Deep Underground Neutrino Experiment (DUNE) Technical Proposal

Software and Computing Overview

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Chapter 1

² Overview of Software and Computing

3 1.1 Overview

Offline computing for DUNE faces new and considerable challenges due to the large scale and 4 diverse physics goals of the experiment. In particular, the advent of liquid argon (LAr) TPC's with 5 exquisite resolution and sensitivity, combined with enormous physical volumes, creates challenges 6 in acquiring and storing large data volumes and in analyzing and reducing them. The computing 7 landscape is changing rapidly, with the traditional HEP architecture of individual cores running 8 Linux being superseded by multi-core machines and GPU's. At the same time, algorithms for LAr 9 reconstruction are still in their infancy and developing rapidly. As a result, we have reason to be 10 optimistic about the future but we are not able to predict it accurately. The ProtoDUNE single 11 and dual phase tests at CERN in the fall of 2018 will provide a wealth of data that will inform the 12 future evolution of the DUNE computing models. 13

The DUNE offline computing challenges can be classified in several ways. We will start with the different detector/physics configurations that drive the large scale data storage and reconstruction. This discussion leans heavily on the data acquisition (DAQ) design described in Volume 2: The Single-Phase Far Detector and Volume 3: The Dual-Phase Far Detector of the DUNE Technical Proposal.

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¹⁹ **1.1.1 Detectors**

The DUNE experiment will consist of four 17 kT far detector modules located at the Sanford Underground Research Facility, using either single or dual phase Liquid Argon TPC's, and an, as yet unspecified, near detector at Fermilab. The proposed full-size 17 kT modules for the far detectors will have an active volume 12m high, 14.5m wide and 58m long.

1 1.1.1.1 Single-phase estimates

Each single-phase (SP) module will consist of 150 alternating vertical cathode and anode planes
spaced 3.5 m apart and operated at 180 kV for a 500 V/cm drift field. The anode planes are made
up of anode plane assembly (APA)s which are 6.3 m tall by 2.3 m wide and have 2,560 readout
channels each. Each channel is sampled with 12-bit precision every 500 nsec. For modules of this
size, drift times in the liquid argon are of order 2.5 ms and raw data sizes before compression are
of order 6 GB per module per 5.4 ms readout window. With no triggering and no zero suppression
or compression, the raw data volume for four modules would be of order 145 exaB/year.

9 1.1.1.2 Dual-phase technology

For dual-phase, electrons drift the full height of the cryostat, emerge from the liquid and are 10 collected - after gas amplification, on an grid of instrumented pads at the top of the detector. The 11 WA105 3x1x1 m test of this technology ran successfully in the summer of 2017[2]. Each 17 kT 12 module will have 153,600 channels. Drift time in the liquid argon is 7.5 ms. Given 20,000 samples 13 in an 8 ms readout, the uncompressed event size is 4.2 GB (for 1 drift window). Due to gas 14 amplification, the signal to noise ratio is quite high, allowing loss-less compression to be applied 15 at the front-end with a compression factor of ten, bringing the event size/module to 0.42 GB. 16 Recording the entire module drift window can be considered a pessimistic figure, since events are 17 normally contained in smaller detector regions. A far detector module can be treated as 20 smaller 18 detectors (with similar number of readout channels to the prototype currently being constructed 19 at CERN), running in parallel, each one defining a Region of Interest (ROI). For beam or cosmic 20 events it is possible to record only the interesting ROI(s) with the compressed size of a single ROI 21 being 22 MB. 22

23 1.1.1.3 Beam coincident rates

Requiring coincidence with the Long-Baseline Neutrino Facility (LBNF) beam will reduce the 24 effective live-time from the full 1.2-1.5 sec beam cycle to a 5.4 ms (8 ms for DP) readout window 25 coincident with the 10 microsecond beam spill, leading to an uncompressed data rate for beam-26 coincident events of around 20 GB/sec for four 17 kT single-phase detector modules (~ 16 GB/sec 27 for dual-phase), still too high to record permanently. Only a few thousand true beam interactions 28 in the far detectors are expected each year. Compression and conservative triggering based on 29 photon detectors and ionization should reduce the data rate from beam interactions by several 30 orders of magnitude without sacrificing efficiency. 31

32 1.1.1.4 Near detector

The near detector configuration is not yet defined but we do have substantial experience from T2K and MicroBooNE at lower energies, and MINERvA at the DUNE beam energies on cosmic and $_1~$ beam interactions under similar conditions. We can expect that a near detector will experience \sim

² 5-10 beam interactions/beam pulse and non-negligible rates of cosmic rays, spread over an area of

³ a few square meters. MicroBooNE experience and ProtoDUNE simulations indicate compressed

⁴ event sizes of 100-1000 MB, leading to yearly data volumes of 2-20 PB. Storing and disentangling

⁵ this information will be challenging but comparable to the ProtoDUNE data expected in 2018.

