

Introduction to the COVID-19 Open Research Dataset

The COVID-19 Open Research Dataset (CORD-19) is a collection of over 50,000 scholarly articles - including over 40,000 with full text - about COVID-19, SARS-CoV-2, and related coronaviruses. This dataset has been made freely available with the goal to aid research communities combat the COVID-19 pandemic. It has been made available by the Allen Institute for AI in partnership with leading research groups to prepare and distribute the COVID-19 Open Research Dataset (CORD-19), in response to the COVID-19 pandemic.

During this lab you will learn how to process and analyze a subset of the articles present in the dataset, group them together into a series of clusters, and use Automated ML to train a machine learning model capable of classifying new articles as they are published.

Setup

We will start off by installing a few packages, such as `nltk` for text processing and `wordcloud`, `seaborn`, and `yellowbrick` for various visualizations.

```
In [1]: %pip install nltk
        %pip install wordcloud
        %pip install seaborn
        %pip install yellowbrick
```

Requirement already satisfied: nltk in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (3.7)

Requirement already satisfied: joblib in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from nltk) (1.1.0)

Requirement already satisfied: regex<=2021.8.3 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from nltk) (2022.3.15)

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Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: wordcloud in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (1.8.1)

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Requirement already satisfied: numpy>=1.6.1 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from wordcloud) (1.22.3)

Requirement already satisfied: pillow in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from wordcloud) (9.1.0)

Requirement already satisfied: fonttools>=4.22.0 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from matplotlib->wordcloud) (4.32.0)

Requirement already satisfied: python-dateutil>=2.7 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from matplotlib->wordcloud) (2.8.2)

Requirement already satisfied: cycler>=0.10 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from matplotlib->wordcloud) (0.11.0)

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Requirement already satisfied: pyparsing>=2.2.1 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from matplotlib->wordcloud) (3.0.8)

Requirement already satisfied: six>=1.5 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from python-dateutil>=2.7->matplotlib->wordcloud) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: seaborn in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (0.11.2)

Requirement already satisfied: matplotlib>=2.2 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from seaborn) (3.5.1)

Requirement already satisfied: pandas>=0.23 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from seaborn) (1.4.2)

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Requirement already satisfied: six>=1.5 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from python-dateutil>=2.7->matplotlib>=2.2->seaborn) (1.16.0)
 Note: you may need to restart the kernel to use updated packages.
 Requirement already satisfied: yellowbrick in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (1.4)
 Requirement already satisfied: scipy>=1.0.0 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from yellowbrick) (1.8.0)
 Requirement already satisfied: matplotlib!=3.0.0,>=2.0.2 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from yellowbrick) (3.5.1)
 Requirement already satisfied: numpy>=1.16.0 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from yellowbrick) (1.22.3)
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 Requirement already satisfied: kiwisolver>=1.0.1 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from matplotlib!=3.0.0,>=2.0.2->yellowbrick) (1.4.2)
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 Requirement already satisfied: threadpoolctl>=2.0.0 in /anaconda/envs/aiw-ai-kernel/lib/python3.8/site-packages (from scikit-learn>=1.0.0->yellowbrick) (3.1.0)
 Note: you may need to restart the kernel to use updated packages.

We'll first download stopwords and the Punkt tokenizer models present in the `nltk` package, in order to be able to process the articles

In [2]: `import nltk`

```
nltk.download('punkt')
nltk.download('stopwords')
```

```
[nltk_data] Downloading package punkt to /home/azureuser/nltk_data...
[nltk_data]   Package punkt is already up-to-date!
[nltk_data] Downloading package stopwords to
[nltk_data]   /home/azureuser/nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
```

Out[2]: `True`

We'll also import the rest of the modules needed in this notebook, and do a quick sanity-check on the Azure ML SDK version

In [3]: `import os`
`import json`
`from string import punctuation`

```

import pandas as pd
import seaborn as sns
sns.set_palette('Set2')
from wordcloud import WordCloud
import matplotlib.pyplot as plt
from yellowbrick.cluster import KElbowVisualizer
from gensim.models.doc2vec import Doc2Vec, TaggedDocument
from sklearn.manifold import TSNE
from sklearn.decomposition import PCA
from sklearn.model_selection import train_test_split
from sklearn.cluster import KMeans, SpectralClustering, DBSCAN, Birch, AgglomerativeClustering
from sklearn.metrics import roc_auc_score
from nltk import word_tokenize, sent_tokenize
from nltk.stem import SnowballStemmer, PorterStemmer

from azureml.core import Workspace, Datastore, Dataset, VERSION

print("Azure ML SDK Version: ", VERSION)

