HPC + AI WALL STREET



Sr. Distinguished Technologist and Head of Technology Strategy & Evaluation, Office of the CTO



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At HPE, we believe in changing the way people live and work.



The World Before November 2022...

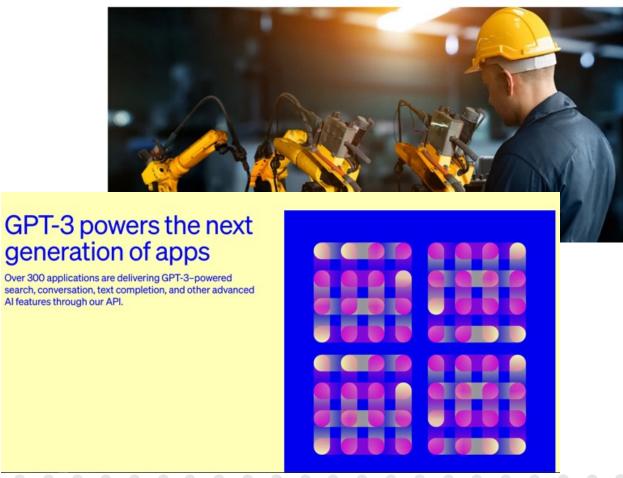


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Using DeepSpeed and Megatron to Train Megatron-Turing NLG 530B, the World's Largest and Most Powerful Generative Language Model



Top 9 Use Cases of Computer Vision in Manufacturing





AMDL

Oct 11, 2021

...and Then This Happened...

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| | | | WALL STREET |
|---|---|----------|-------------|
| OpenAl's chatGPT | The Al Arms Race | | |
| OpenAl's GPT-4 | Siddharth Sharma - Follow 9 min read - Feb 9 | | |
| Google's Bard for Cha | © 2 Q L [‡] ⊙ Ů | t can be | |
| mistak inappro Anthropic's Claude 2 | "It's a new day in search. The race starts today We're going to move fast." — Satya Nadella, Microsoft CEO | | |
| Try Chat(| | | |
| Google's Gemini | | | |
| AWS's Q | | | |
| many more | | | |
| | | | |
| | Altman and Lecun (Forbes) | | |
| | | | |

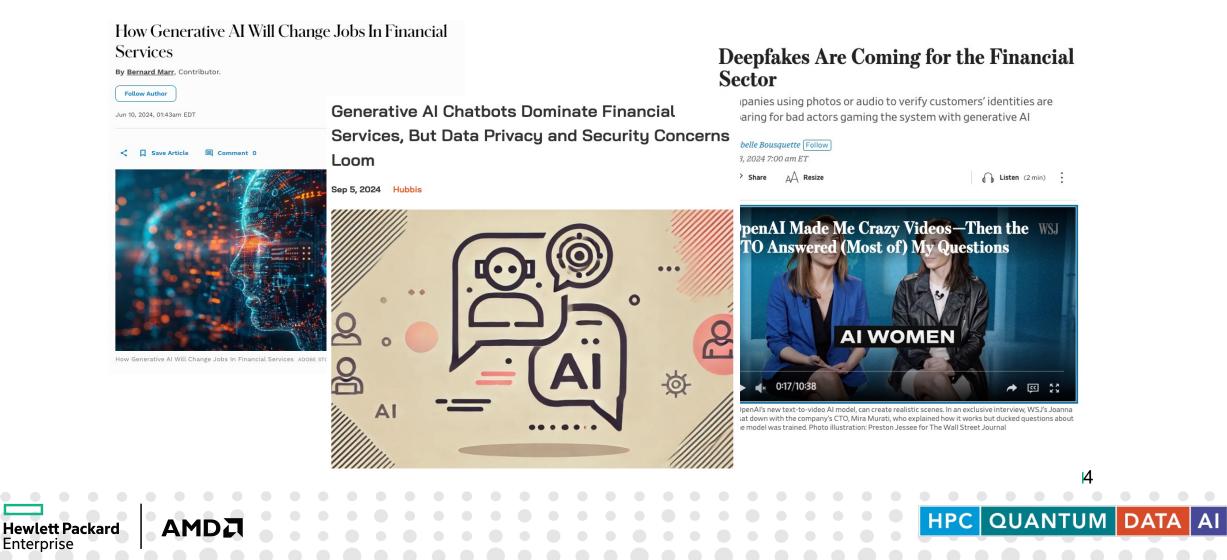
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In recent years, the field of Artificial Intelligence has seen a rapid rise in the development of large language models. These models, based on deep