6 1.1.2 Physics Challenges

7 DUNE physics will consist of several different processes with very different rates and event sizes.

8 1.1.2.1 Long-baseline neutrino oscillations

⁹ Neutrino oscillation measurements will require a near detector operating in a high rate environment and far detectors in which beam-coincident events are rare but in time with the beam spill and of sufficient energy to be readily recognizable. Studies discussed in the DAQ section of Technical Proposal Volumes 2 and 3 indicate that high efficiencies are achievable at an energy threshold of 10 MeV, leading to event rates for beam-initiated interactions of $\sim 6,400/year$ and an uncompressed data volume of around 30 TB/year per 17 kT single-phase module.

^{tab:dag-tabadagtdataprates-dp} Tables 1.1 and 1.2 summarize the event and data rates after appropriate filtering from the DAQ section of Volumes 2 and 3 of the Technical Proposal.

17 1.1.2.2 Processes not in synchronization with the beam spill

These include supernova physics, atmospheric neutrinos, proton decay, neutron conversion and solar neutrinos. These processes are generally at lower energy, making triggering more difficult, and asynchronous, thus requiring an internal or external trigger. In particular, supernovae signals will consist of a large number of low-energy interactions spread throughout the far detector volume over a time period of 1-30 seconds. Buffering and storing 10 seconds of data would require around 2000 readout windows, or around 50 TB per supernova readout. At a rate of one such event/month, this is 600 TB of uncompressed data per module/year.

²⁵ **1.1.2.3** Calibration

In addition to physics channels, continuous calibration of the detectors will be necessary. It is likely that, for the far detectors, calibration samples will dominate the data volume. Cosmicray muons and atmospheric neutrino interactions will provide a substantial sample for energy and position calibration. Dedicated runs with radioactive sources and laser calibration will also generate substantial and extremely valuable samples. Table 1.1 includes estimates for the singlephase far detector. Cosmic ray and atmospheric neutrino signals collected for calibration make Table 1.1: Anticipated annual, uncompressed data rates for a single SP module from the far detector Technical Proposals. The rates for normal (non-SNB triggers) assume a readout window of 5.4 ms. In reality, lossless compression will be applied which is expected to provide as much as a $4\times$ reduction in data volume for each SP module.

| Event Type | Data Volume PB/year | Assumptions |
|---------------------------------------|------------------------|--|
| Beam interactions | 0.03 | 800 beam and 800 rock muons; 10 MeV threshold in coincidence with beam time; include cosmics |
| Supernova candidates | 0.5 | 30 seconds full readout, average once per month |
| Cosmics and atmospherics | 10 | 10 MeV threshold |
| Radiologicals (^{39}Ar snd others. | ≤ 1 | fake rate of \leq 100 per year |
| Front-end calibration | 0.2 | Four calibration runs per year, 100 mea- surements per point |
| Radioactive source calibration | 0.1 | source rate ${\leq}10~\text{Hz};$ single fragment readout; lossless readout |
| Laser calibration | 0.2 | $1{	imes}10^6$ total laser pulses, lossy readout |
| Random triggers | 0.06 | 45 per day |
| Trigger primitives | ≤ 6 | all three wire planes; 12 bits per prim- itive word; 4 primitive quantities; ³⁹ Ar- dominated |

rates-sp

Table 1.2: Anticipated annual, uncompressed data rates for one DP module. The rates for normal (non-SNB triggers) assume a readout window 7.5 ms. These numbers do not include lossless compression which is expected to provide as much as a $10 \times$ reduction in data volume.

| Event Type | Data Volume PB/year | Assumptions |
|---|------------------------|---|
| Beam interactions (DP) | 0.007 | 800 beam and 800 rock muons; this be- comes 700 GB/year if just 2 ROIs/event are dumped on disk |
| Supernova candidates (DP) | 0.06 | 10 seconds full readout, all ROIs are dumped on disk |
| Cosmics/atmospherics (DP) | 2.33 | This becomes 230 TB/year if two ROIs/event are dumped on disk |
| Radiologicals (${}^{39}Ar$ snd other). | ≤ 1 | fake rate of \leq 100 per year |
| Miscellaneous calibrations | 0.5 | similar to SP |
| Random triggers | 0.02 | 45 per day |
| Trigger primitives | ≤ 6 | similar to SP |

¹ up the bulk of the uncompressed SP data volume at ~ 10 PB/year per 17 kT module and will ² dominate the rates from the far detectors.