```

Azure ML SDK Version: 1.39.0

Load the Covid-19 data

CORD-19 has been uploaded to an Azure Storage Account, we will connect to it and use it's API to download the dataset locally. Also please make sure to update the storage account name from the labguide in the below cell.

```

In [4]: covid_dirname = 'covid19temp'

cord19_dataset = Dataset.File.from_files('https://aiinadaystorage800633.blob.core.winc
mount = cord19_dataset.mount()

covid_dirpath = os.path.join(mount.mount_point, covid_dirname)

```

Display a sample of the dataset (top 5 rows).

```

In [5]: mount.start()

# Load metadata.csv, as it contains a list of all the articles and their corresponding
metadata_filename = os.path.join(covid_dirpath, 'metadata.csv')

metadata = pd.read_csv(metadata_filename)
metadata.sample(5)

/tmp/ipykernel_5158/188001440.py:6: DtypeWarning: Columns (13,14) have mixed types. S
pecify dtype option on import or set low_memory=False.
    metadata = pd.read_csv(metadata_filename)

```

Out[5]:

	cord_uid	sha	source_x	title	
48221	qfcoolj9	NaN	Medline	Middle East respiratory syndrome coronavirus i...	10.1128/
113519	pu1jci26	9deb7792c0ff4705670a156adddb64a411384bde	Medline; PMC	Using echocardiography to guide the treatment ...	10.1180 020
117707	nxoj330c	NaN	Medline; WHO	Déjà Vu or Jamais Vu? How the Severe Acute Res...	10.2214/ajr
43435	1r6eg1hh	NaN	Medline	[Rapid review of the use of community-wide sur...	10.211
79284	znafueuo	NaN	Medline	Human internal jugular valve M- mode ultrasound...	

Some of the articles do not have any associated documents, so we will filter those out.

```
In [6]: metadata_with_docs = metadata[metadata['pdf_json_files'].isna() == False]
print(f'Dataset contains {metadata.shape[0]} entries, out of which {metadata_with_docs
```

Dataset contains 134206 entries, out of which 56962 have associated json documents

Display the percentage of items in the dataset that have associated JSON documents (research papers that have extra metadata associated with them).

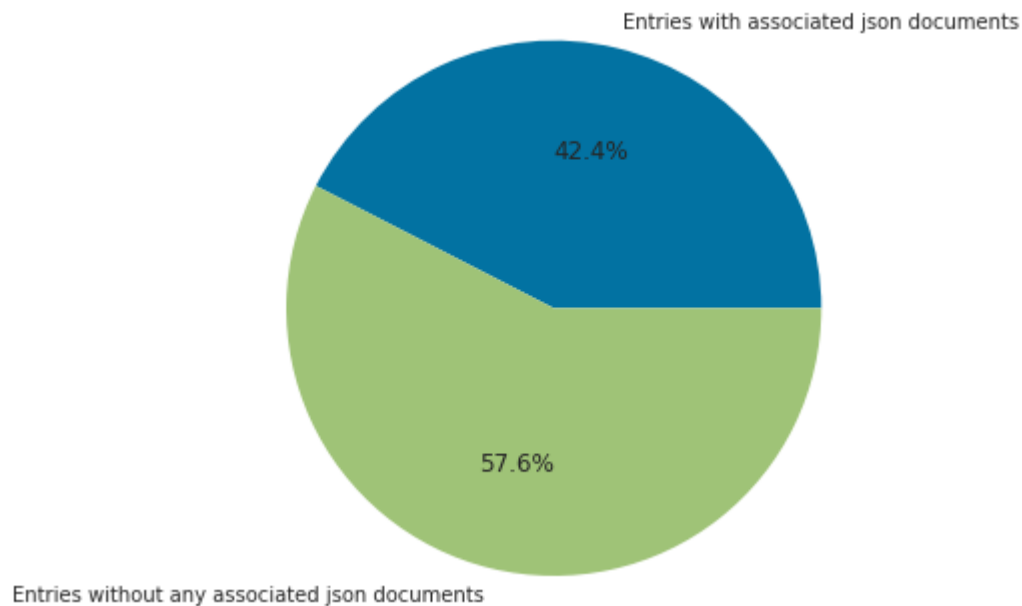
```
In [7]: doc_counts = [metadata_with_docs.shape[0], metadata.shape[0] - metadata_with_docs.shap
doc_labels = ['Entries with associated json documents', 'Entries without any associate

fig, ax = plt.subplots()
ax.pie(doc_counts, labels=doc_labels, autopct='%1.1f%%')
ax.axis('equal')
plt.show()
```

```

findfont: Font family ['sans-serif'] not found. Falling back to DejaVu Sans.
findfont: Generic family 'sans-serif' not found because none of the following familie
s were found: Arial, Liberation Sans, Bitstream Vera Sans, sans-serif
findfont: Font family ['sans-serif'] not found. Falling back to DejaVu Sans.
findfont: Generic family 'sans-serif' not found because none of the following familie
s were found: Arial, Liberation Sans, Bitstream Vera Sans, sans-serif