...and Everything Changed

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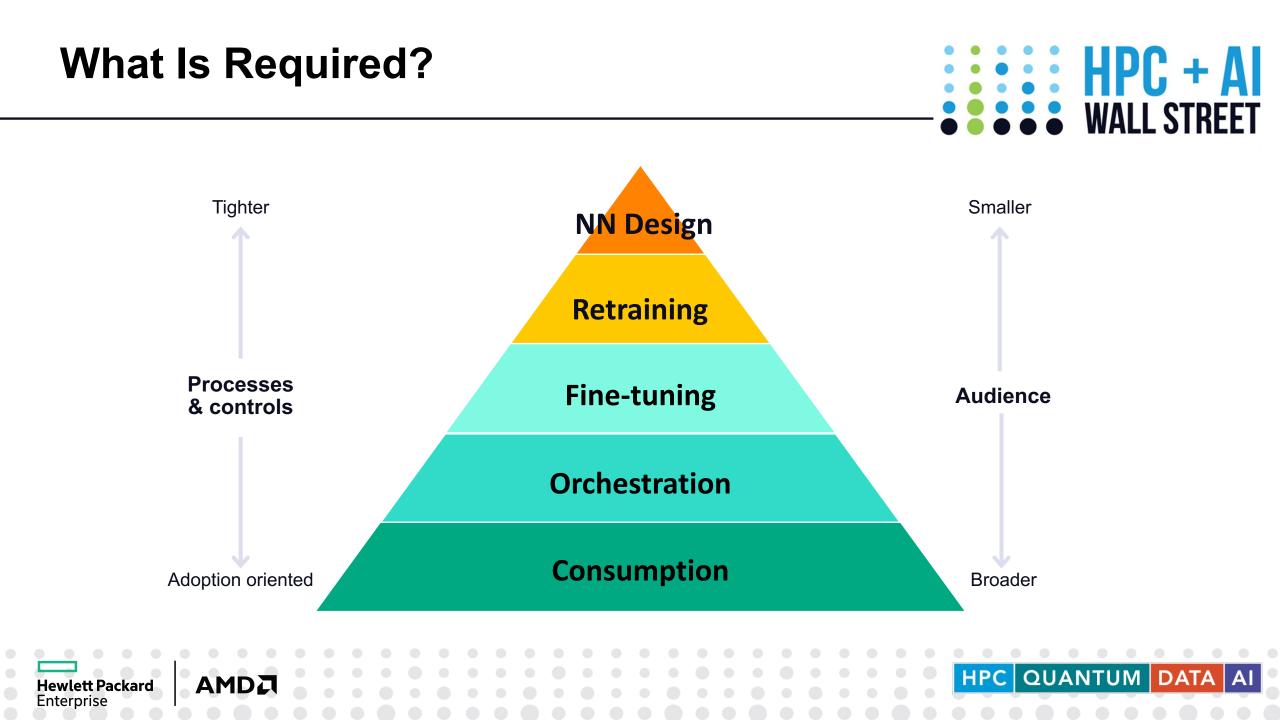
Where Do You Go From Here?

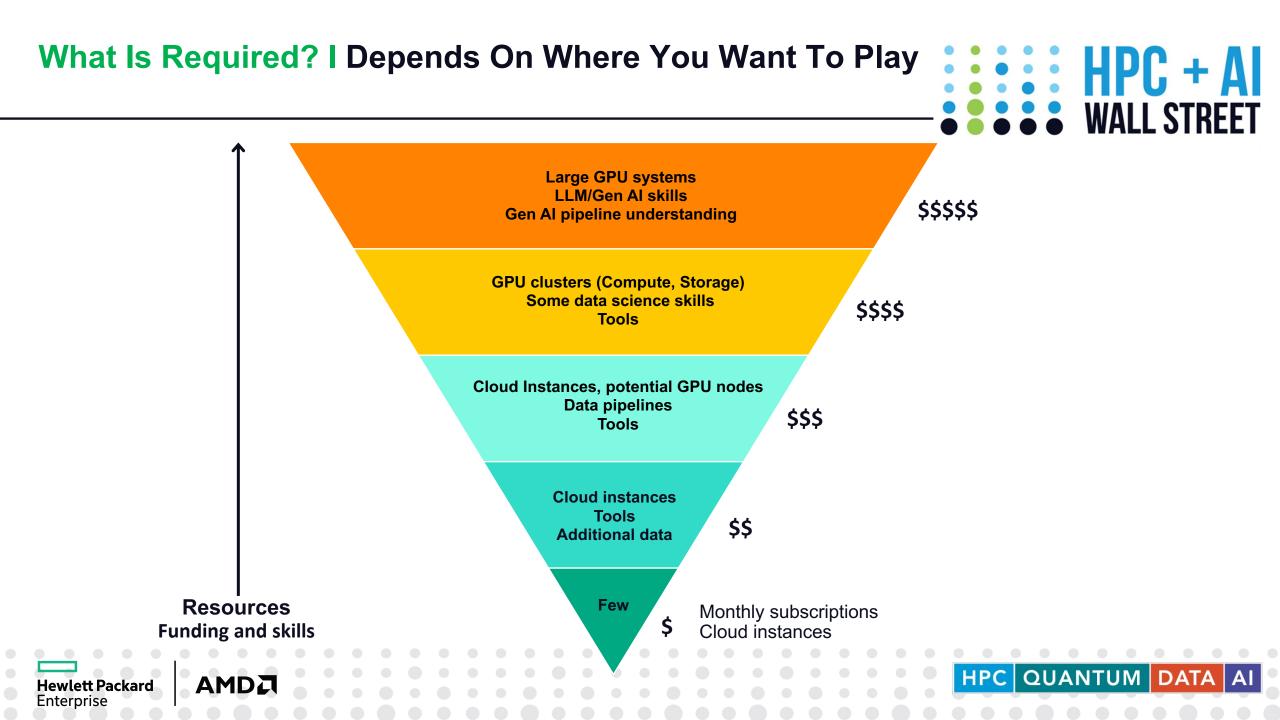




- 1 What is needed to get started ?
- 2 Which **one** (company, model, open source)?
- 3 Is it enterprise ready?
- **4** What about **governance**?
- 5 Which use cases?