3 **1.1.2.4** Zero suppression

⁴ The data volumes discussed above are for un-zero-suppressed data. Efficient zero suppression ⁵ mechanisms can substantially reduce the final data volume but previous experience in HEP indi-⁶ cates that signal processing must be done carefully and often happens well into data-taking when ⁷ the data are well understood. Experience from MicroBooNE and the ProtoDUNE experiments ⁸ will aid us in developing these algorithms but it is likely that they will be applied later in the ⁹ processing chain for single-phase. No zero-suppression is planned for dual-phase.

The constrained environment at the Sanford Lab motivates a model where any further data reduction via zero-suppression is done downstream, either on the surface or after delivery to computing facilities at FNAL or elsewhere. This could be analogous to the HLT's used by LHC experiments. The relative optimization of data movement and processing location is an important consideration

¹⁴ for the design of both the DAQ and offline computing.

15 **1.1.3 Summary**

In summary, uncompressed data volumes will be dominated by calibration for the far detectors ($\sim 10 \text{ PB/year/module SP}$ or $\sim 3 \text{ PB/year/module DP}$) and by beam and cosmic ray interactions in the near detectors (2-20 PB/year). With four far detector (FD) modules but a conservative

¹⁹ factor of four for compression a total compressed data volume of 12-30 PB/year is anticipated.

After discussion with the SP Trigger/DAQ group, we asked them to include as limits in their design a maximum data transfer rate from the far detectors to Fermilab of 100 Gbit/s, which is consistent with projected network bandwidths in the mid 2020's and a limit of 30 PB/year raw data stored to tape.

²⁴ 1.2 Building the computing model

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The DUNE computing model is a work in progress. We can expect that major advances will take place over the next year on several fronts, with data from ProtoDUNE and the full incorporation of lessons from MicroBooNE into LArSoft.

²⁸ The overall model can be divided into several major parts: Infrastructure, Algorithms and Adap-

- ²⁹ tion for the future. These are in different stages of planning and completion. An overarching
- ³⁰ theme is evaluating and using community codes and resources wherever possible.

1 1.2.1 Infrastructure

This category includes the wealth of databases, catalogs, storage systems, compute farms and the software that drives them. HEP fortunately has already developed much of this technology and our plan is to adopt pre-existing systems wherever possible. As DUNE is a fully global experiment, integrating the resources of multiple institutions is both an opportunity and a logistical challenge.

⁶ We are currently planning to have the primary raw data repository at Fermilab, with derived ⁷ samples and processing distributed among collaborating data centers. For ProtoDUNE, raw data ⁸ will also be stored at CERN. Data processing is being designed to run on HEP grid resources ⁹ with significant ongoing effort to containerize it so that we can make use of heterogenous resources ¹⁰ worldwide.

11 1.2.1.1 Core HEP code infrastructure

¹² We plan to use shared HEP infrastructure wherever possible. Notably the ROOT $\begin{bmatrix} root \\ 3 \end{bmatrix}$ and geant4 $\begin{bmatrix} root \\ 4 \end{bmatrix}$

¹³ 5] frameworks. For event simulation, we plan to use and contribute to the broad range of available

¹⁴ generators (GENIE [6], NuWro [7]...) shared with the worldwide neutrino community.

In addition, we are using the infrastructure developed for the LHC and the Intensity Frontier experiments at Fermilab, notably grid infrastructure, the *art* framework and the Sequential Access via Metadata (SAM) data catalog. The NOvA and MicroBooNE experiments are already using these tools for distributed computing and the ProtoDUNE data challenges are integrating CERN and Fermilab storage and CPU resources. We are now extending this integration to the institutions within the collaboration who have access to substantial storage and CPU resources.

