Utilizing **"Big Data**"

to Generate Alpha in Portfolios of Consumer Equities

For additional information please visit: BrandLovalties.com/QWAFAFEW

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Over the past few years

"Big Data"

refers to the vast amounts of data generated – consciously or unconsciously – by billions of people every day when they leave **digital tracks** as they simply go about their plugged-in life during the 21st century.

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Who uses the "Big Data" that people leave behind?



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When doing whatever it is that they do ...



Business "Big Data" comes from billions of on-line:

- Product or merchant searches
- Price comparisons / discount hunting
- Product reviews and recommendations (Yelp)
- Store location and/or business hours queries
- Auction entries or bids, on-line ads
- Help, Travel, Hobby and Investment forums
- Social media postings, "Likes" and "Tweets"
- "Siri: Where is the nearest Starbucks?"
- ... Anything that leaves behind digital tracks

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Business "Big Data" comes from billions of on-line:

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- Price comparisons / discount hunting

This "Big Data" is typically massaged every day by corporate product managers and marketing analysts.

- "Siri: Where is the nearest Starbucks?"
- ... Anything that leaves behind digital tracks

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Our topic here:

Turning "Big Data" derived Business Intelligence Into

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Turning "Big Data" into Investment a:

Let's examine several simple questions:

- Are a corporation's customers ultimately responsible for corporate revenue?
- Are there recent corporate revenues that have not become public knowledge yet?
- Would it be a good thing to know about corporate revenues before they become public knowledge?
- Can "Big Data" tell us what a corporation's customers were buying recerily (e.g. yesterday)?
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Turning "Big Data" into Investment α :

Peter Lynch famously exploited the consumer link to revenues while managing the Magellan Fund at Fidelity Investments during 1977-1990

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Turning "Big Data" into Investment α:

- when the average annual return experienced by Magellan's shareholders was over 29%.

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Turning "Big Data" into Investment α:

One of Peter's key initial equity screens involved selecting only those companies that he knew were selling products again and again to his acquaintances.

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Turning "Big Data" into Investment α:

How can we utilize "**Big Data**" to update Peter Lynch's approach for the early 21st Century?



- Brand Names (e.g., iPhone, iPad, iTunes & Mac)
 - Raw Brand "Citation" Rates
 - Opinions ("Likes" and "Unlikes")
- Corporate Names (e.g., AAPL & Apple, Inc.)
 - Raw Citation Rates
 - Perceptions / Sentiment
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What can be **Deduced** from "Big Data"?

- Current popularity of various brand names
 - Seasonally adjusted "citation" rates
 - Peer relative rankings provide market share data
- Changes in citation rates over time
 YOY changes during given quarters anticipate revenue surprises
- Alerts to developing product or service "events"
 - Statistically significant short term citation "blips"
- Brand -> Revenue relationships
 - Including some that are not otherwise publicly available

























Is "Big Data" really useful?

- Conversely, at any given time roughly 60% of all corporations don't have trailing 8 quarter revenue that correlates well to citation rates.
- Examples of corporations that **do not** have high citation -> revenue correlations include:
 - LMT Lockheed Martin Corporation
 XOM Exxon Mobil Corporation
 MON Monsanto Co.

 - RY Royal Bank Of Canada
 DD E. I. du Pont de Nemours and Company

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Does "Big Data" work for Medium/Small Caps?



Lisefulness is theoretically independent of scale of capitalization

AMGN to JACK capitalization ratio: 43:1 AMGN to JACK citation rate ratio: 1:59

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"Big Data" and Capitalization:

- Usefulness is theoretically independent of scale of capitalization
 In principle methods are cap agnostic
 - The industry and scale of consumer operations are the critical factors

Citation rate -> Revenue correlations are strongest for retailers of discretionary durable goods

etail

- Current citation rates
- YOY citation growth rate
- Citation rate -> revenue correlation

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"Big Data" and Capitalization:

- Usefulness is theoretically independent of scale of capitalization
 - In principle methods are cap agnostic
 - The industry and scale of consumer operations are the critical factors
 - However, size matters: larger cap corporations with national markets and retail facilities generally have higher raw citation rates and better signal/noise ratios
- Most important metrics are:
 - Current citation rates
 - YOY citation growth rate
 - Citation rate -> revenue correlation
 - Citation rate -> earnings / equity price correlation
 Quantify the citation rate -> revenue -> earnings -> equity price causality chain



