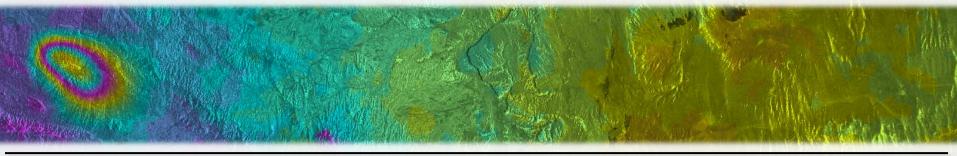
### **GETTING READY FOR (NI)SAR IN THE CLOUD**

#### **Contributors:**

F.J. Meyer<sup>1)2)</sup>, S. Owen<sup>3)</sup>, C. Stoner<sup>2)</sup>, H. Hua<sup>3)</sup>, S. Arko<sup>2)</sup>

<sup>1)</sup>Geophysical Institute, University of Alaska Fairbanks
<sup>2)</sup>Alaska Satellite Facility (ASF), University of Alaska Fairbanks
<sup>3)</sup>Jet Propulsion Laboratory, Pasadena, California









6/2017 HGF Alliance Meeting



• The Cloud and Its Relevance for Large Volume SAR Missions

- The Goals of the Get Ready for NISAR (GRFN) Project
- Current GRFN Status and Progress/Findings
- Conclusions













### THE CLOUD AND ITS RELEVANCE FOR LARGE VOLUME SAR MISSIONS









- Public Cloud [Amazon, Google, Microsoft, ...]
  - Cloud vendors typically have "infinite" resources available
  - Virtual machines handle processing → spinning-up and terminating of VMs provides full performance while only paying for what you use
  - Web Object store for data storage & distribution → full capability but pay-as-you-go
- Private Cloud [your Typical Data Center]
  - Can also be virtual machines for processing, but pay for machines upfront as a sunk cost
  - Build out of data distribution capabilities takes time and must be paid for upfront
- Potential Benefits of the Cloud for Large Volume Remote Sensing Systems
  - Cloud allows you to scale up as you need instead of a big up front sunk cost
  - Cloud also allows researchers to bring their processing to the archive no more waiting to download 100s or 1000s of scenes first!



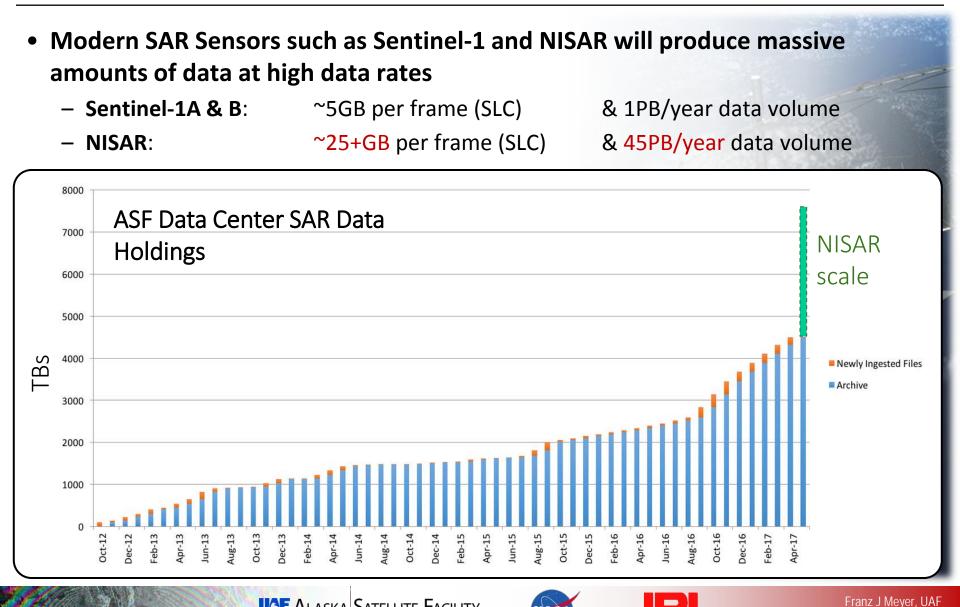




#### (NI)SAR And The Cloud The Relevance of Cloud Processing for Modern SAR Missions



HGF Alliance Meeting, 6/2017 - 5





Jet Propulsion Laboratory



- Modern SAR Sensors such as Sentinel-1 and NISAR will produce massive amounts of data at high data rates
  - Sentinel-1A & B: ~5GB per frame (SLC)
  - NISAR: ~25+GB per frame (SLC)