```



Investigate individual items

Let's load an example entry from the dataset.

```

In [8]: # Change the document index in order to preview a different article
DOCUMENT_INDEX = 0
example_entry = metadata_with_docs.iloc[DOCUMENT_INDEX]

filepath = os.path.join(covid_dirpath, example_entry['pdf_json_files'])
print(f'Document local filepath: {filepath}')
print(f'Document local filepath: {covid_dirpath}')

filepath = covid_dirpath + '/comm_use_subset/pdf_json/02a009e42054081b441d0f4b203679c4
print(f'Document local filepath: {filepath}')

```

```

Document local filepath: /tmp/tmpemaxbzly/covid19temp/document_parsing/pdf_json/d1aafb
70c066a2068b02786f8929fd9c900897fb.json
Document local filepath: /tmp/tmpemaxbzly/covid19temp
Document local filepath: /tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/02a009
e42054081b441d0f4b203679c4b0cae38d.json

```

Next, we will display the list of elements that are available for the selected document.

```

In [9]: try:
        with open(filepath, 'r') as f:
            data = json.load(f)

    except FileNotFoundError as e:
        # in case the mount context has been closed
        mount.start()
        with open(filepath, 'r') as f:

```

```
data = json.load(f)

print(f'Data elements: { ", ".join(data.keys())}' )
```

Data elements: paper_id, metadata, abstract, body_text, bib_entries, ref_entries, back_matter

Please make sure to update the storage account name from the labguide in the below cell.

```
In [10]: from azureml.core import Dataset
cord19_dataset = Dataset.File.from_files('https://aiinadaystorage800633.blob.core.winc
mount = cord19_dataset.mount()
```

View the full text version of the document.

```
In [11]: for p in data['body_text']:
          print(p['text'], '\n')
```

Hantaviruses are negative-sense RNA viruses transmitted to humans from small animal hosts. Different viral species are associated with one of two disease syndromes: hemorrhagic fever with renal syndrome (HFRS), or hantavirus pulmonary syndrome (HPS) [1]. Hantaan virus a1111111111 a1111111111 a1111111111 a1111111111 a1111111111 (HTNV), primarily found in Asia, is among the most prevalent HFRS-causing hantaviruses with a case fatality rate of between 1-15% [2]. Puumala virus (PUUV) causes most HFRS cases in Europe, though its case fatality rate is lower at <1% [3, 4]. There are currently no FDA licensed vaccines or therapeutics for either HFRS or HPS [5].

The Syrian hamster (*Mesocricetus auratus*) is the typical animal used to model hantavirus infection and disease. Andes virus (ANDV), an HPS-causing hantavirus, causes lethal disease in immunocompetent hamsters [6], while numerous other HPS-causing hantaviruses including Sin Nombre Virus (SNV) and Choclo virus cause lethal disease in hamsters immunosuppressed with dexamethasone and cyclophosphamide [7, 8]. In contrast to HPS-causing hantaviruses, exposure of hamsters to HFRS-causing hantaviruses such as HTNV, PUUV, Dobrava (DOBV) and Seoul (SEOV) leads to asymptomatic infection, despite viral dissemination, even when immunosuppressed (Hooper Lab, unpublished data) [8] [9] [10] [11]. In these studies hamsters were exposed to high doses of HTNV and PUUV, far exceeding the infectious dose 99% (ID₉₉) for the virus. Development and characterization of a uniformly infective, low-dose challenge model, enhances the hamster model's usefulness in vaccine and therapeutic testing. In this report we present a low-dose hamster infection model for both HTNV and PUUV infected animals.