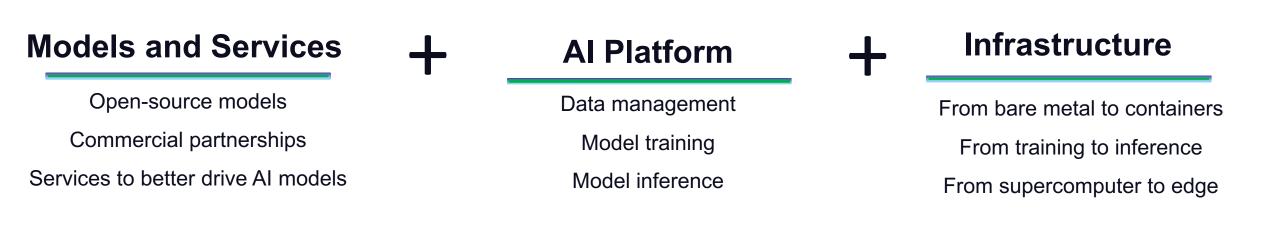




Artificial Intelligence (Gen Al and Beyond)



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One platform | Vendor neutral | Cloud neutral | Al accessible for all



| What Are Your Key Considerations? | HPC + AI WALL STREET | | | | | |
|-----------------------------------|--------------------------------|--|--|--|--|--|
| Models and Services | | | | | | |
| Al Platform | | | | | | |
| Data Services | | | | | | |
| Infrastructure Software | | | | | | |
| Hardware Infrastructure | | | | | | |
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Infrastructure | Good models require good data and also a lot of compute

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If this cat were elected president, its first order of business would be to...



| Completion | Base—64 GPUs for 37 days—56,832 GPU hours |
|------------------|---|
| make sure the ea | conomy is strong. |

 Completion
 Extended—256 GPUs for 22 days—135,168 GPU hours

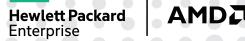
 declare war on the dog.



Infrastructure | Where to Deploy?

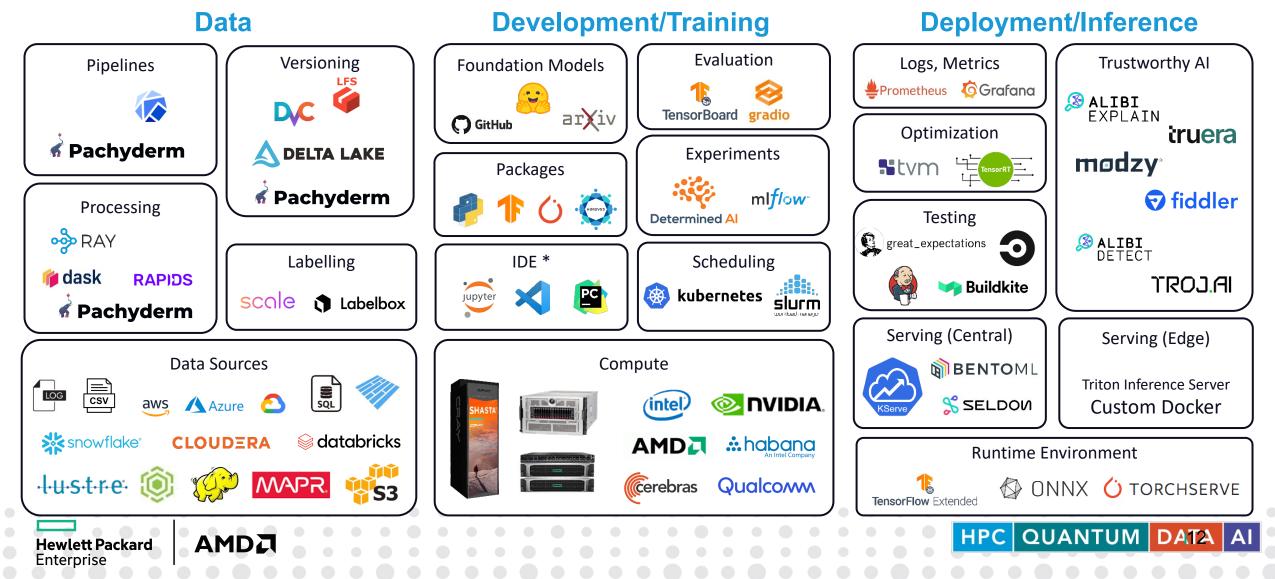


| | Pros | Cons |
|---------------|---|---|
| Private Cloud | Enhanced Security & Privacy Customization Predictable performance Compliance adherence Reduced Risk of vector lock in | Higher Initial Investment Limited Scalability Resource underutilization Complexity |
| Public Cloud | Cost Efficiency Scalability Global Accessibility Maintenance and Updates Innovation | Security and Privacy Concerns Dependency Limited Control Potential Compliance Issues Vendor Lock-In |



