

Semiconductors

The 4th Tectonic Shift in Computing: To a Parallel Processing / IoT Model

Key Takeaway

Every 15 years the computing industry experiences a tectonic shift, which dramatically changes the computing model and competitive landscape. We believe we are at the start of a tectonic shift now, driven by low memory prices, improved parallel processors/software, and improved AI technologies like Neural Networks. We view NVDA, AMD, XLNX, CAVM, MXIM, TXN, ADI, MCHP as beneficiaries. We think INTC has the most to lose as the data center incumbent.

There Have Been 3 Tectonic Shifts in Computing. Every 15 years, an accumulation of technical innovations translates to tectonic shifts in the computing model. In the 60s the industry shifted from Mainframes to Mini-Computers, in the early 80s it shifted to PCs, and in the late 90s it shifted to a cell phone / datacenter model. Each computing model shift brought a shift in the beneficiaries: IBM in Mainframes to DEC in MiniComputers, to INTC/MSFT in PCs, to AAPL/Samsung/INTC/MSFT in the cell phone / datacenter model.

We Believe We are at the Start of the 4th Tectonic Shift Now, to a parallel processing / IoT model, driven by lower memory costs, free data storage, improvements in parallel processing hardware and software, and improvements in AI technologies like neural networking, that make it easy to monetize all the data that is being stored.

Those Architected for Parallel Best Positioned. Computing platforms historically relied on higher clock speeds in successive generations for improved platform performance. However Moore Stress is now at play, preventing higher clock speeds translating to higher-and-higher processor core devices. We think those companies that have architected their hardware and software platforms from the ground up for parallel processing are best positioned to benefit - NVDA, XLNX, CAVM and AMD make that list.

NVDA Appears Best Positioned as the Leading Parallel Processing Platform. NVDA was the first to recognize and successfully invest in a HW/SW platform (GPU/CUDA) targeted specifically at parallel processing applications, and our field checks suggest it is years ahead of its competition.

We Also Like AMD, XLNX and CAVM as Parallel Processing Beneficiaries. With their own parallel processing platforms, we think AMD, XLNX and CAVM also benefit as the industry shifts to a parallel processing paradigm. We think the market is underestimating the longer term opportunity for these companies, and upgrade XLNX and CAVM to Buy in conjunction with this note.

We Favor MXIM, TXN, ADI, and MCHP as IoT Beneficiaries. As more companies figure out how to monetize data generated by IoT, we expect higher demand for these devices, which we expect benefits MXIM, TXN, ADI and MCHP.

INTC has the most to Lose as the Dominant Datacenter Processor Incumbent. With dominant market share in the datacenter, we think INTC has the most to lose as the industry shifts to a parallel processing / IoT model. While we don't expect INTC's DCG revenues to disappear or dramatically decline, we do believe that they continue their deceleration to the mid-single digit range for the next 12-24 months. We lower our rating on INTC to Underperform.

Mark Lipacis *

Equity Analyst

(415) 229-1438 mlipacis@jefferies.com

Delos Elder, CFA, CPA *

Equity Associate

(415) 229-1511 delder@jefferies.com

Sean Dorsey, CPA *

Equity Associate

(415) 229-1574 sdorsey@jefferies.com

* Jefferies LLC

^Prior trading day's closing price unless otherwise noted.

Contents

TECTONIC SHIFTS IN COMPUTING	3
Previous Tectonic Shifts in Computing.....	3
<i>Chart 1: Tectonic Shifts in the Computing Paradigm over the Past 60 years</i>	<i>4</i>
The Next Tectonic Shift is Happening— to a Parallel Processing / IoT Paradigm. 5	
<i>Chart 2: Drivers of Tectonic Shift in Computing Platform</i>	<i>6</i>
Jefferies Artificial Intelligence Summit Series	14
IMPORTANCE OF THE PLATFORM – CASE STUDIES.....	17
PARALLEL PROCESSING PLATFORM MAP.....	19
A Guide to Understanding Parallel Processing Platforms (Hardware and Software)	20
<i>Chart 23: Parallel Processing Platform Map</i>	<i>20</i>
Deep Learning Platform Summaries	21
<i>Chart 24: NVDA Deep Learning Platform Summary.....</i>	<i>21</i>
<i>Chart 25: Google Deep Learning Platform Summary.....</i>	<i>22</i>
<i>Chart 26: INTC Deep Learning Platform Summary</i>	<i>23</i>
<i>Chart 27: XLNX Deep Learning Platform Summary.....</i>	<i>24</i>
<i>Chart 28: AMD Deep Learning Platform Summary</i>	<i>25</i>
PARALLEL PROCESSING / IOT WINNERS AND LOSERS.....	26
<i>Chart 29: Tectonic Shift Winners & Losers.....</i>	<i>26</i>

Tectonic Shifts in Computing

Previous Tectonic Shifts in Computing

There has been a tectonic shift in the computing industry about every 15 years

Today's cell phone is a descendant of the 1946 ENIAC, three tectonic shifts removed

Innovations in TTL-transistors led to a tectonic shift to Mini-Computers from Mainframes in the mid-'60s

Mainframes cost \$100,000s
Mini-Computers cost \$10,000s

The tectonic shift to PCs from Mini-Computers was driven by both technical and business innovations

With the PC, computing ASPs declined another 10x into the \$1,000s, while units increased 100x into the 100s of millions

Faster air-interface standards and efficient centralized computing set up a tectonic shift to the Cell-Phone / Datacenter computing model

With the Cell-Phone, computing ASPs declined by 10x and units increased 10x vs the PC

Each tectonic shift in computing model brought a shift in beneficiaries:

- Mainframes -> IBM
- Mini-Computers -> DEC
- PCs -> MSFT + INTC
- Cellphone/Datacenter -> AAPL, MSFT+INTC

Just about every 15 years over the past 60, the accumulation of innovations has caused tectonic shifts in the computing industry. Describing these shifts as tectonic makes sense to us because tectonic shifts can change the landscape to the point where it is not recognizable. Consider that today's computing device of choice, the cellphone, is actually a descendant of ENIAC, the first mainframe computer. The ENIAC was built in 1946 and comprised of 6,000 mechanical switches and 18,000 vacuum tubes. Its footprint was 2,000 square feet; it weighed 30 tons and cost \$500,000. Three tectonic shifts later, you carry your \$600 cellphone in your pocket – it has a \$30 processor, and with wireless access to the cloud, has infinitely more processing power than the ENIAC.

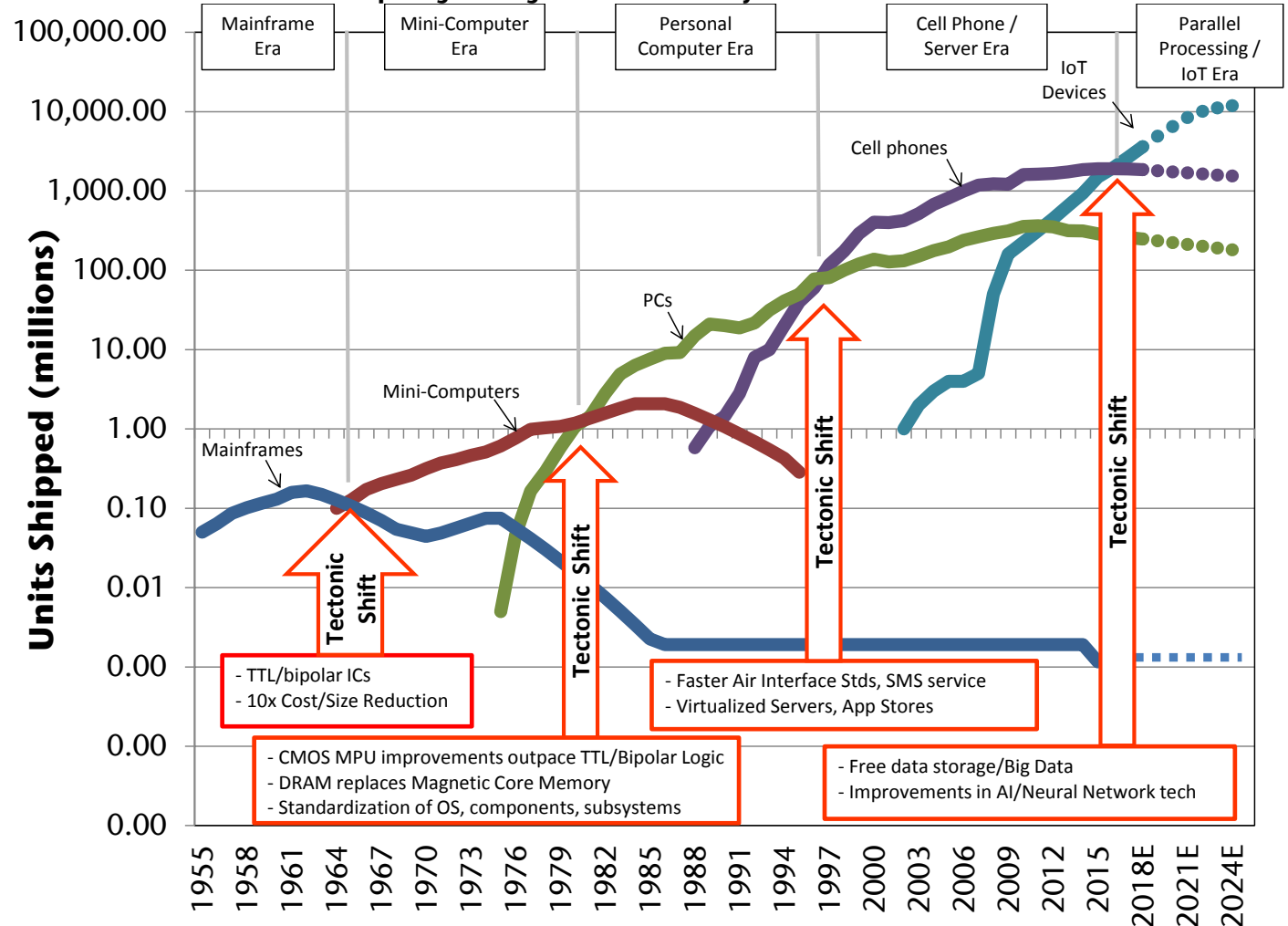
Mainframes in the 1950s saw improvements from the ENIAC, but were still measured in terms of \$100,000s in cost and 1,000s of square feet. By the mid-'60s, improvements in TTL-bipolar transistor technology led to a tectonic shift to mini-computers, the poster-child of which was Boston-based Digital Equipment Corporation (DEC). DEC's mini-computer price-tags were quoted in \$10,000s, and the space required to host them was measured in 10s of square feet. While mainframe sales peaked in the 100,000 unit range annually, annual mini-computer units reached to the 1,000,000 unit range.

The tectonic shift toward Personal Computers was driven by both technological and business innovations. CMOS microprocessor design improvements outpaced TTL-bipolar transistors, which allowed for much higher levels of integration, while DRAM replaced magnetic core memory. So both logic and memory were put on the Moore's Law size and cost model. On the business innovation front, IBM anointed Microsoft and Intel as their PC standard, and IBM shared its PC architecture with the world. The computing industry quickly shifted to a horizontal model from a vertical one, allowing component and subsystem vendors to benefit from global economies of scale, and PC-OEMs to benefit from a low-cost, assembly line manufacturing model. The PC model shifted by two orders of magnitude in units to 100s of millions, and one order of magnitude in costs to 1,000s of dollars.

By the late '90s, faster air-interface standards and efficient, centralized computing in the form of data centers set the computing industry up for a tectonic shift to a cell phone / datacenter paradigm, where a \$30 processor in your hand could potentially get you access to the computing power in all the datacenters of the world. VMWare was founded in 1998 and server virtualization dramatically lowered the cost of server compute cycles. A killer app called SMS was introduced, and Apple and Google created App Stores, harnessing the power of a huge ecosystem of software developers that drove utility into the cell phone today that few could have imagined in the '90s.

Importantly, each tectonic shift in the computing model brought with it a tectonic shift in its beneficiaries. IBM dominated the Mainframe Era, but DEC replaced IBM as the face of the Mini-Computer Era. IBM resurfaced in the PC Era to set PC standards, but Microsoft and Intel ended up controlling the standards and dominating the PC industry profits. The Cell Phone / Datacenter Era was dynamic – Nokia, with its Symbian OS and candybar form factor captured the lion's share of industry profits earlier, but is a shadow of its former self as Apple and distant second (as measured by operating profits) Samsung, dominate today. Intel and Microsoft have been the biggest beneficiaries on the backend of the Cell Phone / Datacenter Era.