Ferrets (*Mustela putorius furo*) have become a popular animal model for a number of respiratory pathogens including influenza [12], coronavirus [13], Nipah virus [14], and morbillivirus [15], due to the similarity in lung physiology to humans. In addition, they have recently been described as a disease model of two hemorrhagic fever viruses, Bundibugyo virus and Ebola virus [16, 17], supporting viral replication without prior adaptation. Most hantavirus-related human disease occurs by aerosolized transmission of the virus from the excreta or secreta of infected rodents [18, 19], a model of viral infection for which the ferret is well suited. In this study we demonstrate that ferrets are capable of being infected by high titers of HTNV and PUUV, though aside from gradual weight loss infected animals exhibit no clinical symptoms or impaired renal function.

It has been established that infection of rhesus macaques (*Macaca mulatta*) with HFRS-causing hantaviruses (DOBV, SEOV, HTNV, and PUUV) leads to asymptomatic infection and seroconversion [9], while infection of cynomolgus macaques (*Macaca fascicularis*) with PUUV leads to a mild disease characterized by lethargy, mild proteinuria and hematuria, and kidney pathology, similar to mild HFRS in humans [20]. However, the macaques' large size and cost limits their usefulness in therapeutic studies, especially when test article availability is limited, as is often the case in passive transfer studies. The common marmoset (*Callithrix jacchus*) is becoming more popular for infectious disease studies. Its genetic similarity to humans, cost, relative safety, and small size make it an attractive alternative to traditional non-human primate species [21]. Marmosets have been used as a disease model for other viral agents including Dengue virus [22], Hepatitis C virus [23], influenza virus [24], Lassa fever virus [25], orthopox viruses [26] [27] [28], Rift Valley Fever virus [29], Eastern Equine Encephalitis virus [30], and filoviruses [31]. In this study we demonstrate that exposure of marmosets to HTNV leads to asymptomatic infection characterized by high levels of neutralizing antibodies. This is the first report of hantavirus infection in marmosets.

Medical countermeasures are products including biologics (e.g., vaccines and antibodies) and small molecule drugs that can be used to prevent or combat infectious disease outbreaks. This study presents three animal models of HTNV infection, and two models of PUUV infection that can be used to evaluate the efficacy of medical countermeasures that are intended to prevent or mitigate infection (e.g., vaccines) by these viruses through induction of sterile immunity.

HTNV strain 76-118 [32] , PUUV strain K27 [33] , and PUUV strains Beaumont, and Seloignes (gifts of Piet Maes, Leuven, Belgium) were propagated in Vero E6 cells (Vero C1 008, ATCC CRL 1586) in T-150 flasks and cEMEM media (Eagle's minimal essential medium with Earle's salts (EMEM) containing 10% heat inactivated fetal bovine serum, 200 μ M glutamine, 1% non-essential amino acids, 10 mM HEPES pH 7.4; and antibiotics [penicillin (100 U/ml), amphotericin B (250 μ g/ml), and gentamicin (50 mg/ml)]. Virus was collected from infected monolayer supernatants. Cell debris was removed by low speed centrifugation (2500 rpm in a table top centrifuge). HTNV and PUUV strain K27 were twice plaque purified according to published methods [6] . Virus stocks were aliquoted and stored at -60°C or colder. Virus identity has been confirmed by sequencing of the stocks.

Female Syrian hamsters 6-8 wks of age (Envigo, Indianapolis, IN) were anesthetized by inhalation of vaporized isoflurane using an IMPAC 6 veterinary anesthesia machine. Once anesthetized, animals were injected with the indicated concentration of virus diluted in PBS. Intramuscular (i.m.) injections (in the caudal thigh) consisted of 0.2 ml delivered with a 1ml syringe with a 25-gauge, 5/8 in needle. Intranasal (i.n.) instillation consisted of 50 μ l total volume delivered as 25 μ l per nare with a plastic pipette tip. Blood sampling from the vena cava occurred under previously stated methods of anesthesia, and was limited to 7% of a hamster's total blood volume per week. At time of arrival animals were randomized into experimental groups. Animals were housed in small animal pans, not exceeding four animals to a pan, in a climate and humidity controlled animal biosafety level 3 (ABSL-3) with a 12-hour light/dark cycle. Animals had pelleted food and water provided ad libitum. Enrichment in the form of toys, nesting material and supplemental treats was provided. Humane endpoint conditions were established as decreased mobility (inability to obtain food and water) and subdued response to stimulation, and animals were monitored daily during the experiment. As infected animals did not become ill, these criteria were not met and animals were euthanized by terminal blood collection from the heart after administration of Ketamine-xylazine (KAX)(USAMRIID, Fort Detrick MD) and prior to intracardiac injection with pentobarbital sodium (USAMRIID) at the end of the study. Due to lack of illness no pain relief, aside from anesthesia during procedures, was required.

confidence intervals reaching out to 0.7log 10 (approximately 5-fold) under monotone assumptions of response profiles for intermediate doses. As our initial selection of dosages did not meet the desired infection rate span, some dosage groups were repeated leading to 20 hamsters per group. In the serial euthanasia study three hamsters per group were used. This is the minimum number required to provide collection of sufficient samples for detection of antibodies and viral kinetics in tissues. For each experiment pre-sera from animals served as negative control.