• Signal "Luminosity"

- Baseline customer commerce brand name citations
- "Event" generated citations:
 - Product/Service nightmares (recalls, outages, operational disasters)
 - Corporate nightmares (scandals, M&A, layoffs, fiscal implosions)

• Signal "Relevance"

- Citations -> Revenue correlations
- Citations -> Revenue -> Earnings -> Equity Price causality chain













30%

35%

40%















- "Big Data" collecting issues:
- Time granularity needs to match desired actionable signals
 - Quarterly (or better) resolution for general research
 - Weekly (or better) resolution to anticipate earnings reports
 - Daily resolution for alerts to unfolding PR "events"
- Scale of collected data must match extent of
 - How many languages must be scanned / translated?How many character sets must be utilized?

 - Are Simplified Chinese citations relevant? Mandarin? Cantonese?

"Big Data" collecting issues:

- Dealing with Low Luminosity Signal / Noise Ratios
 - Especially critical for Small / Micro Cap Corporations
 - Aggressive Signal Amplification Methodologies
 - Seeking mean signal level that is >= twice the sample's sigma
 - Adjusting sample time aperture dynamically to create sliding meta-samples
 - Utilize growth percentiles to handle vast citation scale differences

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"Big Data" processing issues:

- Processing requirements often scale geometrically with scope of data
 Pattern searches scale geometrically with number of target variables
- If you are collecting tens of billions of daily communications
 - Think grids with thousands to tens of thousands of processors
 - Think NSA scale processing resources (and "black" budgets)
- Storage of exabyte scale (10¹⁸) historic data can become a challenge
 Data needs to remain "on-line" to facilitate historic correlations
 - Data set integrity does not permit time based purges

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So, how can analysts utilize "Big Data" as a new "fundamental" data source to help Optimize Investment Decisions?