Chart 1: Tectonic Shifts in the Computing Paradigm over the Past 60 years



Source: Jefferies Equity Research, Mainframe computer data sourced from IBM Company Filings, "The Early Computer Industry: Limitations of Scale and Scope", A. Gandy; Minicomputer "History of Computer Communications" J. Pelkey; Personal Computer data sourced from "Total Share: Personal Computer Market Share 1975-2010", J. Reimer, Gartner; Mobile devices sourced from Counterpoint Research, Canals Research, "Smartphones" Research Report, M. Ilyas, S. Ahson; IoT devices sourced from Gartner.

The Computing Industry has seen Tectonic Shifts about every 15 years

Each shift in computing model is 10x larger in units, and 10x smaller in ASP

We believe we are at the beginning of the next Tectonic Shift in Computing, toward a Parallel Processing / IoT Paradigm

We believe the computing industry is at the beginning of the next tectonic shift, to a Parallel Processing / IoT model

Price declines in memory have led to “free storage” for many, and are an important driver of this shift

Improvements in AI, Neural Networking, Parallel Processors and Parallel software are also driving this tectonic shift

Parallel processing software is difficult to get right

NVidia recognized the market potential early and invested in its parallel processing platform 10 years ago, well ahead of its peers

Like the PC market in the ‘80s and the cell phone market in the ‘90s, Neural Networking technology is creating markets that don’t exist – consequently they are difficult to forecast

Consistent with the 10x increase in units every tectonic shift, the IoT market is approaching 10 billion

Three tectonic shifts from today, your cellphone will be your grandkids’ ENIAC

The Next Tectonic Shift is Happening– to a Parallel Processing / IoT Paradigm

It has been about 15 years since the shift to the Cell Phone / Datacenter model, and the accumulation of innovations since then leads us to believe that the computing industry is at the beginning of its next tectonic shift.

We view price declines in memory as one marker of this shift. Fifteen years ago, \$10 bought you enough memory to store 500, 2MB photos. Today, that same \$10 gets you enough memory to store over 100,000 photos. That price decline has translated to massive elasticities in demand – to the point where 10s of petabytes of data are now being stored (think of a group of 100 million bits, stored in 1 billion different places). Data storage for many today is practically free.

But if someone is allowing you to store all that data for free, then they likely want to monetize it – right? This is where improvements in Artificial Intelligence, more specifically, Neural Networking, as well as in parallelized software and parallel processing architectures play a role in our tectonic shift thesis.

Computer Science theory has always held that parallel computing architectures could be more efficient than serial ones for many applications. The challenge to that theory, however, had been that there never was good software to distribute your processing stream across parallel processors and then pull the outputs from those processors back together in the right order. This is a very difficult problem to solve. Consequently, computing architectures and software, have been largely serial in nature, up until recently.

However, 10 years ago last month, NVidia launched a software platform, called CUDA, that effectively did just that. The idea was that the CUDA software would enable programmers to leverage the parallel architecture in GPUs for general purpose computing, using widely deployed high-level languages like C++. Initial versions of CUDA weren’t perfect, but five generations later, in 2012, the CUDA efforts bore fruit. A team from the University of Toronto used Deep Neural Networks and NVidia GPUs to smash its competition at the annual ImageNet competition, where teams compete to identify objects from a database of over 10 million images. The event caused a tectonic shift in the AI industry itself, shifting it from a “Knowledge-based” to “Data-based” discipline – and creating demand for GPU compute cycles.

Much like the PC market in the ‘80s and the cell phone market in the ‘90s, the size of the market for Neural Networking and parallel processing is difficult to quantify, because the markets they are creating simply don’t exist. At four different AI conferences Jefferies hosted over the last 8 months (Boston, London, Hong Kong, Zurich) we heard how 30 different startups were using AI and Neural Networking for new business models, for example: 1) identify and sort recyclables from garbage; 2) write large portions of 10Qs and 10Ks from 3 years of financial statements, 3) do facial recognition on video from the far-flung CCTV network in China to catch criminals; and 4) using a wrist-worn heart rate monitor to predict with a 90% accuracy rate whether you will have a heart attack within the next two hours.

The data shows that each new computing model is 10x the size of the previous one in units. If cell phone units are measured in the billions, then the next one has to be in the 10s of billions. Gartner estimates that IoT unit shipments are currently approaching 10 billion units. With data storage effectively free, we expect that the data generated from all these IoT devices will be stored, and processed, using parallel processing platforms and AI technologies. The data tells us that we are at the beginning of the next tectonic shift in computing to a parallel processing and IoT platform paradigm. It is hard to imagine when you are so close to it, but three tectonic shifts from today, the cellphone in your pocket will likely become your grandkids’ ENIAC.

We believe that there have been three tectonic shifts in computing, each of which happened roughly every 15 years:

- 1) From Mainframes to Mini-Computers in the mid-'60s
- 2) From Mini-Computers to PCs in the early 1980s
- 3) From PCs to Mobile Phones / Datacenters in the late '90s

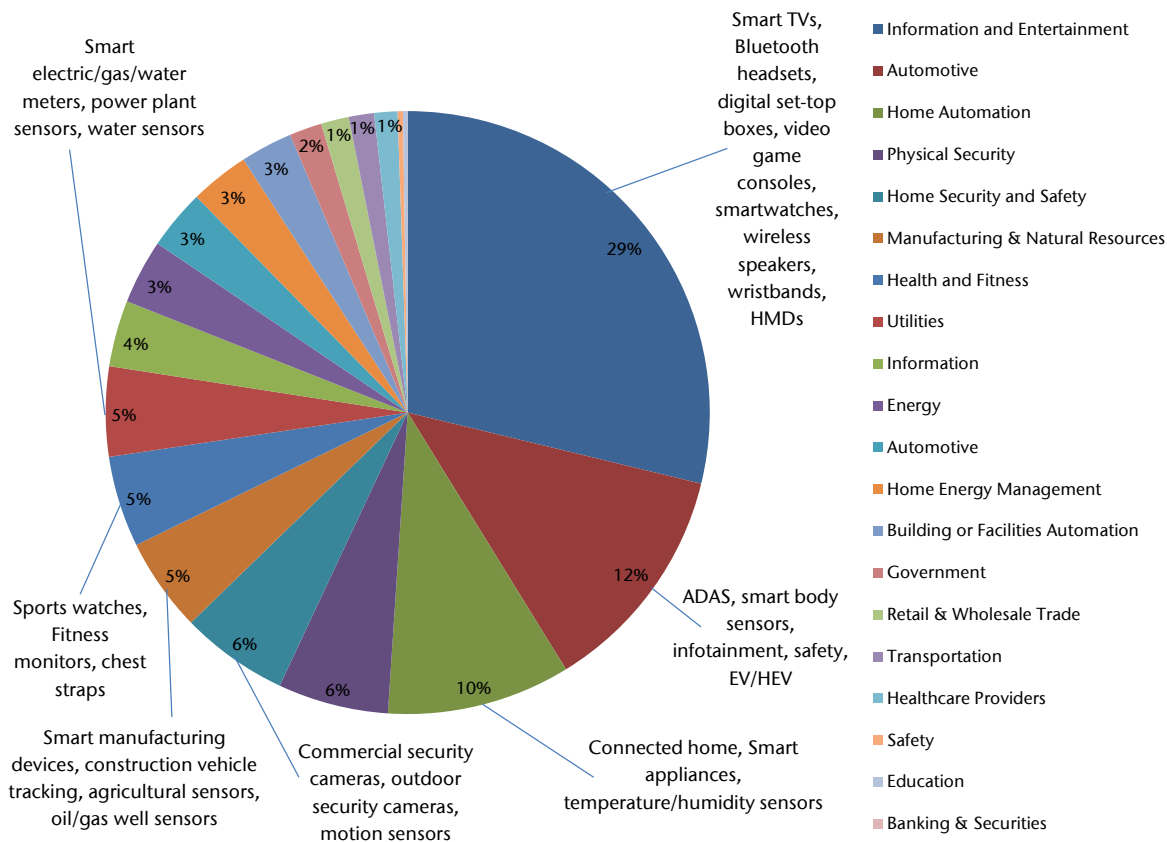
It has been about 15 years since the last tectonic shift, and we think we are at the beginning of the next shift now, to a Parallel Processing / IoT paradigm, driven by improvements in memory costs, as well as by improvements in parallel processing hardware and software, and AI and neural network technologies

Each tectonic shift can be characterized by a 10x magnitude change in price, size and weight. We think that the Parallel Processing / IoT model fits this rule of thumb

Chart 2: Drivers of Tectonic Shift in Computing Platform

<u>Tectonic Shift</u>	<u>Years</u>	<u>Drivers</u>
Mainframe Computers to Minicomputers	Mid 1960s	<ul style="list-style-type: none"> • TTL (transistor-transistor logic) bipolar logic circuits replaced slower DTL (Diode Transistor Logic) circuits • Space requirement lowered to 10s of sq ft from 100s/1000s of sq ft • ASP declines to 10s of thousands of dollars from 100s of thousands of dollars
Mini-Computers to Personal Computers	Early 1980s	<ul style="list-style-type: none"> • CMOS MPU integration outpaces TTL/bipolar logic ICs • DRAM replaces Magnetic Core Logic • Standardization of the OS motivated a software ecosystem • Standardization of components and subsystem enabled lower costs associated with production-line assembly and global economies of scale • ASP lowers to the thousands of dollars from 10s of thousands of dollars • Space requirement lowered to 1 sq ft from 10s of sq ft
PCs to Cell Phones / Servers	Late 1990s	<ul style="list-style-type: none"> • Centralizing Processing Power - Faster Air Interface standards enable a \$30 CPU to potentially access all the processing power located in the data centers in the world • SMS/texting services launched • Virtualized server software lowers the cost/data center compute cycle • Apple and Android App Stores drive utility into smartphones
Cell phones / Servers to Parallel Processing / IoT	Mid-2010s	<ul style="list-style-type: none"> • Low cost of storage and "free" data storage translate to petabytes of data stored • Improvements in AI and Neural Networking technology enable cloud players to process and monetize all the new data stored • Moore Stress changes historical compute cost curve and results in more cores being added to processors

Source: Jefferies; "Rise and Fall of Minicomputers", Engineering and Technology History Wiki

Chart 3: IoT: Internet of Things – Endpoints and Associated Services 2017

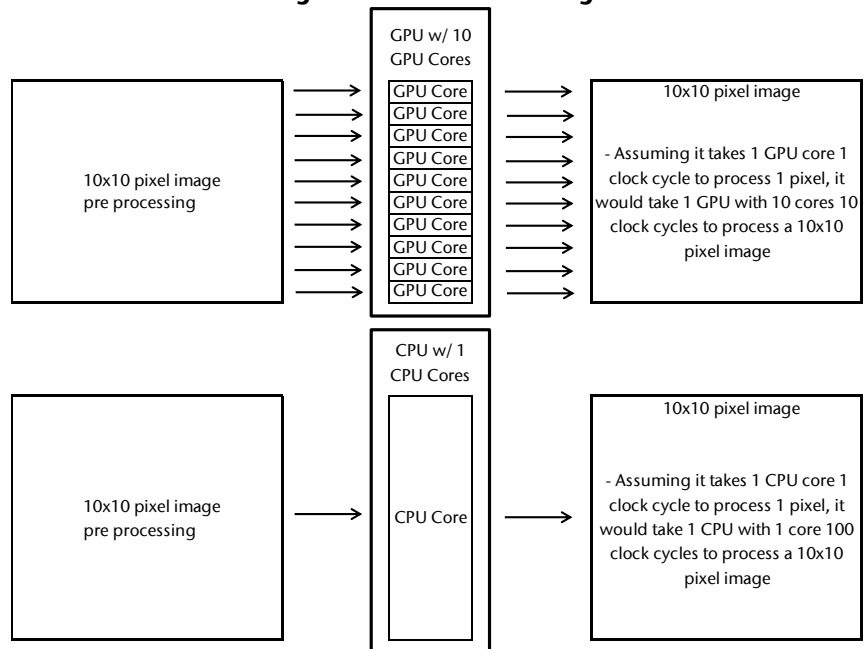
Source: Gartner (December 2016)

Applications run on a serial processor (such as the single core CPU depicted below) use one core at a time and process tasks in sequential order

Compare that to a parallel processor that breaks up tasks amongst several cores and re-assembles the completed tasks

Parallel processors can multiply the processing power of a similarly clocked serial processor *if* the application code is written to optimize the architecture