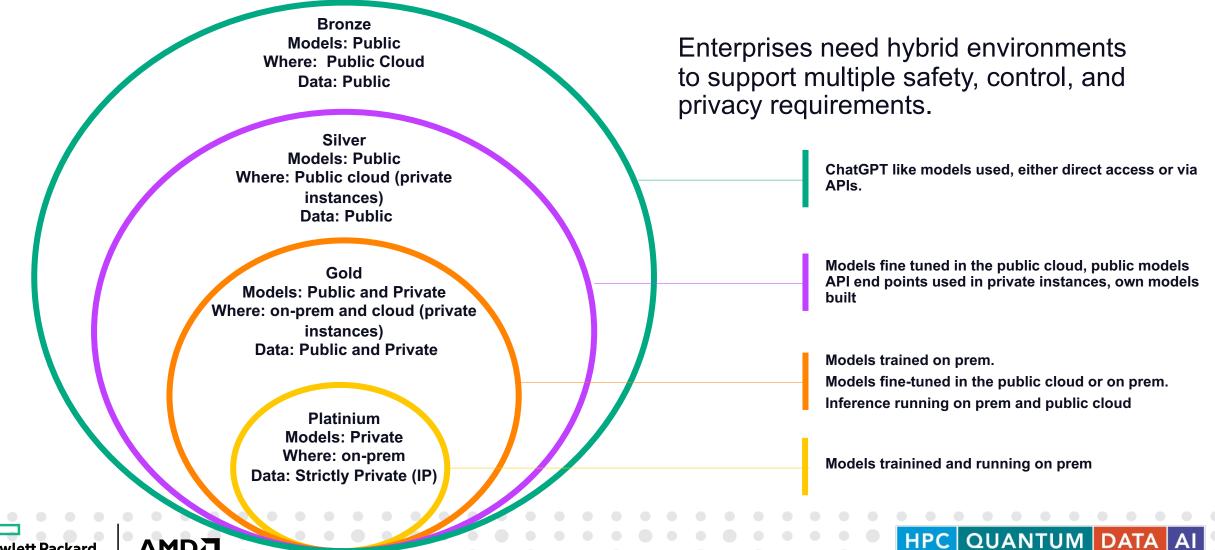
Platform | A Very Complicated Landscape

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Models | How Do You Choose the Right One? It Depends

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Models | Choosing the Right Foundation Model



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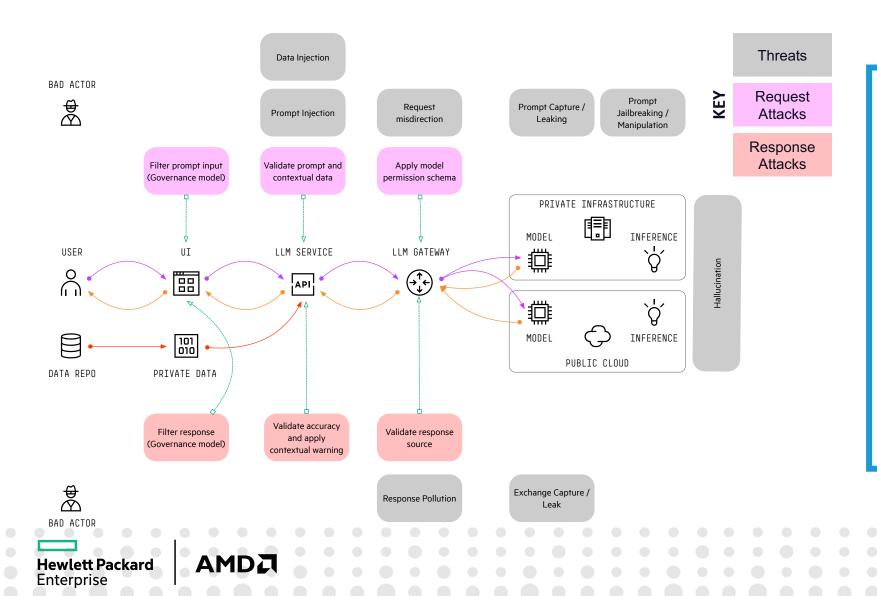
- Select the best AI foundation model for your use case
- Evaluation categories:
 - Project Requirements
 - Model Capabilities
 - Operational & Ethical Considerations
 - Ethical & Governance Considerations
- Align model choice with project goals and organizational policy