21 1.2.2 Algorithms

This category includes the simulations, signal processing and reconstruction algorithms needed to reconstruct and understand our data. Algorithms are currently under development but are informed by existing general codes (for example GENIE and geant4) and the experience of other liquid argon experiments as encoded in the shared LArSoft project. Simulations are quite advanced but full understanding of reconstruction algorithms will need real data from ProtoDUNE.

27 1.2.2.1 External products

The image-like nature of TPC data allows us to make use of external machine-learning systems such as TensorFlow[8], Keras[9] and Caffe[10]. Many of these are being evaluated for pattern recognition. While they encapsulate a wealth of experience, they are also somewhat volatile as driven by external needs. We must have access to and preserve the underlying source codes in order to maintain reproducibility.

1 1.2.3 Adaptability

As the experiment will be expected to run at least two decades past the present we must be prepared for the inevitable and major shifts in the underlying technologies that will occur. The ability to keep operating over decades almost requires that we emphasize open source over proprietary technologies for most applications. We should also plan to be able to utilize and support a large range of compute architectures in order to fully utilize the resources available to the collaboration.

Table 1.3 summarizes the responsibilities of the Software and Computing group and Reconstruction
 and Algorithms groups for both DUNE and ProtoDUNE.

Table 1.3: Computing Tasks - see the ProtoDUNE section for details on current status.

| Task | Status |
|---|---------------------------------|
| | |
| Code management | in place |
| Documentation and logging of DAQ and detector | in design |
| configurations | |
| Data movement | design rates achieved for short |
| | periods |
| Grid processing infrastructure | early version in use for data |
| | challenges |
| Data catalog | sam, in place |
| Beam instrumentation and databases | ifbeam, in test |
| Calibration and Alignment processing | needs development |
| Calibration and Alignment databases | needs development |
| Noise reduction | tested in simulation |
| Hit finding | tested in simulation |
| Pattern recognition algorithms | tested in simulation |
| Event simulation | use existing software |
| Analysis formats | no common format |
| Distribution of analysis samples to collaborators | needs development |

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⁹ 1.2.4 Downstream activities

¹⁰ The previous sections have concentrated on movement and recording of raw data, as that is most

¹¹ time-critical and drives the primary data storage requirements. Basic simulation and reconstruction

¹² algorithms are in place, but other components, in particular physics analysis models, are in a much

¹³ earlier stage of development.

1 **1.2.4.1** Simulation

Our simulation efforts will build on the combined experience of multiple neutrino experiments 2 and theorists for inputs. We already have a solid foundation of event and detector simulation 3 codes thanks to prior work by the LArSoft and event generator teams. However, even with good 4 software in place, detector simulation in detectors of this high resolution is highly CPU and memory 5 intensive and we are actively following projects intended to exploit HPCs for more efficiency. As 6 simulation is much less I/O and database intensive than raw data reconstruction, (due in part to 7 our ability to trigger efficiently on signal), we can anticipate resource contributions to this effort 8 being distributed across the collaboration and grid resources worldwide. Simulation sample sizes 9 orders of magnitude larger than the number of beam events in the far detector will be reasonably 10 easy to achieve while near detector samples would need to be prohibitively large to equal the 11 millions of events that will be collected every year. 12

13 **1.2.4.2** Reconstruction

¹⁴ We have working frameworks for large-scale reconstruction of simulated and real data in place ¹⁵ thanks to the LArSoft effort. These, and the simulations, have been exercised in large scale data ¹⁶ challenges. Optimization of algorithms awaits data from ProtoDUNE.

17 1.2.4.3 Data Analysis

The data analysis framework has not been defined yet. We are working to build a distributed model, where derived data samples are available locally and regionally, similar to the LHC experiments. Provision of samples of ProtoDUNE data and simulated samples for the Technical Design Report will help define the analysis models that are most useful to the collaboration. However, previous experience on the Tevatron experiments indicates that data analysis methods are often best designed by end-users rather than imposed by central software mandates.

²⁴ 1.3 Planning inputs

²⁵ 1.3.1 Running experiments

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²⁶ The Fermilab intensity frontier program experiments (MINOS [11], MINERvA [12], MicroBooNE [13]

²⁷ and NOvA 14 have developed substantial computing infrastructure for the storage, reconstruction

²⁸ and analysis of data on size scales of order 5% that of full DUNE and comparable to the Proto-

²⁹ DUNE experiments. While the LArTPC technology requires unique algorithms, the underlying

³⁰ compute systems, frameworks and database structures already exist and are being adapted for use

31 on both ProtoDUNE and DUNE.