While CPUs tend to have an easily countable number of cores, GPUs can have thousands of low-powered cores being better suited for running simpler simultaneous calculations (e.g. graphics, matrix algebra)

Chart 4: Parallel Processing versus Serial Processing

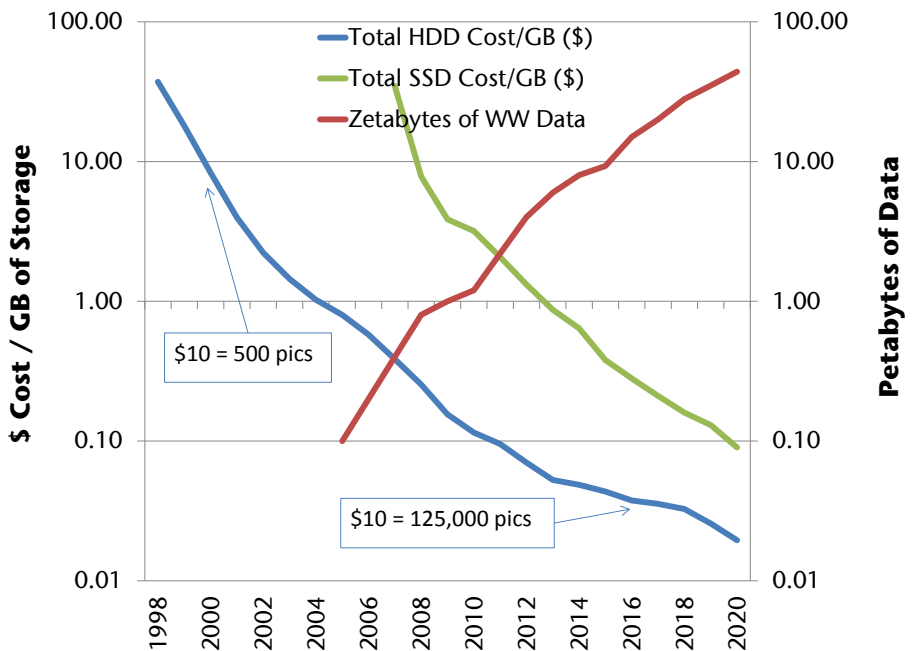
Source: Jefferies

In 2000, \$10 purchased enough memory to store 500, 2MB photos. Today, that same \$10 buys you enough memory to store over 100,000 photos

Fifteen years of memory price declines have driven massive price elasticities in memory demand. Petabytes of memory (imagine 100 million bits stored in 1 billion different places) are now being stored at very low cost

Today, for many, data storage is effectively free

Chart 5: Data Growth versus Data Storage Cost

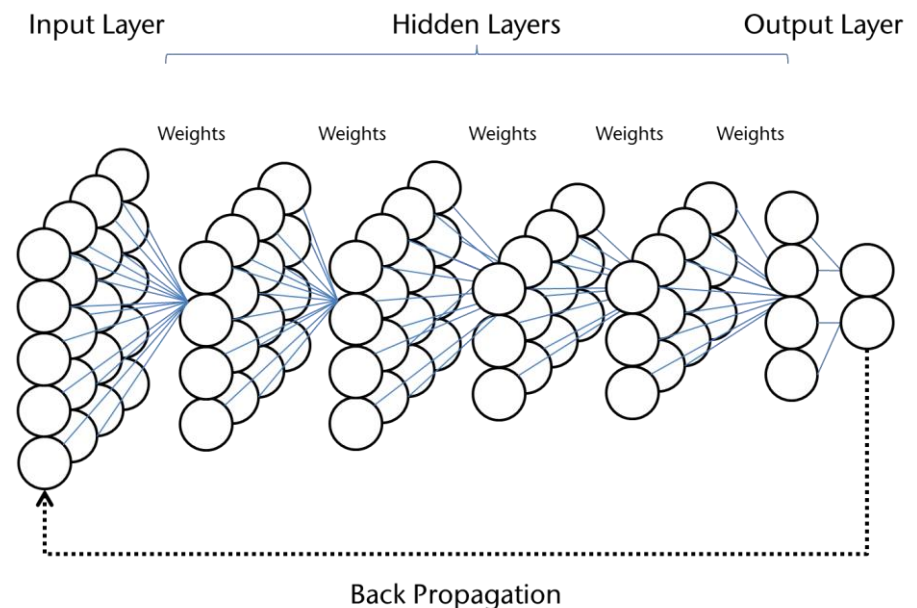


Source: Jefferies, Gartner

Improvements in parallel software and parallel processing architectures, like Nvidia's CUDA and graphics processors, have enabled powerful parallel processing AI frameworks like deep neural networks, to mine and monetize all the data that is being stored

As successful as some companies have been at harnessing the power of deep neural networks (Facebook, Google, Amazon, Baidu), these technologies have only recently been deployed in scale, we still consider the industry to be in its early stages

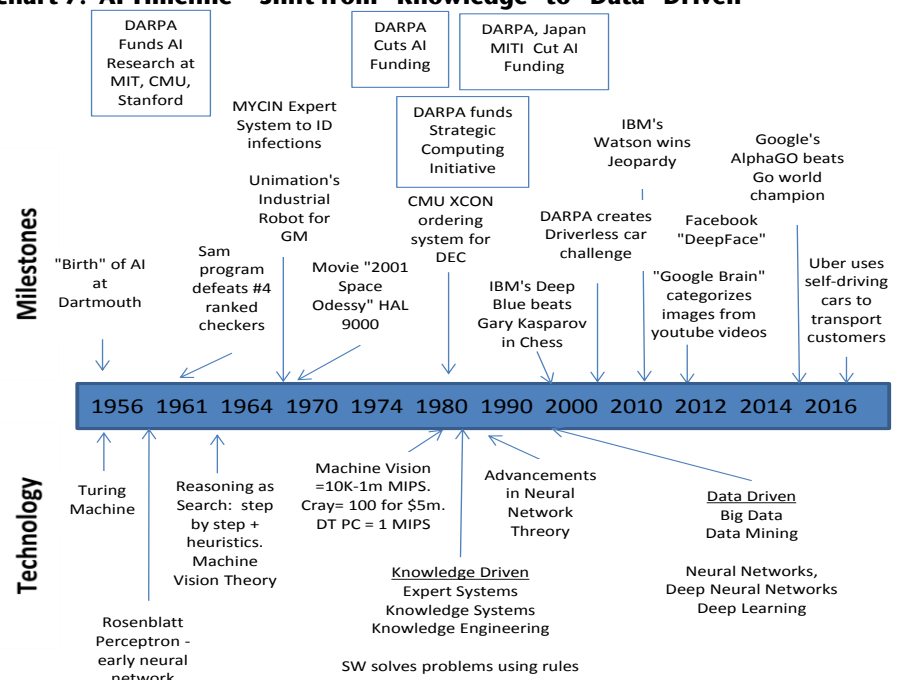
Chart 6: Depiction of a Neural Network



Source: Jefferies

Deep Neural Networks and easily programmed parallel processing platforms have shifted the AI industry from a “Knowledge” driven model to a “Data” driven one

Chart 7: AI Timeline – Shift from “Knowledge” to “Data” Driven

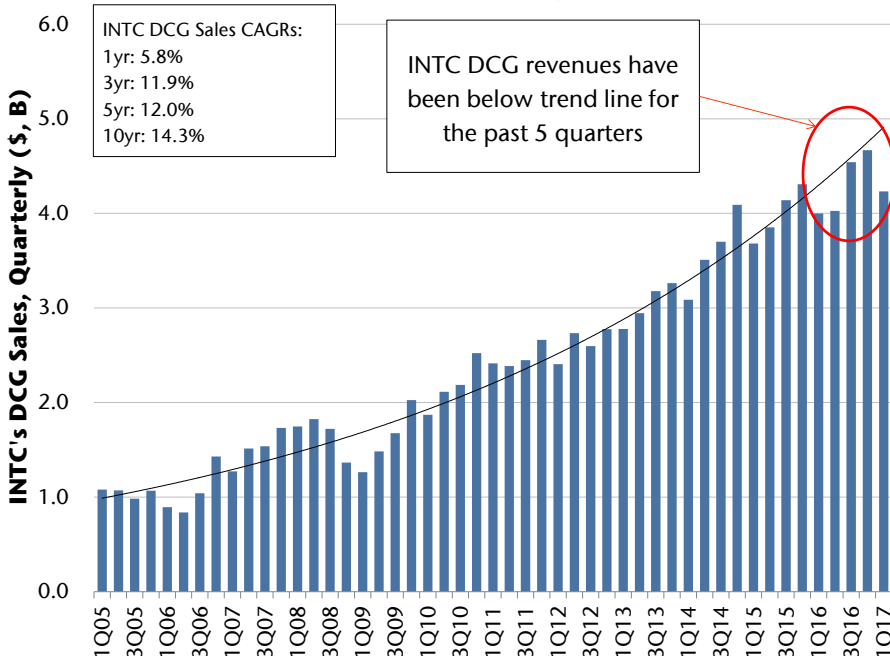


Source: Jefferies, Public Domain

Intel's Datacenter business has grown at a low-to-mid double-digit CAGR over the past 3, 5 and 10 years

However, its 1-yr growth rate decelerated to 6% in the most recent quarter, and the past 5 quarters have been below trendline

Chart 8: INTC Datacenter Processor Revenues - \$

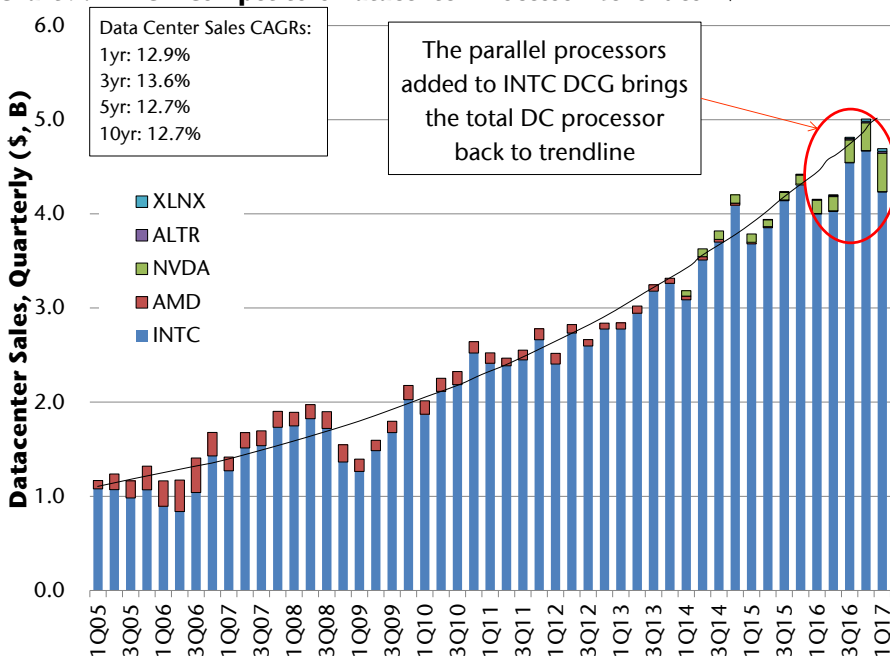


Source: Jefferies

When we add revenues from companies that sell parallel processors to revenues from Intel's Datacenter business, the aggregate datacenter processor revenue bars come back up to trendline

This suggests that processor growth in the datacenter has shifted to parallel processors from Intel x86 processors

Chart 9: INTC + Competitors Datacenter Processor Revenues - \$

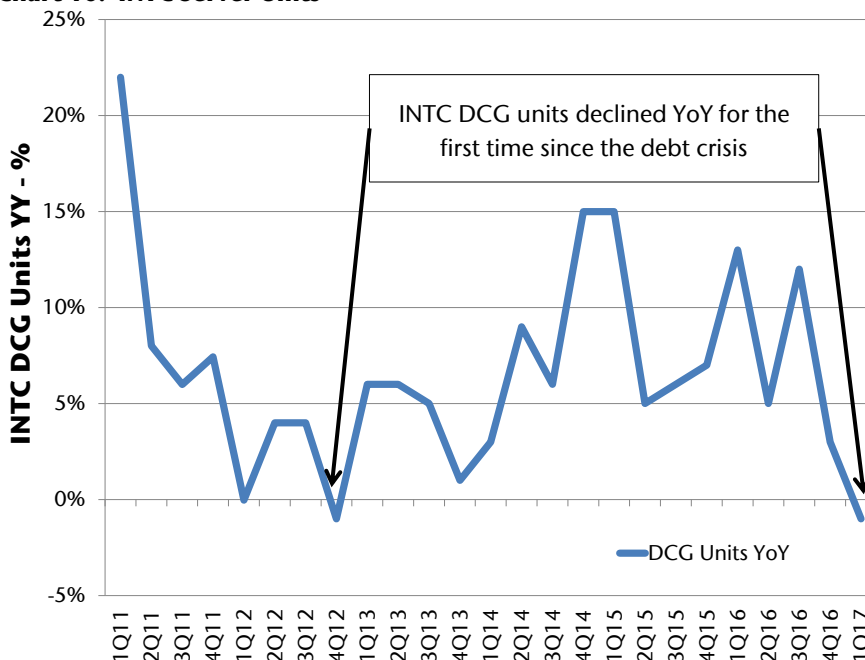


Source: Jefferies

Intel Datacenter MPU unit shipments declined in 1Q17 for the first time since the debt crisis