& 1PB/year data volume & 45PB/year data volume

- Traditional Architecture Won't Scale
  - For Data Centers, too expensive (in cost and time) to scale up to 100s of PB, mostly due to data movement (i.e. processing to storage, user downloads, etc.)
  - For Researchers, a typical two year deep stack will be hundreds of TBs in size
- The Big Data Challenges of Large Volume SAR Missions
  - How to cost effectively ingest and store large volumes?
  - How to scale up and serve large volumes?









#### **Relevance of Cloud Concepts for Mission Operators**

#### • Voluminous SAR Data Becoming a Forcing Function:

 Becoming too large to process SAR in traditional ways e.g. download L0/L1 to process to L2 interferograms and L3 time series

#### Advantages for Processing & Storage

- Bring processing to storage
  - $\rightarrow$  Fewer data movements
  - ightarrow More efficient data handling and processing

#### • Advantages for Serving out Data

- Move user processing to storage
- Spend your time processing, instead of downloading hundreds or thousands of large volume data
- Examples:
  - estimated size of NISAR SLC is 25GB → 5 hours download time per file on 10mbps line → one-year stack (30 SLCs) requires >150 hours of download time







#### Why Should You Care About The Cloud



**Relevance of Cloud Concepts for Science and Applications Communities** 

#### • Infrastructure and Logistics Savings

- Avoid downloading large data volumes (e.g. days to weeks of download time)
- No local storage infrastructure needs
- No expensive compute infrastructure needs
- Massively Parallel Processing Capabilities
  - Opportunity to harness tens of thousands of compute machines
  - Process in hours what used to take days or weeks
  - Limited only by amount you can pay, but pay only for hours that you use

#### Easier Fusion Products

- Opportunity to create products with other datasets in the cloud
- Without the cost to store
- Without the time to download





10 ing



- Processing in the Cloud is Plagued by Same Familiar Issues Related to ...
  - Metadata and data formats
  - Interoperability
- Cloud Resources still have same "Hardware" Failures
  - network timeouts and storage failures
- The Cloud Is not Necessarily Cheaper ...
  - Cloud cost models designed to be similar to total cost of ownership (TCO) of onpremise solutions
- But ... the Cloud Provides ...
  - "infinite resources"
  - pay-as-you-go options
  - Computing resources closer to the archive











# THE GOALS OF THE <u>GET READY FOR NISAR</u> (GRFN) PROJECT





#### The Get Ready For NISAR (GRFN) Project Project Goals



- Project Goals:
  - Understand cost implications of various cloud-based and hybrid architectures for NISAR science data system (processing) and data center (storage)
  - Get science community familiar and comfortable early on with interacting with and working on large SAR datasets in a cloud environment.
- Approach:
  - Build a prototype NISAR processing system in the cloud based on Sentinel-1 SAR data
  - Derive Sentinel-1 SAR data products to socialize SAR data products to science communities
    - L2 products (interferograms; covariance information) for expert SAR community
    - L3 products (deformation rates; time series; ...) for broader science community
    - On-demand processing
- SAR products used to understand how scientists will interact with NISAR data in the cloud to accurately estimate cost implications







#### **The Get Ready For NISAR (GRFN) Project** Expected Key GRFN Science Data System Capabilities



- Prototype NISAR processing system with Sentinel-1 as proxy
- Up to L3 science data products for science focus areas [solid earth; ecosystems; cryosphere; applications]
- Simulate NISAR processing scenarios in the cloud
  - Forward stream processing ("keep up")
  - Bulk reprocessing
  - On-demand processing
  - Urgent response
- Cloud-based collocation of processing system with ASF data center
  - Establish cloud-based high-performance data delivery of L2 Sentinel-1 science data products from processing system to the ASF data center.
- Costing cloud economics
  - Perform analysis necessary to produce costing reports needed for NISAR.