| Connect your Greenlake enric | Correcard hardware to HPE ch your experience | | App Versior 9.9.202 name@ Financ | 24 hpe.com ce | ssessm | ent | |
|---|--|---------------------------|--|----------------------|---------------|---------------------------|--|
| Recomme | ndations th | at fit my rec | quiren | nents . | | | |
| Loan Status As | sessment Sco | recard | | | | | |
| Details | Sugger | ted Models | | | | | |
| Author ID | | | | | | | |
| name@hpe.com Version | LLa | MA2 | | | | | |
| 1 | Tota | Score: 45 | | | | | |
| Date 9.9.2024 | | | | | | | |
| Application | GP1 | | | | | | |
| Loan Status Assessment App Description | | | | | | | |
| Description Assessing loan status from clie | | I Score: 44 | | | | | |
| records | | | | | | | |
| Finance | BLC | | | | | | |
| | Tota | I Score: 42 | | | | | |
| Compare side-by-side | | | | | | | |
| Scoring Legend : | | | | | | | |
| Score of 0 | Score of 1 | Score of 2 | | Score of 3 | | Score of 4 | Score of S |
| Not applicable/not a focus | | Meets some criteria; sign | | Adequately meets cri | | Meets criteria well; mino | Fully meets |
| for the model. | poor capability. | improvements needed. | | improvements neede | 1. | improvements suggeste | criteria/excellent capability. |
| Attribute | | BLOG | DM | 1 | GPT-4 | | LLaMA2 |
| Subscription Fee | | No si | ubscription fe- | es; significant oper | ChatGPT Plus | : \$20/month, Teams: \$ | No subscription fees; significant oper |
| One-time License Fee | | Open | and permissi | we under the Resp | N/A | | Free for research and commercial us |
| | | | ational costs o | depend on resourc | | 4 tokens, Output: \$60/ | Operational costs depend on resourc |
| Tokenizing Fee | | N/A | | | N/A | | N/A |
| Image Generation Fees | | | ational costs (| depend on resourc | Included with | in tokenization fee | Operational costs depend on resourc |
| Image Generation Fees Inference Cost | | | | | | | Depends on model size: 78: 16GB V |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi | irement | Туріс | | multiple high-end | | | |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment | irement | Typic yes | | multiple high-end | yes. | | yes |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment On-Premise | | Туріс | | multiple high-end | | | |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment | | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment On-Premise Language Understanding and Complex Reasoning | d Generation | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment On-Premise Language Understanding and | d Generation | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum GPU Memory Requi Cloud Deployment Cn-Premise Language Understanding and Complex Reasoning Customizable and Adaptive A | d Generation | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum GPU Memory Repui Cloud Deployment On Premisie Language Understanding and Complex Reasoning Costomizable and Adaptive A Knowledge Utilization Personalization and Recomm | d Generation | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum (PT) Memory Regul Cloud Deployment On-Premite Language Understanding and Complex Reasoning Costonizable and Adaptive A Costonizable and Adaptive A Costonizable and Adaptive A Nonededge Utilization Human Adjørment Human Adjørment | d Generation N | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Meinimus (RFM Memory Reput Cool Deployment Complex Understanding and Complex Understanding and Complex Understanding and Construited and Adaptive A Consolidade Utilization Personalization and Recorem Haman Alignment Hodal Hodality, Text Dota Use and Recrising Prah | d Generation M endation Systems licies: Consent-Based | Typic yes | | multiple high-end | | | yes |
| Image Generation Fees Inference Cost Minimum (PT) Memory Regul Cloud Deployment On-Premite Language Understanding and Complex Reasoning Costonizable and Adaptive A Costonizable and Adaptive A Costonizable and Adaptive A Nonededge Utilization Human Adjørment Human Adjørment | d Generation M endation Systems licies: Consent-Based ations: Highly Doplainable | Typic yes | | multiple high-end | | | yes |

| Select t | sment | est alig | Loan Status Assessment Scorecard What industry?* Finance | | Defin | ne the application that the model is being use | d for* |
|---|----------------------------------|--------------------|---|--|---------|---|--------------------|
| Language Ur Creative Con Structured D | sment ne scenario(s) that b | est alig | | | Defie | as the application that the model is being use | dfor* |
| Select th Language Ur Creative Con Structured D | ne scenario(s) that b | est alig | Finance | | | or the spinor of the second second as | |
| Language Ur Creative Con Structured D | | est alig | | | A | ssessing loan status from client rec | ords |
| Creative Con Structured D | | narios | n with your application's requ | uirements | | Key Considerations | |
| Creative Con Structured D | derstanding and Generation | Langua | ge modeling, translation, summarization, QA | , sentiment analysis, er | ntity | Bias in language generation, | ensuring relevance |
| Structured D | | extract | | tua and as orr | | | |
| | | | e writing, music composition, and other creat nthesis, text-to-SQL conversion, and similar | | | Reproducing nuances of crea Handling complex coding tas | |
| | | | nthesis, text-to-SQL conversion, and similar book and open-book QA, fact extraction | 211 0/10160 19342 | | Handling complex coding tas Updating knowledge base, va | |
| Complex Rea | | Logic a | nd problem-solving, knowledge and symboli | c reasoning, mathemat | ical | Processing logical rules, abst | |
| | | reasoni Tasks o | ng ombining text, images, and audio, like text-to | o-image translation and | d audio | | |
| Multimodal 1 Computer Vi | | process | | | | Synthesizing information acre Interpreting visual data, envir | |
| Conversation | | | ewriting, music composition, and other creat | | | Understanding nuanced lang | |
| Human Align | | | | inguage conversations, charbot performance assessment | | context Incorporating ethical conside | rations |
| Dericealization and Decommendation | | | | recommendations based on user behavior and preferences | | Analyzing user data, protecti | |
| | th External Environment | | | | x†s | | |
| _ | | | | ustry-specific adaptation, learning from new data | | Modularity, scalability, flexibil | |
| Propose a new one | | | | | | | |
| Data se ta Sensitivity ect Options Highly Confide Internal Use 0 Public | | pecific | ations | | | | |
| ecifications del capabilities, op | erational considerations, and si | afety & eth | ical guardraits | | | | |
| a use policies | | | Transparency of model operations | | 6.4.4 | measures and ethical guardralls | |
| onsent-Based Dat | alise | ~ | Highly Explainable | ~ | | ic Safety Features | ~ |
| | | - | | ÷ | | | + |
| nership type | | | Context window | | Models | | |
| elect the item | | ~ | Select the item | ~ | Selei | ct the item | ~ |
| iality* | | | Languages* | | | | |
| bd | | ~ | Spanish | ~ | | | |
| | | | | | | | 6 Models ava |

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Implementation | What About Security?