Intel's 1Q17 suffered from a difficult YoY comparison due to an extra week in the 1Q16 quarter. Intel also may have suffered from an "air-pocket" in demand in front of its next gen server MPU, Purley. That said, we still believe something secular is going on and expect Intel DCG growth rates to remain subdued

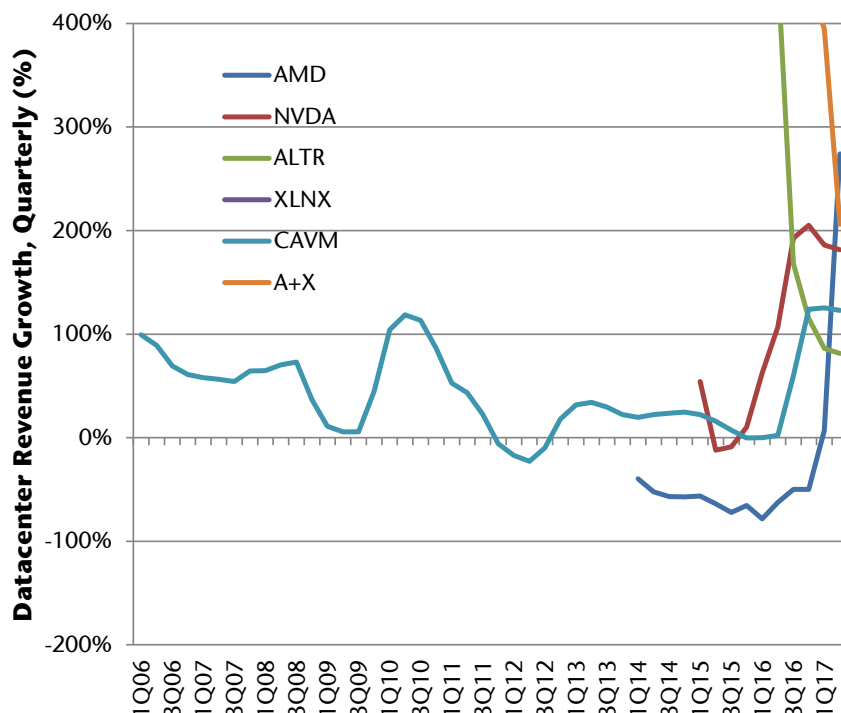
Chart 10: INTC Server Units



Source: Jefferies

Other companies that sell processors into the datacenter are posting impressive growth rates, albeit largely off of small bases

Chart 11: Datacenter Processor Growth - %

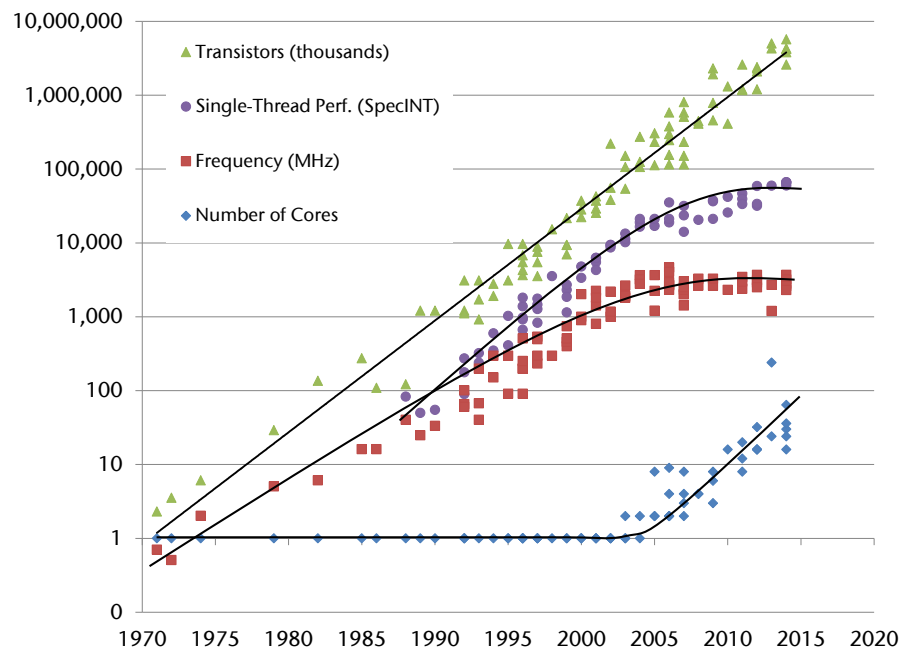


Source: Jefferies, Company Data

Multiple sources highlight that while microprocessors appear to be staying on Moore's Law transistor density curve, that single threaded performance and clock frequency have not kept up, suggesting that the MPU industry could be at an inflection

The bottom line on this chart suggests that MPU makers are trying to make up for the stagnating single threaded performance with more processor cores per MPU – consistent with our thesis that the industry is moving to a parallel processing paradigm

Chart 12: 40 Years of Microprocessor Trend Data

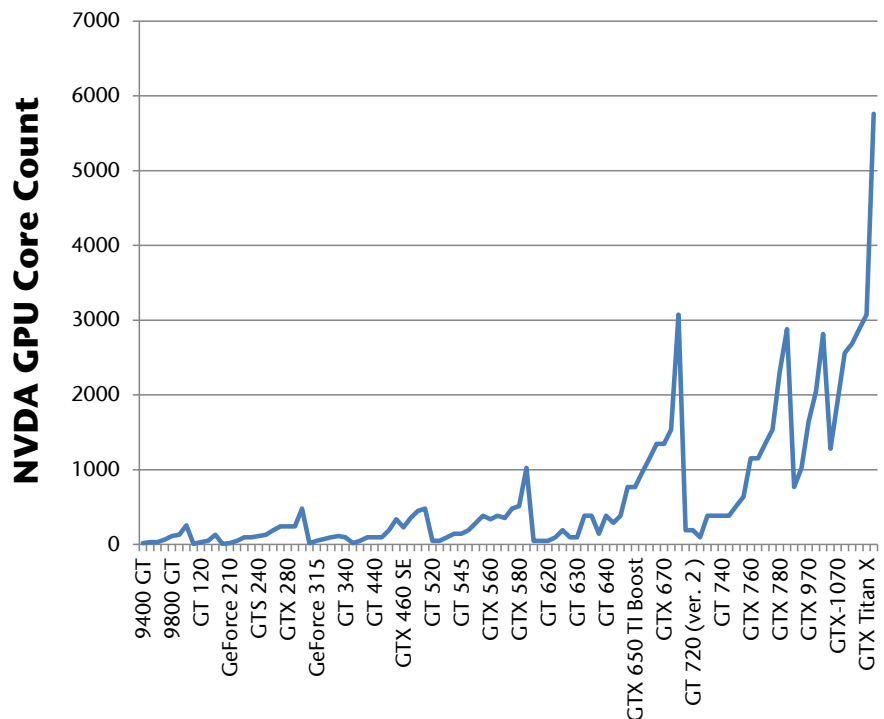


Source: Jefferies. Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten. New plot and data collected for 2010-2015 by K. Rupp.

NVidia has also been increasing its own core count in its GPUs

NVidia's Titan X can have up to 6,000 GPU cores – two full orders of magnitude higher than the highest core count MPUs

Chart 13: NVDA Discrete GPU Core Count (2008-2016)



Source: Jefferies, NVDA

The chart below highlights examples of non-x86 processors being deployed by the Super 7 cloud players, suggesting that the center of gravity of cloud processing is drifting away from x86

Chart 14: "Super 7" Non-x86 Data Center Announcements

Date	" Super 7 "	Product	Processor	Comments
9-Mar-17	Facebook	Big Basin	NVDA GPU	FB Big Basin AI compute adopts NVDA Tesla P100 for data centers
13-Jul-16		Big Sur	NVDA GPU	FB Big Sur uses NVDA M40s for neural network training
15-Nov-16	Google	GCP	AMD GPU	GCP announces AMD FirePro S9300 for remote workstations
15-Nov-16		↓	NVDA GPU	GCP announces NVDA Tesla P100 and K80s for deep learning
8-Mar-17	Microsoft	Azure	CAVM ARM	MSFT announces collaboration with CAVM using its 64-bit ARM processors
4-Aug-16		↓	NVDA GPU	MSFT Azure announces NVDA K80 accelerators for GA on 1-Dec-16
19-Apr-17	Amazon	AWS	XLNX FPGA	AWS announces EC2 F1 instances for GA
29-Sep-16		↓	NVDA GPU	AWS adopts NVDA K80 accelerators in its EC2 PC instances
17-Apr-17	Baidu	Baidu Cloud	NVDA GPU	Baidu integrates NVDA's Pascal-based Tesla GPU accelerators
17-Oct-16		↓	XLNX FPGA	Baidu adopts XLNX in its data centers
3-Dec-16	Alibaba	AliCloud	ARM CPU	BABA to use SoftBank's ARM CPUs in data center
14-Oct-16		↓	AMD GPU	BABA reaches deal with AMD to use Radeon Pro GPUs in data center
20-Jan-16		↓	NVDA GPU	BABA teams with NVDA in \$1b bet on cloud computing
29-Mar-17	Tencent	Tencent Cloud	NVDA GPU	Tencent integrates NVDA's Pascal-based Tesla GPU accelerators

Source: Jefferies

Jefferies Artificial Intelligence Summit Series

The four charts below highlight AI startup companies Jefferies hosted over the past 8 months in Boston, London, Hong Kong and Zurich.

The examples highlight that it is difficult to quantify the market for AI/Neural Networking and Parallel Processing, because with the new technologies, these companies are creating new markets.

Chart 15: Jefferies Artificial Intelligence Summit, Boston 9-Nov-16

Location	Company	Description
Boston	Finn Advisors	Finn.ai is a white label virtual banking assistant, powered by AI (Siri for your bank). Finn helps digitally savvy financial institutions put a 'personal banker' in every customer's pocket, helping them to manage their money wherever they are, whatever they need via a simple, natural conversation.
Boston	Google Research	Research at Google (Google Brain team) is at the forefront of innovation in Machine Intelligence, with active research exploring virtually all aspects of machine learning, including deep learning and more classical algorithms.
Boston	IronNet Cybersecurity	IronNet Cybersecurity (founded in 2014 has raised \$33m in venture funding) is focusing on solutions for the commercial sector to make our networks safe through the implementation of behavioral modeling, statistical analysis, and machine learning techniques.
Boston	MIT Computer Science and AI Laboratory	Tommi Jaakkola, Ph.D., Professor of Electrical Engineering and Computer Science, focuses on machine learning and natural language processing. His academic work on statistical inference and estimation is well known in the field of machine learning.
Boston	Nvidia	Nvidia Corporation is an American technology company based in California. It designs graphics processing units for the gaming and professional markets, and its progress advancing the deep learning industry has resulted in a lead versus its competitors.
Boston	Sentient Technologies	Sentient, founded in 2008, is well known in the tech circles for being the highest funded Artificial Intelligence company in the world. Co-founder Antoine Blondeau, helped invent the technology behind Apple's Siri and has been developing a distributed AI platform across multiple disciplines including marketing, e-commerce and finance markets.