### The Get Ready For NISAR (GRFN) Project

**Key GRFN ASF Data Center Investigations** 



- Ingest at NISAR Scale (bandwidth and volume)
  - Achieved through shared storage by collocation of SDS and data center

#### • Archive and Distribution (storage & distribution costs, bandwidth)

- Lifecycle and storage temperature (hot  $\rightarrow$  cold  $\rightarrow$  hot)
- Data distribution via Earthdata Search Client & Vertex

#### • Cost and Performance Implications of On-demand Processing Scenarios

- Standard product creation on-demand (virtual archive)
- Bulk re-processing from various storage types
- End-user processing system (bring processing to the data)

#### • Science Community Outreach

- L2 product usability and convenience packaging
- Fully Public Data access as a Beta product











### CURRENT GRFN STATUS AND FINDINGS







What Has Been Built/Done So Far



- GRFN SDS and Storage System Built in Public Cloud Environment
  - Now co-located with data center in AWS us-east region
- Collocated SDS and Data Center to Avoid Large L0-L2 Data Volume Movement
  - Capable of ingesting 10 Gbps at forward processing rates and 50 Gbps forward processing plus bulk reprocessing load
- Automatic L2 (Solid Earth) Processing, Ingestion, and Distribution
  - Products available via GRFN website, Vertex, and Earthdata Search client
- On-Demand L2 (Solid Earth) Processing, Ingestion, and Distribution
  - Limited scale, but also available via GRFN website, Vertex, and Earthdata Search client
- Multi-Temperature Storage Prototype
  - Used AWS native lifecycle policies and tracking to determine best (lowest cost) mix of storage classes based on Sentinel-1 distribution activity





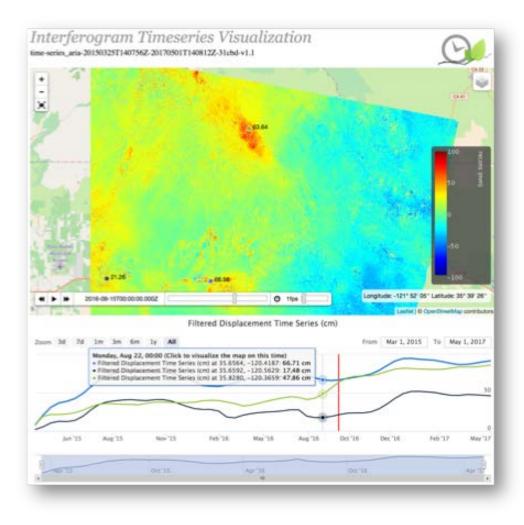


#### **Current GRFN Status and Findings**

What Has Been Built/Done So Far



- Cloud-Based Prototype of On-Demand L3 Displacement Time Series
  - SBAS-type time series solution computed in AWS cloud
  - Time series analysis collocated with L2 interferogram stack







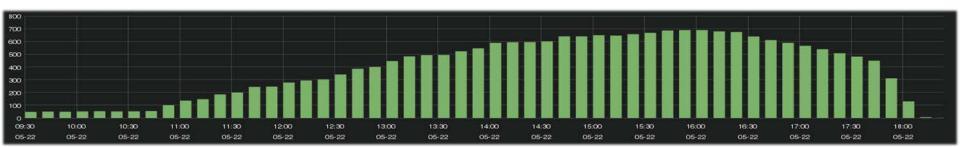


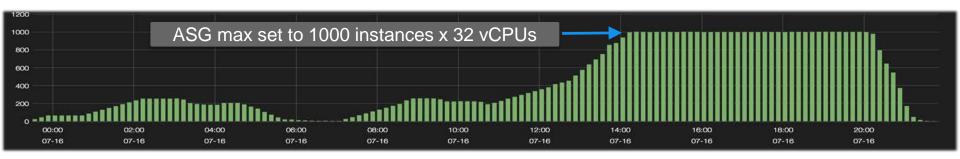
#### Load Testing: An Example of Cloud Scalability



- The size of the science data system compute nodes can automatically grow/shrink based on processing demand
- Auto Scaling group policies can be added
  - E.g., setting max scaling size to curb costs

Auto scaling enabling runs of over 100,000 **vCPUs** 













#### **Studying Performance and Cost Implications of Cloud Architectures**

#### 1. Egress Costs:

- Every file downloaded from the cloud will incur costs
- Depending on the download behavior of users, egress can be major cost factor for cloud-based architectures!