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Synkol	Equity	Percentile Ranking ⁽¹⁾	Enice Correlation(2)	Optimal Leg ⁽³⁾	Industry ⁽⁴⁾	Change Parcentile	Bevenue Correlation Percentile (9)	Brands Moad Porcentile (7)	Event Risk Percentile (f)	Price Correlation Percentile (7)	Equity Moad Parcentile (10)	Investor Interest Percentik (11)
15.	AMERICAN AREA DO DE DE TAC	47.0%	0.40	73	Air FreightDelivery Services	42.9%	61.4%	45.2%	76.4%	40.0%	39.5%	77.9%
1.67	ADVANCE AUTO PARTS, NO	29.9%	0.03	220	Other Specialty Stares	16.5%	24.3%	52.3%	93.1%	62.3%	96.1%	100.0%
1.6%	APPLE NC	80.9%	0.56	132	Computer Manufacturing	47.2%	69.4%	34.8%	72.6%	27.6%	85.6%	25.1%
487	ABOTT LABORATORIES	52.5%	0.50	92	Major Pharmace sticals	11.6%	47.4%	0.7%	2.7%	33.0%	32.8%	15.7%
ala?	ARCTIC LATING	58.7%	0.84	59	Industrial Speciatios	55.1%	94.6%	70.4%	79.6%	1.7%	74.0%	67.3%
a da	ACCENTIBE R.C.		0.40	204	Business Services	74.4%	23.6%	42.5%	22.3%	37.0%	16.8%	71.5%
1.088	ADDRESS TOTEMENT.	47.3%	0.50	173	Computer Software: Prepackaged Software	32.3%	39.5%	50.6%	88.2%	33.3%	73.9%	58.8%
4.00***	ADDIAL AU	25.4%	0.29	203	Shoe Matufacturing	6.1%	22.9%	34.7%	96.8%	46.7%	73.7%	88.4%
100	AUTOMATIC DATA RECEIPTION	64.4%	0.20	71	EDP Services	56.7%	44.0%	18.1%	47.6%	52.7%	21.6%	23.6%
0:1	attan Data	19.4%	0.52	174	Computer Software: Prepackaged Software	29.9%	51.4%	51.8%	4.0%	30.3%	95.0%	30.0%
υt	ation kate	76.6%	0.33	76	Clothing/Bhau/Accessory Stores	93.3%	62.0%	81.7%	40.2%	45.0%	48.2%	9.9%
rrelation to Equity		62.2%	0.02	138	Medical Specialities	28.5%	18.0%	77.7%	96.0%	62.8%	83.6%	84.9%
		12.0%	0.15	249	Savings Institutions	48.4%	91.5%	21.9%	46.9%	55.6%	73.5%	66.5%
riaa	Movemente	85.6%	0.79	111	Accident SHealth Insurance	20.7%	77.6%	63.5%	52.8%	6.1%	35.5%	14.1%
rice	wovements	84.9%	0.49	2	Major Pharmace sticals	21.6%	19.6%	63.0%	5.3%	34.5%	97.7%	5.6%
		90.7%	0.54	121	Property-Casualty Insurers	49.4%	75.9%	87.6%	71.6%	29.8%	27.6%	68.2%
164	ACCURANT INC	30.4%	0.74	46	Accident Mealth Insurance	5.1%	30.3%	69.2%	92.2%	10.0%	91.3%	98.2%
0.588	AGAINST TECHNOLOGIES INC.	95.7%	0.80	13	Business Services	15.4%	47.0%	86.6%	86.1%	4.1%	86.6%	91.3%
0.08	ALCO STORES, NO	93.3%	0.90	67	Department/Specialty Retail Stares	57.9%	4.7%	66.2%	32.2%	0.3%	49.9%	36.2%
18.45	ALASKA AR SECON INC.	19.9%	0.23	188	Air Freight/Delivery Services	17.3%	59.1%	74.0%	25.9%	51.0%	86.8%	56.8%
dia.	THE ALLSTATE CORPORATION	82.0%	0.17	138	Property-Casualty Insurers	15.2%	2.3%	2.2%	27.6%	54.7%	44.2%	52.3%
4,78	41,7254.00	86.0%	0.59	57	Semiconductors	64.6%	38.4%	77.9%	26.3%	23.8%	35.3%	4.9%
185	ACTING TO MERCINE AND AND	35.7%	0.63	132	Semiconductors	5.5%	62.3%	40.2%	61.1%	20.9%	10.7%	58.1%
1801	AMON N.	57.3%	0.26	39	Bistechnology: Biological Products (No Diagnostic Substances)	45.7%	7.2%	77.1%	31.2%	48.7%	4.1%	97.0%
107	AMERICAN TRANSPORT	41.3%	0.55	21	Investment Managers	12.8%	18.9%	67.5%	48.0%	28.8%	73.2%	17.8%
1000	TO AMERICADE HOLDING CORPORATION	27.3%	0.04	45	Investment Bankera/Boloma/Service	99.4%	36.8%	62.0%	24.0%	62.1%	77.1%	0.3%
100	Asses(10x-110x, nc)	5.1%	0.14	178	Catalog/Specialty Distribution	4.3%	15.0%	5.7%		56.1%	38.9%	41.5%
40	AL ¹ (0.4)(1.0)	72.5%	0.14	247	Automotive Altermarket	33.7%	99.5%	95.5%	46.7%	56.7%	94.0%	6.2%
	AREACONDOLE A PROPERTY OF	36.1%	0.89	132	Clothing/Shoa/Accessory Stores	47.8%	5.0%	25.8%	89.3%	0.4%	34.8%	53.4%
1.00		82.64	0.44	4.20	Although and a second second	20.4%	97.7%	15 000		20.0%	52.2%	41.49







"Big Data" Summary:

- For certain classes of equities it can provide useful business intelligence and investment fundamental data
 - Typically those involved in discretionary durable goods trade



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"Big Data" Summary:

- The usefulness of the extracted data depends on:

 - Signal luminosity: baseline commerce and "event" generated
 Signal relevance: correlations to revenue and equity pricing







"Big Data" Summary:

- Such a can be shown to be consistent over an entire market cycle
 - Risk mitigation 2006-2008
 - Explosive upside 2009-2010
 - Persistence during stability 2011-2013



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