Source: Jefferies

Chart 16: Jefferies Artificial Intelligence Summit, London 23-Jan-17

Location	Company	Description
London	Babylon Health	Babylon Health is the UK's leading digital healthcare service. Its purpose is to democratise healthcare by putting an accessible and affordable health service into the hands of every person on earth by building an AI doctor to diagnose illness via its app. Babylon Health has raised over \$60m in venture funding.
London	Cortexica Vision Systems	Cortexica Vision Systems is the leading provider of image and object recognition systems. Through utilising the latest AI techniques, Cortexica provides unique and personalised solutions to solve various business problems and needs.
London	Deepomatic	Deepomatic helps companies and scientists build massive, high quality, image-based datasets to unlock the new generation of AI. Deepomatic's technology is based on deep learning, a method to turn basic visual inputs into meaningful concepts. The company has designed a visual search engine, capable of handling millions of search requests among millions of products.
London	Featurespace	Featurespace is the world-leader in Adaptive Behavioural Analytics and creator of the ARIC platform, a machine learning software system developed out of the University of Cambridge.
London	Intelligent Voice	Intelligent Voice Limited is a global leader in the development of proactive compliance and eDiscovery technology solutions for voice, video and other media. Intelligent Voice turns your company and your client's calls, email and IM into smart data, using a powerful Machine Learning Engine and the "World's Fastest" Speech to Text Engine.
London	re:infer	re:infer is an artificial intelligence solution that provides structured, human-like understanding of B2C conversations. B2C businesses are talking more and more with their customers -- be that on in-app or on-website chat, third party messaging platforms, social, calls or emails.
London	Sadako Technologies	Sadako Technologies is a Spanish innovative company incorporated in 2012 as a result of the concern of its founders to use technology and engineering to build a better world. Specialists in the development of latest generation technologies in the areas of computer vision and robotics.
London	Stepsize	Stepsize's first product solves a problem developers working in teams face every day: what is this piece of code, who wrote it, and why? Since these questions emerge from the code, the company aggregates metadata from the tools they use daily and provide a context layer available directly from their editor.
London	V.U. Amsterdam	Dr. Eiben is a professor of Computational Intelligence at VU University Amsterdam and Visiting Professor on the University of York, UK. His research lies in the field of Artificial Intelligence, Evolutionary Computing, Evolutionary Robotics, Artificial Life, and Adaptive Collective Systems.

Source: Jefferies

Chart 17: Jefferies Artificial Intelligence Summit, Hong Kong 6-Apr-17

Location	Company	Description
HK	Baidu Big Data Lab	Baidu's Big Data Lab focuses on large-scale machine learning algorithms and applications in areas such as predictive analysis, large data structure algos and intelligent systems research. Haishan Wu is a senior data scientist and leads a team focusing on mining spatial-temporal big data of Baidu.
HK	CloudMinds Technology	CloudMinds is researching avenues for many rapidly advancing AI engines in the cloud to make robots smarter. Bill Huang is the CEO of CloudMinds, where he is implementing a vision to provide households around the world with cloud base intelligent robots.
HK	Hangzhou Hikvision Digital Technology	The image recognition technology developed by Hikvision Robust Reading platform is currently the world's most influential contest in OCR technology field.
HK	HKUST and WeChat Joint Lab on AI	Research areas of WHAT LAB include intelligent robotic systems, natural language processing, data mining, speech recognition and understanding. The lab will bring together top researchers in the development of innovative artificial intelligence application with the database of WeChat.
HK	LeEco	Kai Ni is a Vice President in the Autonomous Driving Division at LeEco. Mr. Ni is responsible for leading the advanced engineering activity of autonomous driving and is in charge of product development of ADAS and automated driving systems.
HK	Megvii Technology / Face++	Megvii is producing China's top cutting-edge vision technology AI platform, Megvii products are now being used by Financial Institutions, real estates, security industries all around China.
HK	SenseTime	Founded in 2014, SenseTime's face recognition technology has an error rate below one in 100,000. It also provides text, vehicle and image recognition to mobile Internet companies, financial services and security companies. SenseTime has raised over \$120 million in venture funding.
HK	UB Tech	Ubtech is China's biggest robot maker, a \$5b technology company based in Shenzhen. The University of Sydney partnered with Ubtech to create the Ubtech AI Research Center which aims to bring researchers from different academic fields together to collaborate on artificial intelligence.

Source: Jefferies

Chart 18: Jefferies Artificial Intelligence Summit, Zurich 13-Apr-17

Location	Company	Description
Zurich	Demiurge Technologies AG	Demiurge Technologies AG is a Switzerland-based artificial intelligence company developing the next generation of deep neural networks and brain chips for mobile robots.
Zurich	Ditto AI	Ditto Labs is a leading provider of vision-as-a-service for enterprises. The award winning company offers enterprise-class machine-learning and computer vision to businesses demanding a cloud-based API to label, filter and search visual media using artificial intelligence.
Zurich	Gamaya	Gamaya fits drones with a hyperspectral imaging system. This type of imaging can give scientists information about the type of ground, vegetation, and even building materials used. Drones are then flown over fields and industrial farmers can accurately measure the health of crops and their environment.
Zurich	Sentifi	Sentifi is building the largest online ecosystem of crowd-experts and influencers in global financial markets (Sentifi Crowd) to generate Sentifi Signals, market intelligence that is not available via traditional news media using our proprietary technology Sentifi Engine.
Zurich	Starmind International AG	Starmind Brain's vision is people to be able to think with the power of 1,000 brains by interlinking millions of people with billions of experiences and making this collective knowledge available in real time anywhere, anytime.
Zurich	Swiss AI Lab IDSIA	Prof. Marco Zaffalon, Swiss AI Lab IDSIA (Dalle Molle Institute for Artificial Intelligence), Founder & Head of the Imprecise Probability Group.
Zurich	University of Zurich	Professor Davide Scaramuzza, is the Director of Robotics & Perception Group at the University of Zurich. Professor Scaramuzza's main research interest is computer vision applied to the autonomous navigation of visually-guided ground and micro flying robots.

Source: Jefferies

The smartphone and PC markets illustrate the importance of delivering a platform in order to capture industry profits. Interestingly, while Nokia and its Symbian/Cellphone platform dominated up until 2008, Apple and its iPhone platform dominated since

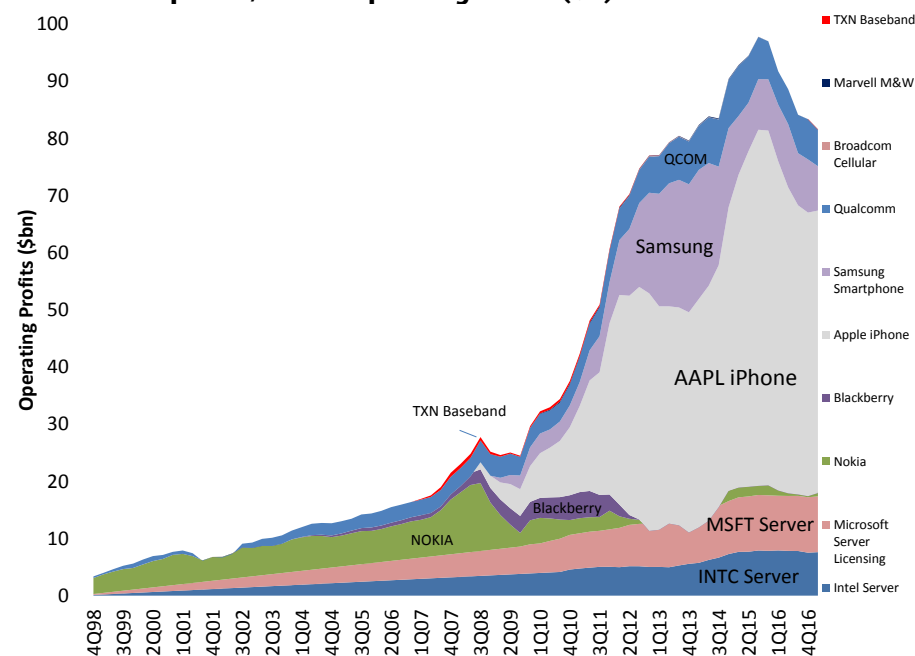
Since an important part of the smartphone value proposition is its ability to harness the computing power of datacenters around the world, we include Intel and Microsoft as proxies for the other half of the Cell Phone / Datacenter computing paradigm

Apple's iPhone emerges as the dominate platform of this computing model today, capturing the majority of profits as it owns all the critical components of the platform: the SoC (Apple's AX SoCs), operating system (iOS), system-level hardware (iPhone) and App Store

On the backend of the Datacenter, Intel and Microsoft account for about 20% of this computing models' industry profits

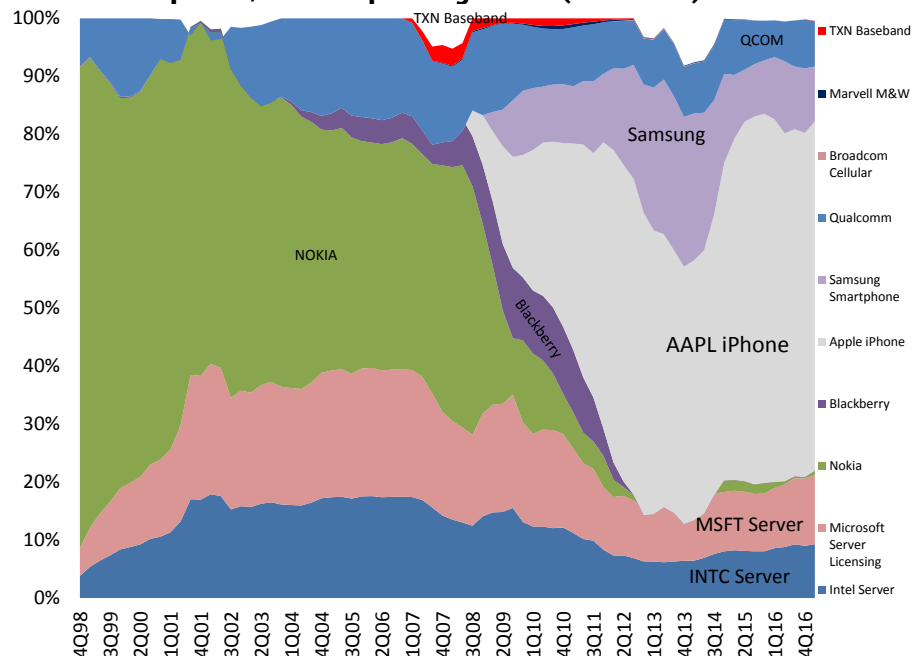
Importance of the Platform – Case Studies

Chart 19: Cellphone / Server Operating Profits (\$Bn)



Source: Jefferies, company data

Chart 20: Cellphone / Server Operating Profits (% of Total)



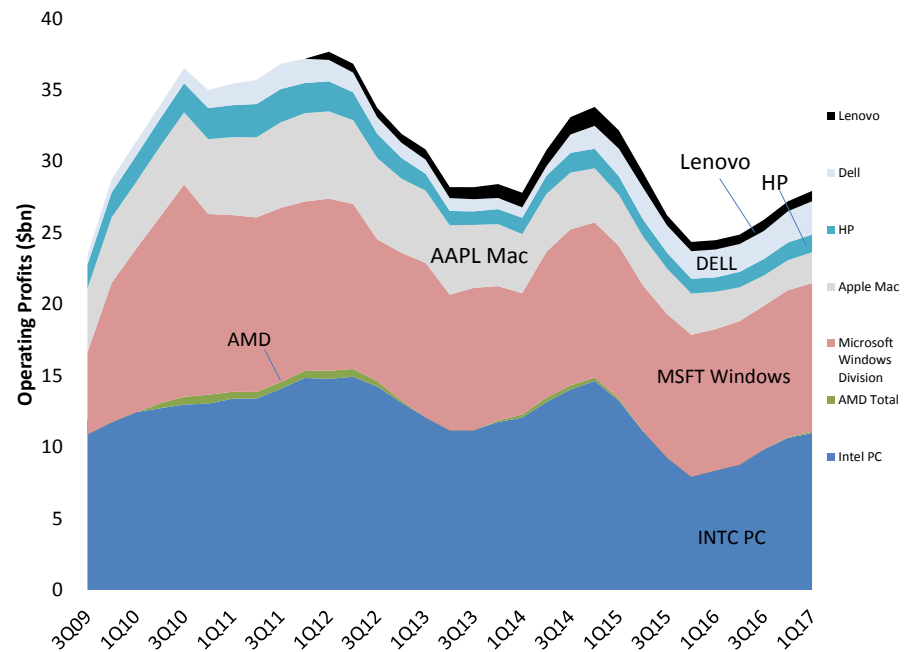
Source: Jefferies, company data

In the PC computing era, IBM anointed Intel and Microsoft as the standards of their PC, and shared its architecture, which enabled the IBM PC clone industry, and exposed component and subsystem makers to global economies of scale