- For algorithms, the MicroBooNE 15 experiment has been running since 2015 with a liquid Argon-1
- TPC which shares many characteristics with the DUNE APA's. MicroBooNE has, over the past 2
- year, published studies of noise sources and signal processing [16, 17], novel pattern recognition 3
- strategies [18, 19] and calibration signatures such as Michel electrons and cosmic rays [20, 16]. 4 DUNE shares both the LArSoft software framework and many expert collaborators with Micro-
- BooNE and is taking direct advantage of their experience in developing simulations and recon-6
- struction algorithms. 7

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1.3.2 ProtoDUNE

The ProtoDUNE single and dual-phase experiments will run in the Fall of 2018. While the detectors 9 themselves have only 4-5% of the channel count of the final far detectors, the higher beam rates 10

(up to 100 Hz) and the presence of cosmic rays make the expected instantaneous data rates of 11

2.5 GB/sec from these detectors comparable to those from the full far detectors and similar to 12

those expected for a near detector. 13

In addition, the entire suite of issues in transferring, cataloging, calibrating, reconstructing and an-14

alyzing these data are the same as for the full detectors and are driving the design and development 15

of a substantial array of computing services necessary for both ProtoDUNE and DUNE. 16

Substantial progress is already being made on the infrastructure for computing, through a series of 17 data challenges in late 2017 and early 2018. Development of reconstruction algorithms is currently 18 restricted to simulation but is already informed by the experience with MicroBooNE data. 19

In summary, most of the important systems are already in place or are in development for full 20 ProtoDUNE data analysis and should carry over to the full DUNE. We have indicated where 21 infrastructure is in place in table 1.3. 22

Single-Phase prototype 1.3.2.1 23

The single-phase prototype (ProtoDUNE-SP) will utilize six prototype APAs with the full drift 24 length envisioned for the final far detector. In the single-phase detector, the readout planes are 25 immersed in the liquid Argon and no amplification occurs before the electronics. ProtoDUNE-SP 26 is being constructed in the NP04 test beamline at CERN and should run with tagged beam for 27 around 6 weeks in the Fall of 2018. In addition cosmic ray commissioning beforehand and cosmic 28 running after the end of beam are anticipated. Table 1.4 shows the anticipated data rates and 29 sizes. 30

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| Parameter | Value |
|--|-------------------------|
| Trigger rate | 25 Hz |
| Spill duration | 4.5 s |
| SPS Cycle | 22.5 s |
| Readout time window | 5 ms |
| # of APAs to be read out | 6 |
| Uncompressed single readout size (per trigger) | 230.4 MB |
| Lossless compression factor | 4 |
| Instantaneous data rate (in-spill) | 1440 MB/s |
| Average data rate | 576 MB/s |
| 3-Day buffer depth | 300 TB |
| Planned total statistics of beam triggers in 42 beam | 18M |
| days | |
| Planned overall storage size of beam events | 1.0 PB |
| Requested storage envelope for ProtoDUNE-SP | 5 PB at FNAL, 1.5 PB at |
| | CERIN |

| Table 1.4: Parameters for the ProtoDUNE-SP run at (| CERN |
|---|------|
|---|------|

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1.3.3 Dual-Phase prototype

The Dual-Phase prototype will either run in the NP02 beamline in late Fall 2018, or run at high rate on cosmics soon thereafter. Given the most recent construction schedule for the Dual-Phase prototype is now likely that the collaboration will forgo beam data taking and focus on detector performance assessment with two charge-readout plane (CRP)s read out and cosmics only. ProtoDUNE-DP will then run with cosmics at a rate going from 20 to 100 Hz from late Fall 2018 to at least April 2019. During six months of operation, with 50% efficiency, ProtoDUNE-DP is expected to collect about 300 million cosmic triggers at various rates, corresponing to a total data values of 2.4 BP

⁹ data volume of 2.4 PB.