- HPC + A WALL STREET
- Fail gracefully and secretly
- Audit user authorization
- Secure access to external resources
- Parameterize and validate all inputs and outputs
- Avoid persisting changes when possible

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Adversarial testing

Implementation | Governance

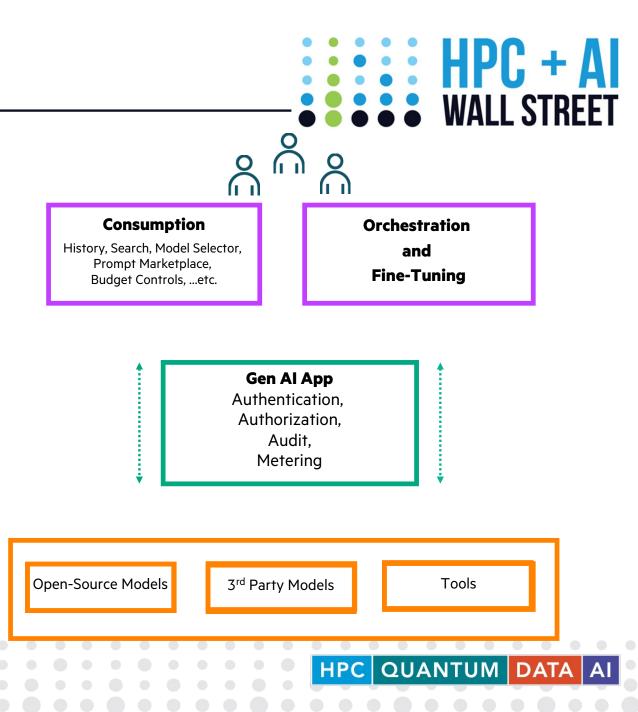
- Allow employees to securely and auditably use new technology to assist with researching, writing, and code generation.
- Keep your **sensitive data on premise**.
- Deliver the ability to choose your model.
- Enforce company policy.
- KEY DELIVERABLES
 - Al guardrails

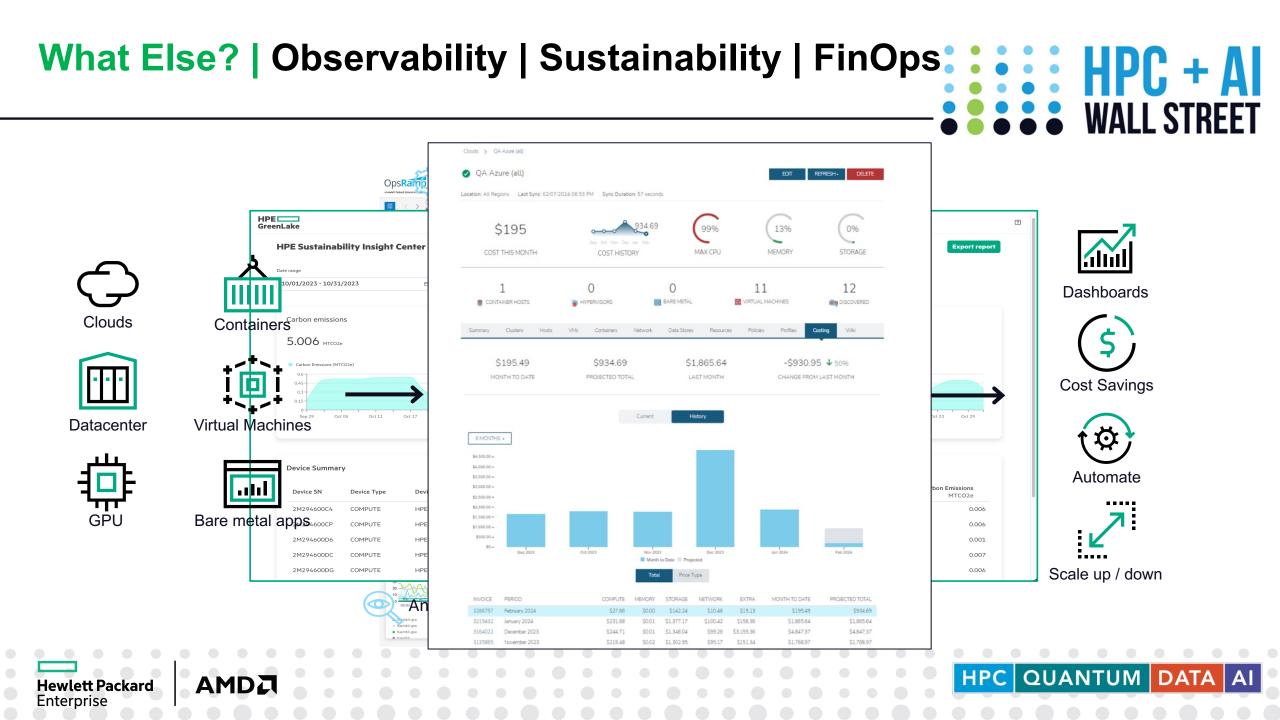
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Model Token Management