Intel and Microsoft largely define the PC platform (Wintel) and have captured the lion's share of the PC industry profits

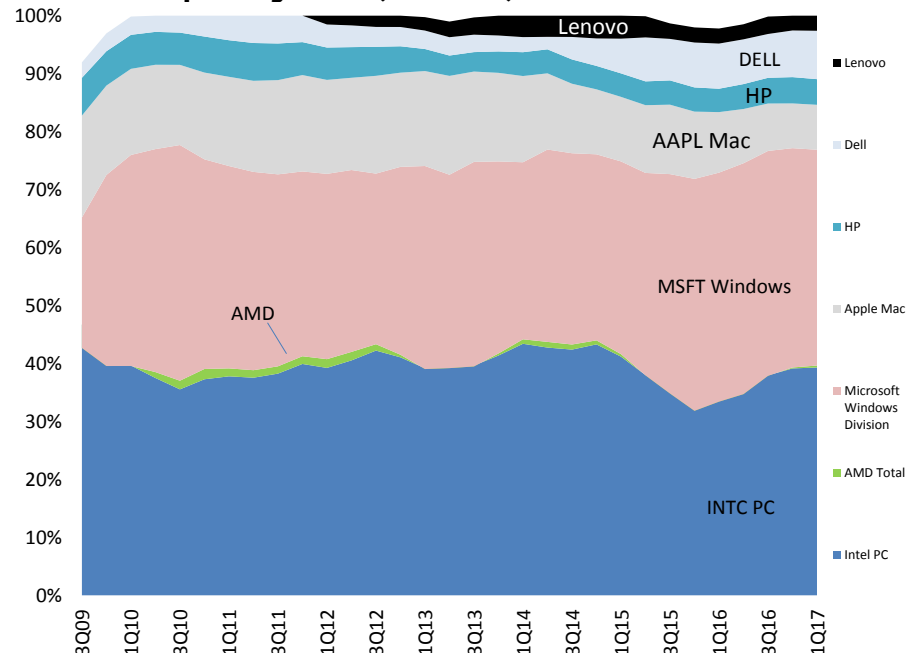
PC OEMs capture the remaining 20% of the industry profits—only Apple, by virtue of developing its own OS (MacOS), capable of capturing an outsized portion relative to its size

Chart 21: PC Operating Profits (\$B)



Source: Jefferies, company data

Chart 22: PC Operating Profits (% of Total)



Source: Jefferies, company data

Parallel Processing Platform Map

In this section we provide a framework for understanding how the different parallel processing platforms (hardware and software) relate to one another and to the high level deep learning frameworks that exist in the market today. For each platform, we highlight the Processor, Native Languages, the higher level Libraries and Primitives, as well as the high level Deep Learning Programming Languages and Frameworks that support that platform

Based on this framework, and our own field checks, we come to the following conclusions:

- 1) NVidia is the best Positioned Platform Provider at the Processor and Software Level. NVidia was early in recognizing the potential for the parallel processing market and 10 years ago started investing in a Software and Hardware Platform to prosecute this market. NVDA's proprietary CUDA software is on its 9th generation, where most other parallel software programs are on their 1st or 2nd. Not only did NVidia invest in CUDA, but it also invested in placing CUDA and its GPUs at engineering and science programs at top universities around the world. At the four Jefferies hosted AI conferences, we heard contacts from MIT, Princeton and VU in the Netherlands say things like "we all use NVDA and CUDA at our university." At the processor level, NVidia launched its Pascal GPU with mixed precision support, and is expected to launch its next gen GPU, Volta, with tensor cores, which we believe means that it has at least a 2-year lead on its processor and at least a 2-to-5 year lead on its CUDA software. We think that being a first mover is key, due to the transaction costs associated with porting software to a new platform.
- 2) Google's TensorFlow and TPU (Tensor Processor Unit) appear to be Second. Google introduced its TPU1 for inferencing last year, and recently announced its TPU2 for training last month. Google has its own deep learning framework, called TensorFlow, which is a high level language for writing code to train Neural Networks, that makes the platform underneath it invisible to developers. In the limit, Google's TPU2 could disrupt NVDA's training, but we think it would be limited to disruption at Google. We believe Google is still purchasing NVDA GPUs, and as of yet, Google has not opened its TPU2 up for public use.
- 3) Intel Has Made Good Efforts in its Software and Processors, but is still lagging NVidia and Google. Intel has three platforms... arguably four: based on Xeon, Xeon PHI, Altera FPGA (Programmable Solutions Group), and Lake Crest (Nervana). On the positive side, Intel has expanded its Math Kernel Library (MKL) to support deep neural network libraries (MKL-DNN), and its Neon Deep Learning Framework has promise. However, its parallel processing solution Xeon PHI does not support mixed precision processing, nor does it have tensor cores for matrix multiplication – key for Deep Learning Training.
- 4) AMD and XLNX have new deep learning products – but yet to see broad adoption. AMD's MiOpen and XLNX SDAccel have been recently introduced to the market. These solutions bear watching, but we have not seen broad support for these solutions in production environments.

A Guide to Understanding Parallel Processing Platforms (Hardware and Software)

Chart 23: Parallel Processing Platform Map

Optimized Deep Learning Frameworks / Design Languages								
Apache Singa		Y					Y	
Caffe		Y		Y	Y	Y	Y	
Caffe2		Y		Under Development				
Deeplearning4j		Y					On Roadmap	
Dlib		Y					No	
Keras		Y					Under Development	
Microsoft Cognitive Toolkit CNTK		Y		Y			Y	
MXNet		Y		Y	Under Development		Y	
Neon				Y		Y		Y
OpenNN		Y					No	
Google TensorFlow		Y	Y	Y	Under Development		Y	
Theano		Y		Y			Y	
Torch		Y		Y			Y	
PLATFORM	Libraries / Primitives	Nvidia Deep Learning SDK (cuDNN...)	Google TPU Software Stack (API, Drivers...)	Intel MKL (MKL-DNN, DAAL...)	Xilinx reVISION, RAS (xfDNN, xfBLAS)		AMD MIOpen (OpenCL, HIP...)	Intel MKL (MKL-DNN, DAAL...)
	Native Languages / Drivers	CUDA	C++, CUDA	C++	OpenCL, C++ Vivado, SDAccel	OpenCL Quartus	ROCm (HCC, HIP, OpenCL, Python)	C++, Python
	Processor	Nvidia GPU	Google TPU	Intel Xeon Phi, Xeon	Xilinx "UltraScale" FPGA	Intel (Altera) "Arria 10" FPGA	AMD GPU	Intel Nervana "Lake Crest"

Source: Jefferies Equity Research

Based on the findings in Chart 23 (above), NVDA is the best positioned platform provider at the processor and software level

At the chip level, NVDA's Volta launch extends its lead versus competitors which we estimate to be ~2 years

At the software enablement level, we see an even wider lead as NVDA has so deeply seeded its CUDA software and libraries amongst researchers at top universities around the world

NVDA's GPU are available at every major cloud service provider (Amazon, Microsoft, Google, IBM) and its hardware is optimized on all of the most popular deep learning frameworks today (including Baidu's PaddlePaddle which is not presented above).

We view this as a major competitive advantage that is not fully appreciated by investors

Deep Learning Platform Summaries

Chart 24: NVDA Deep Learning Platform Summary

Parallel Processing Platform Element	Comments
Deep Learning Frameworks / Design Languages	Compared to other parallel processing platforms, Nvidia has the broadest support amongst the design languages, which is consistent with its first mover advantage and with its parallel processing native language and extensive libraries.
Libraries / Primitives	<p>Nvidia arguably has the most comprehensive and libraries for deep neural network training and parallel processing. The "Nvidia Deep Learning SDK" includes libraries like cuDNN (Neural Net building blocks), TensorRT (inferencing production deployment), DeepStream (inference, transcoding), cuBLAS and cuSPARSE (Linear Algebra), NCCL (collective communication modules).</p> <p>The table above is largely focused on parallel processing as it relates to deep learning, however one could consider a broader view of NVidia's parallel processing capabilities. In Automotive, Nvidia's SDK is called "DriveWorks," which includes libraries for sensor fusion, map localization, vehicle control and ADAS rendering. In gaming, NVidia's SDK is called "GameWorks," which includes libraries like VisualFX (face, hair, wave, shadow movements), PhysX (clothing, explosions) and OptiX (ambient light, acoustics, line of sight). Importantly, all of NVidia SDKs are based on CUDA and can be integrated with cuDNN.</p>
Native Languages	Nvidia's GPU native language is its proprietary CUDA. CUDA is like a proprietary C++ for Nvidia GPUs that was constructed to support parallel processing. Nvidia originally started investing in CUDA over 10 years ago, and the company has been seeding CUDA in engineering and science programs at Universities for the past 10 years - and consequently the company can boast it has 500,000 developers that know CUDA.
Processor	<p>All of Nvidia's GPUs support CUDA, which means that software developed on one class of GPU can be run on a different class. Its GPUs for the datacenter include its Tesla-class, including its most recently launched DGX-1 (aka supercomputer in a box), which are designed for Neural Network Training, and P1 (or P40?) for inferencing. NVidia's gaming family of GPUs include its Titan family, as well as its GeForce. NVDA's Automotive solutions are called Drive-CX (infotainment, digital dash) and Drive-PX (ADAS, autonomous driving). And its mobile solution is called Tegra.</p> <p>Importantly, NVidia's most recently announced GPU architecture, Volta, has "tensor cores" which allow for rapid matrix multiplication calculations critical for neural network training and inferencing.</p>

Source: Jefferies, company data

In our view, Google's Tensor Processing Unit optimized for TensorFlow appears to be second

In our estimation, it would take months to optimize the TensorFlow code for alternate deep learning frameworks. This limits the use cases for researchers and developers who have already committed to another type of framework (e.g. Caffe or Theano)

Google introduced its TPU1 for inferencing last year, and recently announced its TPU2 for training last month. Google has stated that it will offer Volta-based NVDA GPUs in its Cloud Platform, but we view the TPU as a disruptor in the industry

The TPU puts pressure on the other hyperscale cloud providers to develop a similar offering to remain competitive

Chart 25: Google Deep Learning Platform Summary

Parallel Processing Platform Element	Comments
Deep Learning Frameworks / Design Languages	Google's Tensor Processing Unit (TPU) is only compatible with TensorFlow, Google designed open sourced deep learning framework. Google's TPU 2.0 (announced 2Q17) was optimized for TensorFlow so researchers who develop in TensorFlow abstraction can expect better performance, while developers using competing software engines (such as Caffe and Torch) will have to conform to adopt the hardware. Bear in mind TensorFlow is the most widely used deep learning framework.
Libraries / Primitives	Google's engineering team developed a compiler and a proprietary software stack that translates API calls from TensorFlow to TPU instructions. Compared to a CPU that can compute 10's of operation per cycle, a TPU can compute hundreds of thousands a cycle.
Native Languages	C++, Python, CUDA are the native languages. Like a GPU the TPU stack is split into a User Space Driver and a Kernel Driver. The Kernel Driver is lightweight and handles only memory management and interrupts - designed for long-term stability. The User Space driver changes frequently. It sets up and controls TPU execution, reformats data into TPU order, translates API calls into TPU instructions, and turns them into application binary. The User Space driver compiles a model the first time it is evaluated, caching the program image and writing the weight image into the TPU's weight memory; the second and following evaluations run at full speed.
Processor	Google's Tensor Processing Unit (TPU) is a custom-ASIC that was conceived in 2013 as a need for larger computational demands for neural networks were identified within the company leading to a requirement to double the datacenter footprint. In just over a year the chip was 28nm silicon was designed, built and shipped -- unprecedented for the industry. Matrix multiplication operations were targeted within the organization to limit reliance on 16- and 32-bit CPUs and GPUs.