Table 1.5: Parameters for a 6 month ProtoDUNE-DP cosmic run at CERN

| Parameter | Value |
|--|-----------------|
| Trigger rate | 20-100 Hz |
| CRPs read out | 2 |
| Uncompressed single readout size (per trigger) | 80 MB |
| Lossless compression factor | 10 |
| Maximum data rate | \leq 800 MB/s |
| Cosmic rays over a 6 month run | 300 M |
| Requested cosmic storage envelope for ProtoDUNE- | 2.4 PB |
| DP | |

1 1.3.4 Data Challenges

² Computing and software is performing a series of data challenges to ensure that systems will ³ be ready when the detectors become fully operational in the summer of 2018. To date we have ⁴ performed challenges using simulated single and double-phase data and real data from cold-box ⁵ tests of single-phase electronics. We anticipate average rates of ~ 600 MB/sec but have set our ⁶ design criteria at 2.5 GB/sec for data movement from the experiments to CERN Tier-0 storage ⁷ and from there to Fermilab.

In data challenge 1.5 in mid-January 2018, dummy data based on non-zero-suppressed simulated
events were produced at EHN1 and successfully transferred via 10-50 parallel transfers to the EOS
(EOS) disk systems in the CERN Tier-0 data center at a sustained rate of 2 GB/sec. Transfers to
dCache/Enstore at Fermilab achieved rates of 500 MB/sec.

¹² Data challenge 2.0 was performed in early April 2018 is still being analyzed but preliminary ¹³ estimated rates of 4 GB/sec from Experiment Hall North One (EHN1) to the tier-0 were achieved ¹⁴ over several days. Rates to Fermilab disk cache were 2 GB/sec. Movement from FNAL disk ¹⁵ cache to tape was substantially slower due to configuration for a lower number of drives than ¹⁶ needed and contention for mounts with other running experiments. Fermilab is in the process of ¹⁷ upgrading their tape facilities but we may require additional offsite buffer space if data rates from ¹⁸ the experiments exceed the ~ 600 MB/sec expected.

A subsample of the data was used for data quality monitoring at CERN and the full sample was reconstructed automatically on the grid using resources at multiple sites, including CERN. Our overall conclusion from this test is that most components for data movement and automated processing are in place. Remaining issues are integration of beam information, detector configuration and calibrations into the main processing stream, and faster tape access.

24 1.3.5 Monte Carlo Challenges

The collaboration has performed multiple Monte Carlo challenges to create samples for physics studies for the Technical Proposal and in preparation for the Technical Design Report in early 2019. In the last major challenge, MCC10 in early 2018, 17M events, taking up 252 TB of space were generated and catalogued automatically using the central DUNE production framework in response to requests by the Physics groups.

30 1.3.6 Reconstruction tests

Reconstruction tests have been performed on simulated single-phase ProtoDUNE test beam interactions with cosmic rays and an electronic noise simulation based on MicroBooNE experience. Hit finding and shaping is found to take around 2 minutes/event with a 2 GB memory footprint, leading to a reduction in data size of a factor of four. Higher-level pattern recognition occupies 10-20 minutes/event with a 4-6 GB memory footprint. For real data, calibration, electric field
 non-uniformities and other factors will likely raise the CPU needs per event. We will learn this
 when real data starts to arrive in late summer.

1.4 Resource planning and prospects

The DUNE computing effort relies heavily on the human and hardware resources of multiple 5 organizations, with the bulk of hardware resources at CERN, and national facilities worldwide. 6 The DUNE computing organization serves as an interface between the collaboration and those 7 organizations. Computational resources are currently being negotiated on a yearly basis, with 8 additional resources available opportunistically. Human assistance is largely on a per-project 9 basis, with substantial support when needed but very few personnel as yet permanently assigned 10 to the DUNE or ProtoDUNE efforts. We are working with the laboratories and funding agencies to 11 identify and solidify multi-year commitments of dedicated personnel and resources for ProtoDUNE 12 and DUNE, analogous to, but smaller than, those assigned to the LHC experiments. In-kind 13 contributions of computing resources and people can also be an alternative way for institutions to 14 make substantial contributions to DUNE. 15

The ProtoDUNE efforts in 2018-2019 will exercise almost all computing aspects of DUNE, although at smaller scale. Much of the infrastructure needed for full DUNE, in particular databases, grid configurations and code management systems need to be fully operational for ProtoDUNE. We believe that the systems in place (and tested) will be adequate for that purpose.