• Access to models either on-prem or public cloud





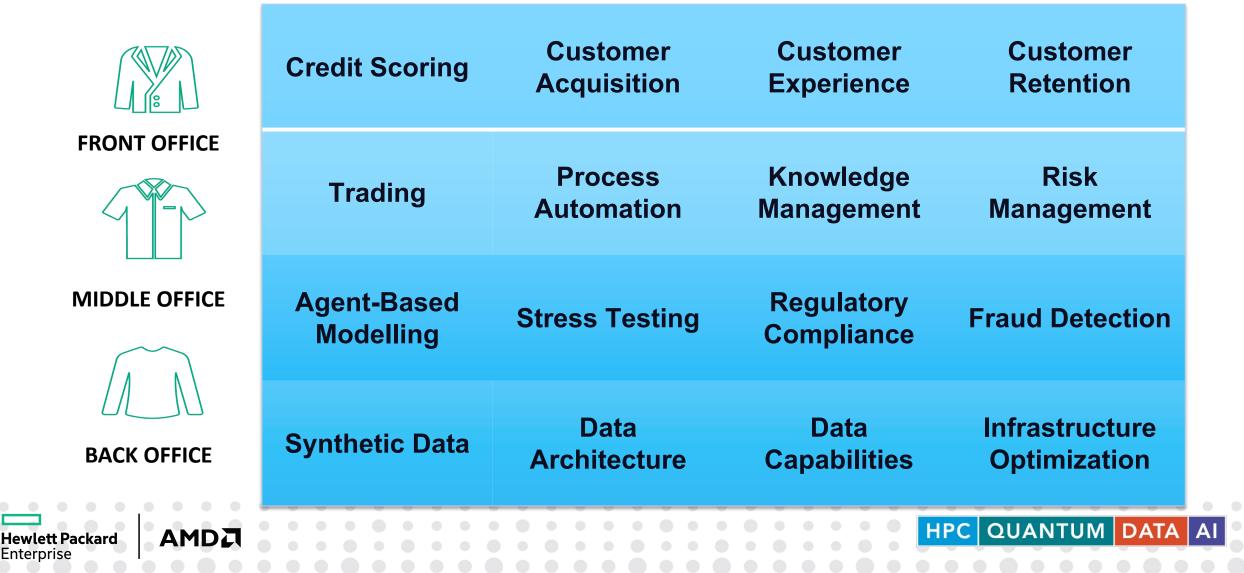


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| Models and Services | Accelerate the journey to Gen AI value for the Enterprise | Model Hub HPE A&PS Services | Model- aaS | |
|----------------------------|--|---|---------------|--------------------------|
| AI Platform | Drive collaborative AI model development from Data preparation to Deployment | Machine Learning Development Environment Machine Learning Data Management HPE AI Essentials | Al PaaS | Observability |
| Data Services | Manage data governance, availability, and security | Data Fabric Data Observability Machine Learning Data Management | Airado | Sustainability FinOps |
| Infrastructure Software | Allow AI models to run efficiently anywhere | Private Cloud for BMaaS, VMaaS, and CaaS. | | |
| Hardware Infrastructure | Optimize the performance of AI workloads with AI native infrastructure | Accelerators: Heterogenous support for Nvidia/AMD/Intel Compute: Cray, Proliant Storage: GL4FL, Networking: Slingshot, Mellanox Best in Class DLC | laaS | |



AI Use Case Families in Financial Services: Spanning Front, Middle, and Back Office



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Getting Started: Pick Your Use Case

| | | | • • • | | L JINLLI |
|-------------------------------|---|---|-------|--------------------------|--------------------------|
| | Regulatory Compliance | Credit Scoring | | Development | Production |
| End-User Group | Internal Users Employee domain SME's (CPA's, compliance officers, Accountants, auditors, (CAO/CFO) Small Userbase, Highly Specialized Able to provide feedback on early iterations Low exposure to bad actors | External and/or Internal Users Customers or customer-facing employees (Loan officers, salespeople, brokers) Large Userbase, Experience levels may vary High exposure to bad actors | | | |
| Use Case Regulations | Regulations mandate what to report to who and when, with specific instructions on preparation methodology "Automate as much as possible to minimize human error" -regulators | Fair lending, anti-discrimination, industry regs High-risk Use Case- AI Act (EU) GDPR/CCPA/Data protection regulations "AI can't do anything a human wouldn't be allowed to do" -regulators | | Security | Observability |
| Source Data (Unstructured) | Public information (Regulatory text, instructions, etc.) Trusted source- directly from regulators Low potential for wrongdoing if leaked | Non-public, Personally Identifiable Information Various sources- trustworthiness may vary High potential for wrongdoing if leaked Data sovereignty & data gravity concerns | | Guardrails Governance | Sustainability FinOps |
| Models Used | Open-source foundation model (Llama3 based) Out of box, no customization | Proprietary vendor or internally developed models Highly customized | | | |
| Hardware Infrastructure | Small-scale inference environment or cloud API solution | Private, hosted, secure AI Inference at Scale cluster AI Factory training resources if building your own models | | | |
| | | | | | |

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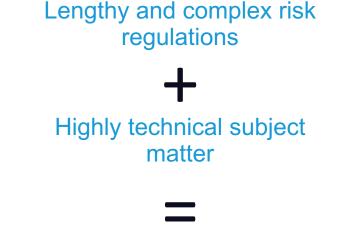
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- Hewlett Packard International Bank DAC, ("HPIB") is a credit institution authorized by the Central Bank of Ireland and owned by HPE.
- The Capital Requirements Regulation 3 (CRR3) is a new regulation spanning 646 pages that goes into effect for banks in the EU on January 1, 2025. It is part of the wider "Basel 3.1"/ "Basel 3 Endgame" regulatory risk framework for global financial institutions.
- The CRR3 requirements have been consistently modified leading up to implementation, and portions still lacked fully defined requirements with less than 1 year to go.
- Risk jargon often requires a CPA with specific experience in the financial sector to understand these standards.