→ To Save Egress Costs, Hybrid Architecture Including Cloud, Edge Locations, and On-Premise Components is Recommended

- Store infrequently used data in the cloud (deep [cheap] storage)
- Serve "hot data" from on-Premise or cached edge locations → cheaper, even incl. direct connect & Hardware costs
- Distribution of data among cloud, edge, and on-premise locations may be learned from user behavior







### **Cloud Architecture and Egress Costs**

#### Learn From Sentinel-1 SLC Download Patterns



Sentinel-1 SLC usage largely localized around Major Hazard Zones

Hot data zones (green areas) take care of >70% of Sentinel-1 SLC data downloads

 $\rightarrow$  Serving out data in green areas from onpremise or edge location can lead to large savings in egress costs











#### 2. Compute Costs

#### → Mix Fast-Path and Slow-Path Processing Streams to Curb Costs

- Spend reserved instance compute time on Minimum Viable Product requirements to get product to serving
- Queue extra compute to the slow path on Spot Market instances to be cheaper but possibly later (i.e. Browse generation)

#### 3. Storage Costs

#### $\rightarrow$ Make storage decision during Ingest

 Per file, likely based on Area of Interest (AOI), decide on most appropriate temperature storage class during ingest (i.e. Glacier [cheapest storage] for products that will likely never be touched)









#### Advantages of Moving End-User Processing to Storage

- Fewer data moves [e.g. egress costs]
- Faster results!
- Process massively scalable next to storage
- No hardware requirement/maintenance for end user
- Fusion products with other datasets in the Cloud

How can we help you get started in the cloud?







#### Conclusion



- GRFN has shown the viability of cloud solutions for SAR data missions
- Various cloud architectures have been tested for performance and cost implications
- Some first findings have been presented today
- Not all functionality has been completely built and not all cost analyses are completed but progress is being made on all fronts

We just finished year #1 of GRFN – two more project years to come!!







#### Looking for Beta Testers to Assess GRFN Products & Services



- Earthdata Search
  - Search for "GRFN"
  - <u>https://search.earthdata.nasa.gov/search?q=GRFN&ok=GRFN</u>
- Vertex
  - Missions Tab, Beta Products
  - <u>https://vertex.daac.asf.alaska.edu/</u>
- GRFN/ARIA Science Data System
  - <u>https://aria-search.jpl.nasa.gov/</u> (data products, account sign up)
  - <u>https://aria.jpl.nasa.gov/</u> (general info)

### Questions?

#### Contact: fjmeyer@alaska.edu







#### **AGU FALL MEETING** New Orleans | 11–15 December 2017

### Session ID: 26762Session Title: G015. Recent Advances in SAR and InSAR Processing, Analysis, and Cloud Computing

Section/Focus Group: Geodesy Link: <u>https://agu.confex.com/agu/fm17/preliminaryview.cgi/Session26762</u>

#### Submit abstracts latest by Wed., August 2, 2017











### A FEW ADDITIONAL LESSONS LEARNED







## NASA

#### • Engineering

- Development in cloud environment is more efficient and agile than traditional means.
- Some cloud vendors (e.g. AWS) have strong community following that builds capabilities in and around cloud vendor (e.g., swiftstack storage)
- Cloud vendors provide suite of hardened and ready-to-use data system tools and services

#### • Infrastructure

- Collocation of compute and storage in same cloud region removes unnecessary data movements (and associated costs)
- Moving processing to data archive improves efficiency for large data volume scenarios such as L3 time series from deep stacks
- Data life-cycle policies can help to lower storage costs