However, ProtoDUNE represents only 4-5% of the final volume of the far detectors and the near de-20 tector technology is, as yet, unknown. At the same time, computing technology is evolving rapidly 21 with increased need for flexibility and the ability to parallelize codes. Liquid Argon detectors, be-22 cause of their reasonably simple geometry and image-like data, are already able to take advantage 23 of parallelization and generic machine learning techniques. We have good common infrastructure 24 such as the LArSoft suite and geant4, and will have an excellent testbed with the ProtoDUNE 25 data, but our techniques will need substantial adaption to scale to full DUNE and to take full 26 advantage of new architectures. This will be one of our major challenges, and opportunities for 27 collaboration, over the next five years. 28

¹ Glossary

| 2 | \boldsymbol{art} A software framework implementing an event based execution paradigm. 6 |
|----------------|---|
| 3 4 | detector module The entire DUNE far detector is segmented into four modules, each with a nominal 10 kt fiducial mass. 14, 15 |
| 5 6 | secondary DAQ buffer A secondary DAQ buffer holds a small subset of the full rate as selected by a trigger command. This buffer also marks the interface with the DUNE Offline. 14 |
| 7 | DUNE Deep Underground Neutrino Experiment. 1, 3, 6, 7, 9, 12, 13 |
| 8 9 10 | anode plane assembly (APA) One unit the SP detector containing the elements sensitive to activity in the LAr. It contains two faces each of three planes of wires, cold electronics and photo detection system. 2 |
| 11 12 13 | data acquisition (DAQ) The data acquisition system accepts data from the detector FE electron- ics, buffers it, performs a trigger decision, builds events from the selected data and delivers the result to the offline secondary DAQ buffer. 1 |
| 14 15 | dual-phase (DP) Distinguishes one of the four 10 kt detector modules of the DUNE far detector by the fact that it operates using argon in both gas and liquid phases iii, 15 |
| 16 | Experiment Hall North One (EHN1) Location at CERN of the ProtoDUNE experiments. 11 |
| 17 | EOS (EOS) The XRootD based distributed file system developed by CERN. 11 |
| 18 19 | far detector (FD) Refers to the detector or more generally the experimental site in or above the Homestake mine in Lead, SD. 5 |
| 20 | liquid argon (LAr) The liquid phase of argon. 1 |
| 21 22 | Long-Baseline Neutrino Facility (LBNF) The organizational entity responsible for developing the neutrino beam used by DUNE. 2 |

Sequential Access via Metadata (SAM) A data handling system to store and retrieve files and
 associated metadata, including a complete record of the processing which has used the files.

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- single-phase (SP) Distinguishes one of the four 10 kt detector modules of the DUNE far detector
 by the fact that it operates using argon in just its liquid phase.. iii, 2, 15
- **HPC** High Performance Computing Facilities generally computing facilities emphasizing parallel
 computing with aggregate power of more than a teraflop.. 7
- LArSoft Liquid Argon Software (LArSoft), a shared base of physics software across Liquid Argon
 (LAr) Time Projection Chamber (TPC) experiments.. 6–8, 13
- MicroBooNE The Liquid Argon TPC-based MicroBooNE neutrino oscillation experiment at Fermilab. 3, 5, 6, 9, 12
- ¹⁰ MINERvA The MINERvA neutrino cross sections experiment at Fermilab. 3, 9
- **NOvA** The NOvA off-axis neutrino oscillation experiment at Fermilab. 6, 9
- ProtoDUNE Two prototype detectors operated in a CERN beam test. One prototyping single phase (SP) and the other dual-phase (DP) technology. i, 1, 3, 5–13
- ¹⁴ **ROI** region of interest. 2
- trigger candidate Summary information derived from the full data stream and representing a
 contribution toward forming a trigger decision. 15
- trigger command Information derived from one or more trigger candidates and which directs
 elements of the detector module to read out a portion of the data stream. 14, 15
- trigger decision The process by which trigger candidates are converted into trigger commands.
 14, 15

Acronyms

- ² **APA** anode plane assembly. 2, 10
- ³ **DAQ** data acquisition. 1, 3, 5, 8, 14
- ⁴ **DP** dual-phase. iii, 4, 5, 10, 15
- ⁵ **EHN1** Experiment Hall North One. 11
- 6 **EOS** EOS. 11
- $_7$ $\,$ FD far detector. 5 $\,$
- **LAr** liquid argon. 1
- ⁹ **LBNF** Long-Baseline Neutrino Facility. 2
- 10 **SAM** Sequential Access via Metadata. 6
- ¹¹ **SNB** supernova neutrino burst. 4
- ¹² **SP** single-phase. iii, 2, 4, 5, 10, 15

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