Source: Jefferies, company data

In our view, INTC has made good efforts in its software enablement platforms as well as at the processor level, but it still lags NVDA and Google in the deep learning field

INTC has a myriad of product offerings (Xeon, Xeon Phi, Arria 10 FPGAs, and Nervana silicon)

We view the software development with its Math Kernel Libraries (DAAL, MKL-DNN, etc.) as a differentiator – but it still lags NVDA's robust software suite and libraries

Chart 26: INTC Deep Learning Platform Summary

Parallel Processing Platform Element	Comments
Deep Learning Frameworks / Design Languages	Compared to other parallel processing platforms, Intel has been investing more nominal dollars in the development of its ecosystem simply because of its size. In our view, Intel's efforts in the field lack the expertise of Nvidia and the impact of Google.
Libraries / Primitives	<p>Intel Math Kernel Library (MKL) attempt to abstract the reliance on any particular hardware (whether that be a CPU, GPU, ASIC, DSP or FPGA). Intel originally released its MKL over a decade ago, but a more stable release (v 11.3) was updated in February 2016. MKL software supports code to optimize mathematical routines specifically for Intel processors (Xeon, Xeon Phi, and Nervana silicon) to increase application performance. Highly vectorize and threaded functions such as linear algebra (BLAS), fast fourier transforms, deep neural networks, vector math, statistics and data fitting optimizations are released for free to customers through a community licensing program.</p> <p>Intel Math Kernel Library - Deep Neural Networks (MKL-DNN) is an open source primitive that is included in Intel's Math Kernel Library to optimize performance of a deep neural net. MKL-DNN is meant to accelerate technologies like image recognition and allows use with Python and Java.</p> <p>Within Intel MKL is the Intel Data Analytics Acceleration Library (DAAL) which helps speed big data analytics by providing optimized algorithmic building blocks for all data analysis stages (pre-processing, transformation, analysis, modeling, validation and decision making). Intel DAAL is optimized to support Intel's Xeon, Xeon Phi, Core and Atom processors.</p>
Native Languages	Initial release of Intel Math Kernel Libraries was August 2015 and supports both C++ and Java interfaces.
Processor	Intel processors targeting the deep learning industry include Xeon, Core and Atom; while accelerators include Xeon Phi, Altera FPGAs, and Nervana. While Intel's approach to tackle the deep learning market does not appear to be as unified as Nvidia's, we note that the widespread adoption of Intel silicon does work to its advantage in the space.

Source: Jefferies, company data

XLNX SDAccel has been recently introduced to the market

This solution bears watching, but we have not seen broad support for this solution in production environments

Chart 27: XLNX Deep Learning Platform Summary

Parallel Processing Platform Element	Comments
Deep Learning Frameworks / Design Languages	While Xilinx's main FPGA competitor at the leading edge (Altera) was acquired by Intel in 2015, the competition in the FPGA accelerator market is intense. This market represents less than 5% of total sales today, but is growing quickly as hyperscale cloud providers like AWS, Azure and Baidu Cloud adopt FPGAs-AAS in the datacenter.
Libraries / Primitives	<p>Xilinx's reVISION software stack for machine learning and computer vision applications at the edge sit on top of Xilinx's Zynq MPSoCs. In the datacenter Xilinx's Reconfigurable Acceleration Stack sit on top of UltraScale+ FPGAs to optimize performance. Xilinx proprietary libraries (xvDNN, xvBLAS, etc.) are highly optimized for Xilinx FPGAs at 16-bit and 8-bit integer data types.</p> <p>Given that Xilinx's SDAccel environment sits on top of OpenCL, we think the pole position in the accelerator market rests with Nvidia largely due to its robust libraries for deep neural network training and parallel processing.</p>
Native Languages	<p>Xilinx SDAccel development environment for OpenCL and C/C++ enables optimized performance of Xilinx's FPGAs in data center acceleration applications. SDAccel combines the optimized compiler supporting OpenCL, C and C++ kernels, along with proprietary libraries, and development boards to streamline run-time performance on an FPGA.</p> <p>Xilinx SDAccel was first released in 2015 and while the initiative is being invested in currently, the company notes that development is significantly behind Nvidia with its proprietary CUDA software.</p> <p>Xilinx's Vivado design suite allows users to leverage C-based design calls.</p>
Processor	<p>Xilinx FPGAs for training and inference (machine learning, computer vision, etc.) are based on the reconfigurable nature, lending itself to a wide range of workloads and new evolving algorithms. Microsoft Azure (using Altera FPGAs) and Amazon AWS (using Xilinx FPGAs) are the two most notable hyperscale cloud vendors to deploy FPGAs for commercial availability. Baidu has also adopted Xilinx FPGAs in its datacenter. FPGA instances in the datacenter are useful in many HPC applications such as genomics research, financial analytics, real-time video processing, big data search, and security.</p> <p>One of the major drawbacks to deploying FPGAs at massive scale is the difficulty to program, but increased levels of software abstraction (such as SDAccel) intend to alleviate this problem.</p>

Source: Jefferies, company data

AMD's miOpen has been recently introduced to the market

This solution bears watching, but we have not seen broad support for this solution in production environments

Chart 28: AMD Deep Learning Platform Summary

Parallel Processing Platform Element	Comments
Deep Learning Frameworks / Design Languages	While AMD's main competitor in the Deep Learning market has the broadest support amongst design languages, AMD has recognized the importance of a GPU as a parallel processor and believes its competitive advantage is its open source strategy.
Libraries / Primitives	<p>MiOpen is AMD's open source library to enable Radeon-GPU acceleration for deep learning and was made available to developers in early July 2017. Analogous to NVDA's cuDNN, miOpen provides GPU-tuned implementations for standard calls into programming languages OpenCL and Python. In our view, AMD's miOpen deep learning library is an important step to optimize performance on Radeon GPU-specific hardware.</p> <p>MiOpen supports 2 programming models - OpenCL and HIP (Heterogeneous-Compute Interface for Portability). HIP is a CUDA porting tool that scans CUDA source code and converts it to source code that can run on AMD GPUs.</p> <p>AMD is targeting a broad range of applications with its parallel processing capabilities including autonomous vehicles, smart homes, drones, etc.</p>
Native Languages	<p>AMD announced its Radeon Open Compute Platform (ROCm) v 1.0 in April 2016 in attempt to provide an alternative to CUDA, NVDA's proprietary GPU language. Various compilers and porting tools within ROCm allow for the possibility to use of alternative languages on the platform (CUDA, C/C++, Lua, Python, R, Ruby, etc.).</p> <p>Theoretically this porting of code from a higher level abstraction layer to the AMD silicon provides a migration path for developers, but we believe performance will inevitably be sacrificed.</p>
Processor	AMD has a unique position in the deep learning stack, as it is the only company with x86 and discrete GPU silicon expertise and has been promoting the concept of a Heterogeneous System Architecture (HSA). The HSA allows the CPU and GPU code to be written in the same source file and support capabilities such as a unified CPU-GPU memory space. AMD's Vega GPU Architecture (expected to be released in 3Q17) is targeting the high end of the discrete GPU market, a market that NVDA has dominated. While we view the product rollout as a positive for AMD, we will have to see what features are unveiled with deep learning in mind, to remain competitive with NVDA's Volta architecture.

Source: Jefferies, company data

Parallel Processing / IoT Winners and Losers

Chart 29: Tectonic Shift Winners & Losers

Company	Winner/Loser	Comments
AMD	Winner	<ul style="list-style-type: none"> We see AMD as a beneficiary of the shift in computing as it is the only other company besides NVDA with GPU technology, and this will be increasingly valued as workloads migrate to parallel, multi-core architectures
CAVM	Winner	<ul style="list-style-type: none"> We see CAVM as a beneficiary of the shift toward a parallel computing model due to its heritage in networking and multi-core processing CAVM is the only commercially viable 64-bit ARM server vendor, and we think it gains share in emerging workloads due to its lower price/performance/watt We think MSFT's adoption of ARM in datacenter catalyzes broader adoption
INTC	Loser	<ul style="list-style-type: none"> We see INTC as structurally disadvantaged in the shift in computing due to its reliance on x86 architecture, and dominant market position x86 single-core performance per dollar has plateaued, while multi-core performance per dollar has increased We see emerging workloads migrating to parallel, multi-core architectures based on NVDA's GPUs and CAVM's ThunderX
NVDA	Winner	<ul style="list-style-type: none"> As the de facto standard in AI, we see NVDA as a primary beneficiary of the shift in computing to parallel, multi-core workloads NVDA dominates the GPU market and has invested heavily in the ecosystem needed to enable developers to use them We see NVDA capitalizing on its dominate position as machine learning workloads grow
XLNX	Winner	<ul style="list-style-type: none"> We see XLNX as a beneficiary of the shift to a parallel computing model as its programmable devices are capable of being parallelized Like NVDA, XLNX has invested heavily in the software ecosystem and expanded its user base 5x by creating development environments (SDAccel, SDSoC, SDNet), and supporting machine learning in embedded vision (reVISION) As the last independent FPGA vendor we see it capitalizing on the tailwind from share gains on the latest leading-edge nodes—65% on 28nm; 80% on 20nm; and 100% on 16nm

Source: Jefferies

Analyst Certification:

I, Mark Lipacis, certify that all of the views expressed in this research report accurately reflect my personal views about the subject security(ies) and subject company(ies). I also certify that no part of my compensation was, is, or will be, directly or indirectly, related to the specific recommendations or views expressed in this research report.

I, Delos Elder, CFA, CPA, certify that all of the views expressed in this research report accurately reflect my personal views about the subject security(ies) and subject company(ies). I also certify that no part of my compensation was, is, or will be, directly or indirectly, related to the specific recommendations or views expressed in this research report.

I, Sean Dorsey, CPA, certify that all of the views expressed in this research report accurately reflect my personal views about the subject security(ies) and subject company(ies). I also certify that no part of my compensation was, is, or will be, directly or indirectly, related to the specific recommendations or views expressed in this research report.

As is the case with all Jefferies employees, the analyst(s) responsible for the coverage of the financial instruments discussed in this report receives compensation based in part on the overall performance of the firm, including investment banking income. We seek to update our research as appropriate, but various regulations may prevent us from doing so. Aside from certain industry reports published on a periodic basis, the large majority of reports are published at irregular intervals as appropriate in the analyst's judgement.

Investment Recommendation Record

(Article 3(1)e and Article 7 of MAR)

Recommendation Published, 00:31 ET, July 10, 2017
Recommendation Distributed, 01:00 ET, July 10, 2017

Company Specific Disclosures

For Important Disclosure information on companies recommended in this report, please visit our website at <https://javatar.bluematrix.com/sellside/Disclosures.action> or call 212.284.2300.

Explanation of Jefferies Ratings

Buy - Describes securities that we expect to provide a total return (price appreciation plus yield) of 15% or more within a 12-month period.

Hold - Describes securities that we expect to provide a total return (price appreciation plus yield) of plus 15% or minus 10% within a 12-month period.

Underperform - Describes securities that we expect to provide a total return (price appreciation plus yield) of minus 10% or less within a 12-month period.

The expected total return (price appreciation plus yield) for Buy rated securities with an average security price consistently below \$10 is 20% or more within a 12-month period as these companies are typically more volatile than the overall stock market. For Hold rated securities with an average security price consistently below \$10, the expected total return (price appreciation plus yield) is plus or minus 20% within a 12-month period. For Underperform rated securities with an average security price consistently below \$10, the expected total return (price appreciation plus yield) is minus 20% or less within a 12-month period.

NR - The investment rating and price target have been temporarily suspended. Such suspensions are in compliance with applicable regulations and/or Jefferies policies.

CS - Coverage Suspended. Jefferies has suspended coverage of this company.

NC - Not covered. Jefferies does not cover this company.

Restricted - Describes issuers where, in conjunction with Jefferies engagement in certain transactions, company policy or applicable securities regulations prohibit certain types of communications, including investment recommendations.

Monitor - Describes securities whose company fundamentals and financials are being monitored, and for which no financial projections or opinions on the investment merits of the company are provided.

Valuation Methodology

Jefferies' methodology for assigning ratings may include the following: market capitalization, maturity, growth/value, volatility and expected total return over the next 12 months. The price targets are based on several methodologies, which may include, but are not restricted to, analyses of market risk, growth rate, revenue stream, discounted cash flow (DCF), EBITDA, EPS, cash flow (CF), free cash flow (FCF), EV/EBITDA, P/E, PE/growth, P/CF, P/FCF, premium (discount)/average group EV/EBITDA, premium (discount)/average group P/E, sum of the parts, net asset value, dividend returns, and return on equity (ROE) over the next 12 months.

Jefferies Franchise Picks

Jefferies Franchise Picks include stock selections from among the best stock ideas from our equity analysts over a 12 month period. Stock selection is based on fundamental analysis and may take into account other factors such as analyst conviction, differentiated analysis, a favorable risk/reward ratio and investment themes that Jefferies analysts are recommending. Jefferies Franchise Picks will include only Buy rated stocks and the number can vary depending on analyst recommendations for inclusion. Stocks will be added as new opportunities arise and removed when the reason for inclusion changes, the stock has met its desired return, if it is no longer rated Buy and/or if it triggers a stop loss. Stocks having 120 day volatility in the bottom quartile of S&P stocks will continue to have a 15% stop loss, and the remainder will have a 20% stop. Franchise Picks are not intended to represent a recommended portfolio of stocks and is not sector based, but we may note where we believe a Pick falls within an investment style such as growth or value.