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Complexity, Confusion, Delays, Increased risk



Problem Statement

- The financial services industry complies with a complex and fast-changing web of regulatory requirements.
- International firms are forced to adapt to the pace of change in the standards and laws of multiple regulators, jurisdictions, and governments
- Failure to comply can lead to civil (and even criminal) liability. Fines and sanctions can amount to billions of dollars for large institutions
- Each regulation or report change has the potential to have no impact, low impact, or very high impact and a long adoption process. Expert interpretation is required to determine "Does this change affect us?" and "What do we need to do differently than before?

The greatest compliance challenges I expect to face in 2022 is/are...



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Initial Experiments

| _ | |
|---|--|
| | |
| | |

Experiment 1

Bulk analysis of documents on regulator website via Python to produce categorized CRR3 Implementation Report

Experiment 2

Analysis triggered by frequent regulator website updates, which can provide automated guidance and alerts for impactful changes and opportunities to comment



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Experiment 3

Chatbot experience for realtime self-service with followup questions





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Transparency: End-to-end tracking of new proposed regulatory changes, opening and closing of public comment periods, adoption of final changes, and phased implementation deadlines.



Operational efficiency: Swift document analysis and reporting. Give users opportunity to ask questions on complex subject matter in natural language and receive proactive updates on what they need to know.



Cost savings: Minimize burden of adapting to new legislation, manual reviews and potential fines for non-compliance, and consulting spend.



Increase collaboration: Create wider awareness of upcoming changes.



Part of a Holistic Compliance Strategy

Generative AI analysis is one part of a comprehensive strategy

AI-Powered Analysis:

- Swift document review and extraction of essential data.
- Predictive modeling for potential compliance risks.
- Real-time notifications on regulatory changes.
- Simplification of complex regulatory texts using NLP.

Training & Development:

- Regular workshops to educate staff on compliance matters.
- Role-specific training modules for targeted learning.
- Case studies and simulations for hands-on experience.

Human Oversight:

- Expert panels for manual review and validation.
- Routine audits to ensure AI-generated results align with compliance requirements.
- Feedback loops for continuous AI model refinement.



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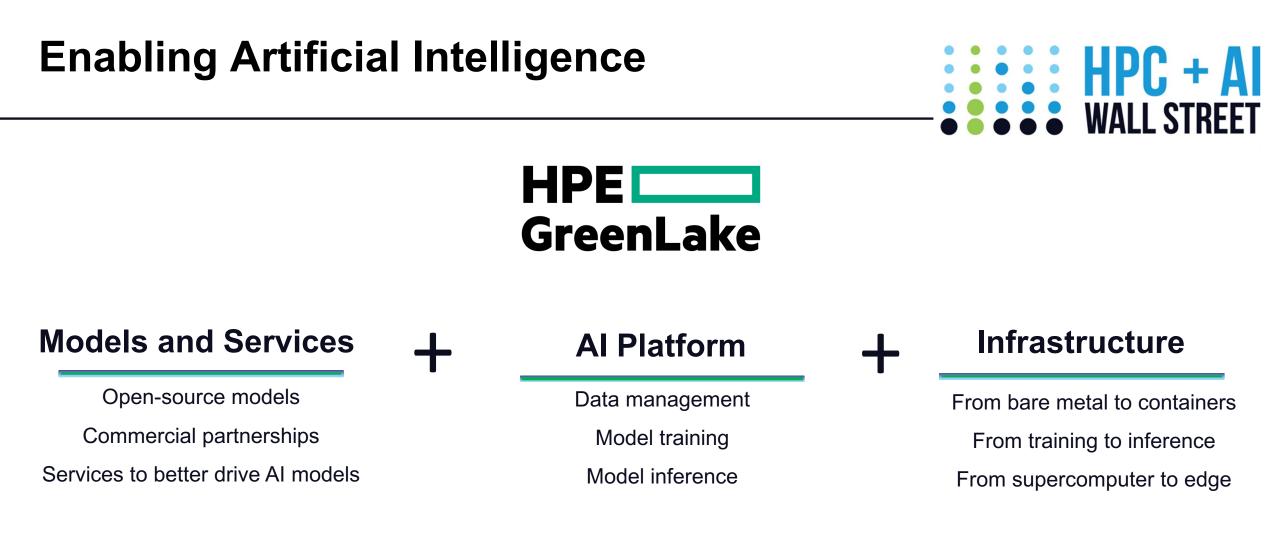
HPE Has Been Preparing for Some Time

November 2018 **July 2021** January 2023 January 2020 **AMP**OL Pachyderm **Ezmeral Unified** Machine Learning November 2016 Analytics Data Management Ezmeral Runtime sgi. Environment -RAY Shasta, ClusterStor, SuperDome Flex, bluedata^{*} Slingshot, and more HPCM, and more "Hyperscale" "Exascale" era era Ezmeral Data Fabric \mathbf{M} Apollo servers Machine Learning introduced MAPR. **Development Environment** MORPHEUS **Hewlett Packard** Enterprise OpsRamp Determined A Frontier operational, 2014 exascale era begins er August 2019 May 2022 June 2021 2024 Enterprise company

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One platform | Vendor neutral | Cloud neutral | Al accessible for all



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2024

Thank You



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