.	N.A.
High	-34%	-12%	13%	27%	15%	-22%	N.A.	N.A.	N.A.
Observations	294	287	294	294	287	294	294	287	294
County FE	Y	Y	Y	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y	Y	Y	Y

Panel B: By Scale

	Open_Small			Open_Big		
	Growth	Stable	Decline	Growth	Stable	Decline
P(Open)	0.143	0.111	0.048	0.313	0.282	0.177
	BC3p	BC3p	BC3p	ВС3р	BC3p	BC3p
Coefficients						
Small	-1.0232***	-0.274	-0.526*	-0.231	-0.160	-0.029
	(0.474)	(0.248)	(0.287)	(0.167)	(0.251)	(0.218)
Big	0.092	0.037	0.035	-0.338***	-0.379***	-0.049
	(0.114)	(0.099)	(0.114)	(0.084)	(0.104)	(0.101)
Marginal Effects						
Small	-0.154***	-0.034	-0.030*	-0.058	-0.033	-0.005
	(0.048)	(0.026)	(0.014)	(0.038)	(0.046)	(0.034)
Big	0.011	0.005	0.002	-0.085***	-0.078***	-0.009
	(0.015)	(0.011)	(0.006)	(0.018)	(0.018)	(0.016)
% Change						
Small	-108%	-31%	-63%	-19%	-12%	-3%
Big	8%	5%	4%	-27%	-28%	-5%
Observations	294	287	294	294	287	294
County FE	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y

## Appendix 6 Openings inside Shopping Centers - Fixed Effect Logit Regressions with Bias Correction

This table is supplementary to Table 3. Logit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. In Panel A, Open\_Low/Mid/High equals 1 if there is any low/mid/high-price openings inside pre-existing shopping centers within the county and 0 otherwise. Each specification contains "All" based on all pre-existing shopping centers and "Large SC" based on pre-existing shopping centers with GLA>400,000 sq.ft. Low/mid/high is the number of low/mid/high-price anchors pre-existing within the county at the beginning of the year of the opening. In Panel B, Open\_Small/Big equals 1 if there is any small/big-scale openings inside pre-existing shopping centers within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for *t*-statistics > 2.58; \*\* for *t*-statistics > 1.96; and \* for *t*-statistics > 1.65.

Panel A: By Price

	Open_Low		Open_Mid		Open_High	
	All	Large SC	All	Large SC	All	Large SC
P(Open)	0.137	0.061	0.159	0.067	0.025	0.017
	BC3p	BC3p	BC3p	BC3p	BC3p	BC3p
Coefficients						
Low	-0.424***	-0.207***	0.085	0.040	-0.050	N.A.
	(0.117)	(0.090)	(0.079)	(0.074)	N.A.	
Mid	0.222**	0.047	-0.571***	-0.188	-0.182	N.A.
	(0.087)	(0.109)	(0.128)	(0.136)	N.A.	
High	-0.132	0.001	0.230	0.249*	-0.663	N.A.
	(0.139)	(0.153)	(0.177)	(0.137)	N.A.	
Marginal Effects						
Low	-0.059***	-0.017***	0.013	0.004	N.A.	N.A.
	(0.013)	(0.006)	(0.011)	(0.006)		
Mid	0.031***	0.004	-0.089***	-0.018	N.A.	N.A.
	(0.011)	(0.008)	(0.017)	(0.011)		
High	-0.018	0.000	0.036	0.023*	N.A.	N.A.
-	(0.017)	(0.011)	(0.025)	(0.011)		
% Change						
Low	-43%	-28%	8%	6%	N.A.	N.A.
Mid	22%	6%	-56%	-26%	N.A.	N.A.
High	-13%	0%	23%	35%	N.A.	N.A.
Observations	875	875	875	875	875	875
County FE	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y

Panel B: By Scale

	Open_Small		Open_E	Big
	All	Large SC	All	Large SC
P(Open)	0.094	0.040	0.195	0.098
	BC3p	BC3p	BC3p	BC3p
Coefficients				
Small	-0.522***	-0.310**	-0.147	-0.143
	(0.182)	(0.142)	(0.106)	(0.107)
Big	0.066	0.129*	-0.210***	-0.078
	(0.062)	(0.075)	(0.065)	(0.064)
Marginal Effects				
Small	-0.053***	-0.018**	-0.027	-0.017
	(0.016)	(0.007)	(0.017)	(0.011)
Big	0.007	0.008*	-0.038***	-0.009
	(0.006)	(0.004)	(0.010)	(0.007)
% Change				
Small	-56%	-46%	-14%	-17%
Big	7%	19%	-20%	-9%
Observations	875	875	875	875
County FE	Y	Y	Y	Y
Year FE	Y	Y	Y	Y

## Appendix 7 Openings of Freestanding Anchors - Fixed Effect Logit Regressions with Bias Correction

This table is supplementary to Table 6. Logit regressions are based on a panel sample of anchor store openings from 2005 to 2011. Counties are units of observations. In Panel A, Open\_Low/Wal-Mart/Target equals 1 if there is any openings of low-price/Wal-Mart/Target within the county and 0 otherwise. Low/mid/high is the number of low/mid/high-price anchors pre-existing within the county at the beginning of the year of the opening. In Panel B, Open\_Big/Wal-Mart/Target equals 1 if there is any openings of big-scale/Wal-Mart/Target within the county and 0 otherwise. Small/big is the number of small/big-scale anchors pre-existing within the county at the beginning of the year of the opening. All the regressions are controlled for county and year fixed effect. BC3p denotes the bias-corrected estimator proposed by Fernandez-Val (2009) when the regressors are treated as predetermined. P(Open) is unconditional probability of openings. % change is calculated as the marginal effect divided by unconditional probability. \*\*\* for *t*-statistics > 2.58; \*\* for *t*-statistics > 1.96; and \* for *t*-statistics > 1.65.

Panel A: By Price

	Open_Low	Wal-Mart	Target	
P(Open)	0.093	0.055	0.021	
	BC3p	BC3p	BC3p	
Coefficients				
Low	-0.037	-0.078	-0.036	
	(0.077)	(0.113)	N.A.	
Mid	0.053	0.127	-0.021	
	(0.095)	(0.113)	N.A.	
High	-0.148	-0.345*	-0.102	
	(0.128)	(0.209)	N.A.	
Marginal Effects				
Low	-0.005	-0.006	N.A.	
	(0.009)	(0.008)		
Mid	0.007	0.010	N.A.	
	(0.011)	(0.008)		
High	-0.019	-0.027*	N.A.	
	(0.014)	(0.014)		
% Change				
Low	-5%	-11%	N.A.	
Mid	8%	18%	N.A.	
High	-20%	-50%	N.A.	
Observations	875	875	875	
County FE	Y	Y	Y	
Year FE	Y	Y	Y	

Panel B: By Scale

	Open_Big	Wal-Mart	Target	
P(Open)	0.106	0.055	0.021	
	BC3p	BC3p	BC3p	
Coefficients				
Small	-0.142	-0.120	N.A.	
	(0.110)	(0.131)		
Big	0.032	0.063	N.A.	
	(0.051)	(0.065)		
Marginal Effects				
Small	-0.021	-0.010	N.A.	
	(0.014)	(0.009)		
Big	0.005	0.005	N.A.	
	(0.007)	(0.005)		
% Change				
Small	-20%	-18%	N.A.	
Big	5%	9%	N.A.	
Observations	875	875	875	
County FE	Y	Y	Y	
Year FE	Y	Y	Y	