Risks which may impede the achievement of our Price Target

This report was prepared for general circulation and does not provide investment recommendations specific to individual investors. As such, the financial instruments discussed in this report may not be suitable for all investors and investors must make their own investment decisions based upon their specific investment objectives and financial situation utilizing their own financial advisors as they deem necessary. Past performance of the financial instruments recommended in this report should not be taken as an indication or guarantee of future results. The price, value of, and income from, any of the financial instruments mentioned in this report can rise as well as fall and may be affected by changes in economic, financial and political factors. If a financial instrument is denominated in a currency other than the investor's home currency, a change in exchange rates may adversely affect the price of, value of, or income derived from the financial instrument described in this report. In addition, investors in securities such as ADRs, whose values are affected by the currency of the underlying security, effectively assume currency risk.

Other Companies Mentioned in This Report

- Advanced Micro Devices, Inc. (AMD: \$13.88, BUY)
- Analog Devices, Inc. (ADI: \$78.96, BUY)
- Broadcom (AVGO: \$253.11, BUY)
- Cavium Inc. (CAVM: \$66.57, BUY)
- Cisco Systems, Inc. (CSCO: \$31.84, BUY)
- Inphi Corporation (IPHI: \$38.99, BUY)
- Intel Corporation (INTC: \$34.73, UNDERPERFORM)
- Lattice Semiconductor Corporation (LSCC: \$6.72, HOLD)
- M/A-COM Technology Solutions Holdings, Inc. (MTSI: \$63.02, BUY)
- Marvell Technology Group Ltd. (MRVL: \$16.25, HOLD)
- Maxim Integrated Products, Inc. (MXIM: \$45.10, BUY)
- Mellanox Technologies, Ltd. (MLNX: \$44.50, UNDERPERFORM)
- Microchip Technology Inc. (MCHP: \$80.88, BUY)
- NVIDIA Corporation (NVDA: \$168.10, BUY)
- NXP Semiconductors NV (NXPI: \$109.86, HOLD)
- ON Semiconductor Corporation (ON: \$15.20, HOLD)
- Orbotech Ltd. (ORBK: \$36.98, BUY)
- QUALCOMM Incorporated (QCOM: \$53.84, BUY)
- Rambus, Inc. (RMBS: \$12.61, HOLD)
- Texas Instruments Incorporated (TXN: \$81.70, BUY)
- Xilinx Corp (XLNX: \$64.88, BUY)

For Important Disclosure information on companies recommended in this report, please visit our website at <https://javatar.bluematrix.com/sellside/Dislosures.action> or call 212.284.2300.

Distribution of Ratings

Rating	Count	Percent	IB Serv./Past 12 Mos.	
			Count	Percent
BUY	1086	50.87%	335	30.85%
HOLD	897	42.01%	180	20.07%
UNDERPERFORM	152	7.12%	15	9.87%

Other Important Disclosures

Jefferies does and seeks to do business with companies covered in its research reports. As a result, investors should be aware that Jefferies may have a conflict of interest that could affect the objectivity of this report. Investors should consider this report as only a single factor in making their investment decision.

Jefferies Equity Research refers to research reports produced by analysts employed by one of the following Jefferies Group LLC ("Jefferies") group companies:

United States: Jefferies LLC which is an SEC registered firm and a member of FINRA.

United Kingdom: Jefferies International Limited, which is authorized and regulated by the Financial Conduct Authority; registered in England and Wales No. 1978621; registered office: Vintners Place, 68 Upper Thames Street, London EC4V 3BJ; telephone +44 (0)20 7029 8000; facsimile +44 (0)20 7029 8010.

Hong Kong: Jefferies Hong Kong Limited, which is licensed by the Securities and Futures Commission of Hong Kong with CE number ATSS46; located at Suite 2201, 22nd Floor, Cheung Kong Center, 2 Queen's Road Central, Hong Kong.

Singapore: Jefferies Singapore Limited, which is licensed by the Monetary Authority of Singapore; located at 80 Raffles Place #15-20, UOB Plaza 2, Singapore 048624, telephone: +65 6551 3950.

Japan: Jefferies (Japan) Limited, Tokyo Branch, which is a securities company registered by the Financial Services Agency of Japan and is a member of the Japan Securities Dealers Association; located at Hibiya Marine Bldg, 3F, 1-5-1 Yuraku-cho, Chiyoda-ku, Tokyo 100-0006; telephone +813 5251 6100; facsimile +813 5251 6101.

India: Jefferies India Private Limited (CIN - U74140MH2007PTC200509), which is licensed by the Securities and Exchange Board of India as a Merchant Banker (INM000011443), Research Analyst (INH000000701) and a Stock Broker with Bombay Stock Exchange Limited (INB011491033) and National Stock Exchange of India Limited (INB231491037) in the Capital Market Segment; located at 42/43, 2 North Avenue, Maker Maxity, Bandra-Kurla Complex, Bandra (East) Mumbai 400 051, India; Tel +91 22 4356 6000.

This material has been prepared by Jefferies employing appropriate expertise, and in the belief that it is fair and not misleading. The information set forth herein was obtained from sources believed to be reliable, but has not been independently verified by Jefferies. Therefore, except for any obligation under applicable rules we do not guarantee its accuracy. Additional and supporting information is available upon request. Unless prohibited by the provisions of Regulation S of the U.S. Securities Act of 1933, this material is distributed in the United States ("US"), by Jefferies LLC, a US-registered broker-dealer, which accepts responsibility for its contents in accordance with the provisions of Rule 15a-6, under the US Securities Exchange Act of 1934. Transactions by or on behalf of any US person may only be effected through Jefferies LLC. In the United Kingdom and European Economic Area this report is issued and/or approved for distribution by Jefferies International Limited and is intended for use only by persons who have, or have been assessed as having, suitable professional experience and expertise, or by persons to whom it can be otherwise lawfully distributed. Jefferies International Limited Equity Research personnel are separated from other business groups and are not under their supervision or control. Jefferies International Limited has implemented policies to (i) address conflicts of interest related to the preparation, content and distribution of research reports, public appearances, and interactions between research analysts and those outside of the research department; (ii) ensure that research analysts are insulated from the review, pressure, or oversight by persons engaged in investment banking services activities or other persons who might be biased in their judgment or supervision; and (iii) promote objective and reliable research that reflects the truly held opinions of research analysts and prevents the use of research reports or research analysts to manipulate or condition the market or improperly favor the interests of the Jefferies International Limited or a current or prospective customer or class of customers. Jefferies International Limited may allow its analysts to undertake private consultancy work. Jefferies International Limited's conflicts management policy sets out the arrangements Jefferies International Limited employs to manage any potential conflicts of interest that may arise as a result of such consultancy work. Jefferies International Ltd, its affiliates or subsidiaries, may make a market or provide liquidity in the financial instruments referred to in this investment recommendation. For Canadian investors, this material is intended for use only by professional or institutional investors. None of the investments or investment services mentioned or described herein is available to other persons or to anyone in Canada who is not a "Designated Institution" as defined by the Securities Act (Ontario). In Singapore, Jefferies Singapore Limited is regulated by the Monetary Authority of Singapore. For investors in the Republic of Singapore, this material is provided by Jefferies Singapore Limited pursuant to Regulation 32C of the Financial Advisers Regulations. The material contained in this document is intended solely for accredited, expert or institutional investors, as defined under the Securities and Futures Act (Cap. 289 of Singapore). If there are any matters arising from, or in connection with this material, please contact Jefferies Singapore Limited, located at 80 Raffles Place #15-20, UOB Plaza 2, Singapore 048624, telephone: +65 6551 3950. In Japan this material is issued and distributed by Jefferies (Japan) Limited to institutional investors only. In Hong Kong, this report is issued and approved by Jefferies Hong Kong Limited and is intended for use only by professional investors as defined in the Hong Kong Securities and Futures Ordinance and its subsidiary legislation. In the Republic of China (Taiwan), this report should not be distributed. The research in relation to this report is conducted outside the PRC. This report does not constitute an offer to sell or the solicitation of an offer to buy any securities in the PRC. PRC investors shall have the relevant qualifications to invest in such securities and shall be responsible for obtaining all relevant approvals, licenses, verifications and/or registrations from the relevant governmental authorities themselves. In India this report is made available by Jefferies India Private Limited. In Australia this information is issued solely by Jefferies International Limited and is directed solely at wholesale clients within the meaning of the Corporations Act 2001 of Australia (the "Act") in connection with their consideration of any investment or investment service that is the subject of this document. Any offer or issue that is the subject of this document does not require, and this document is not, a disclosure document or product disclosure statement within the meaning of the Act. Jefferies International Limited is authorised and regulated by the Financial Conduct Authority under the laws of the United Kingdom, which differ from Australian laws. Jefferies International Limited has obtained relief under Australian Securities and Investments Commission Class Order 03/1099, which conditionally exempts it from holding an Australian financial services licence under the Act in respect of the provision of certain financial services to wholesale clients. Recipients of this document in any other jurisdictions should inform themselves about and observe any applicable legal requirements in relation to the receipt of this document.

This report is not an offer or solicitation of an offer to buy or sell any security or derivative instrument, or to make any investment. Any opinion or estimate constitutes the preparer's best judgment as of the date of preparation, and is subject to change without notice. Jefferies assumes no obligation to maintain or update this report based on subsequent information and events. Jefferies, its associates or affiliates, and its respective officers, directors, and employees may have long or short positions in, or may buy or sell any of the securities, derivative instruments or other investments mentioned or described herein, either as agent or as principal for their own account. Upon request Jefferies may provide specialized research products or services to certain customers focusing on the prospects for individual covered stocks as compared to other covered stocks over varying time horizons or under differing market conditions. While the views expressed in these situations may not always be directionally consistent with the long-term views

expressed in the analyst's published research, the analyst has a reasonable basis and any inconsistencies can be reasonably explained. This material does not constitute a personal recommendation or take into account the particular investment objectives, financial situations, or needs of individual clients. Clients should consider whether any advice or recommendation in this report is suitable for their particular circumstances and, if appropriate, seek professional advice, including tax advice. The price and value of the investments referred to herein and the income from them may fluctuate. Past performance is not a guide to future performance, future returns are not guaranteed, and a loss of original capital may occur. Fluctuations in exchange rates could have adverse effects on the value or price of, or income derived from, certain investments. This report has been prepared independently of any issuer of securities mentioned herein and not in connection with any proposed offering of securities or as agent of any issuer of securities. None of Jefferies, any of its affiliates or its research analysts has any authority whatsoever to make any representations or warranty on behalf of the issuer(s). Jefferies policy prohibits research personnel from disclosing a recommendation, investment rating, or investment thesis for review by an issuer prior to the publication of a research report containing such rating, recommendation or investment thesis. Any comments or statements made herein are those of the author(s) and may differ from the views of Jefferies.

This report may contain information obtained from third parties, including ratings from credit ratings agencies such as Standard & Poor's. Reproduction and distribution of third party content in any form is prohibited except with the prior written permission of the related third party. Third party content providers do not guarantee the accuracy, completeness, timeliness or availability of any information, including ratings, and are not responsible for any errors or omissions (negligent or otherwise), regardless of the cause, or for the results obtained from the use of such content. Third party content providers give no express or implied warranties, including, but not limited to, any warranties of merchantability or fitness for a particular purpose or use. Third party content providers shall not be liable for any direct, indirect, incidental, exemplary, compensatory, punitive, special or consequential damages, costs, expenses, legal fees, or losses (including lost income or profits and opportunity costs) in connection with any use of their content, including ratings. Credit ratings are statements of opinions and are not statements of fact or recommendations to purchase, hold or sell securities. They do not address the suitability of securities or the suitability of securities for investment purposes, and should not be relied on as investment advice.

Jefferies research reports are disseminated and available primarily electronically, and, in some cases, in printed form. Electronic research is simultaneously available to all clients. Additional research products including models are available on Jefferies Global Markets Portal. This report or any portion hereof may not be reprinted, sold or redistributed without the written consent of Jefferies. Neither Jefferies nor any officer nor employee of Jefferies accepts any liability whatsoever for any direct, indirect or consequential damages or losses arising from any use of this report or its contents.

For Important Disclosure information, please visit our website at <https://javatar.bluematrix.com/sellside/Disclosures.action> or call 1.888.JEfferies

© 2017 Jefferies Group LLC