Adult, female neutered and descented ferrets (Marshall Farms, North Rose, NY), were anesthetized by inhalation of vaporized isoflurane using an IMPAC 6 veterinary anesthesia machine, or i.m. injection of Telazol (Zoetis, Parsippany, NJ). Injections (i.m. and i.n.) and blood sampling were conducted under the same condition as hamsters. Intraperitoneal (i.p.) injections consisted of 1 ml delivered with a 3 mL syringe and a 23-gauge needle. Microchips (BMDS, Seaford, DE) were used to identify and ascertain temperature during ferret experiments. In the first ferret challenge study faulty chips lead to inaccurate temperature readings and were only used for identification purposes. Animals were randomized upon receipt into experimental groups. Ferrets were socially housed in metal caging, two to a cage, with sight lines to additional animals in the study, in a climate and humidity controlled ABSL-3 with 12/hour light and dark cycles. Each cage had a nesting box with bedding material, and numerous tubes and shelves for play. Ferrets had access to pelleted food supplemented with treats and potable water, through an automated watering system. Enrichment in the form of manipulanda (tubes, balls, mirrors) and food was provided. Animals were observed daily by trained personnel in addition to general husbandry assessments. Humane euthanasia criteria, defined as both dyspnea, loss of mobility (to obtain food and water) and >20% weight loss. At the end of the experiment, terminal blood samples were collected from the heart after administration of KAX and prior to intracardiac injection with pentobarbital so

dium. No pain relief, aside from anesthesia during procedures, was used.

When disease occurs independently in each of four ferrets with 50% probability, the experiment will have odds about 9:1 in favor of producing at least one diseased ferret. Conversely failure to observe any diseased ferret in a group of four will yield a 95% confidence interval extending from 0-50%. That is, with 95% confidence it will be admitted that the true disease rate may be 50% or less. For this reason groups of four ferrets were used for the experiments. For each experiment pre-sera from animals served as a negative control.

Adult marmosets weighing over 300g were anesthetized by inhalation of vaporized isoflurane using an IMPAC 6 veterinary anesthesia machine. Once anesthetized, animals were injected with the indicated concentration of virus diluted in PBS. I.m. injections (in the caudal thigh) consisted of 0.2 ml delivered with a 1ml syringe with a 25-gauge, 5/8 in needle. Blood sampling from a femoral vein occurred under previously stated methods of anesthesia, and were limited to 7% of each marmoset's total blood volume per week. At time of euthanasia, terminal blood samples were collected from the heart after anesthetization by i.m. injection of Telazol and prior to intracardiac administration of pentobarbital sodium. Animals were housed in containment as previously described [26]. In brief, animals were singly housed in metal cages meeting current standards in a climate and humidity controlled room. Animals were fed pelleted food supplemented with fruits and treats daily, and were provided potable water through an automatic watering system ad libitum. Enrichment in the form of manipulanda (i.e. toys, metal mirrors), foraging devices, treats, and fruit were provided daily. Animals were observed daily by trained personnel in addition to general husbandry assessments. Animals were observed daily by trained personnel in addition to general husbandry assessments. Animals found moribund (defined as labored breathing, decreased food consumption, persistent prostration and moderate unresponsiveness) would be euthanized under humane endpoint criteria, however, as animals did not become ill during the study this criteria was not met. No additional pain relief, aside from anesthesia during procedures, was necessary. All work was performed in an ABSL-3 laboratory.

The marmoset study requires a sample size of 3 for adequate power to determine if the incidence of seroconversion is significantly greater than that which would be expected in the population. This sample size will allow the experimenter to detect seroconversion in at least 2 of 3 animals (66%) versus the expected population constant of <1% at a 95% confidence level using a one-tailed binomial test for proportions.

The enzyme-linked immunosorbent assay (ELISA) used to detect nucleocapsid protein (N) specific antibodies (N-ELISA) was described previously [34]. Species-specific secondary antibodies were used at the following concentrations: peroxide-labeled anti-hamster (1:10,000) (Sera Care, Gaithersburg, MD), peroxide-labeled anti-ferret (1:5,000) (Sigma Aldrich, St. Louis, MO), and alkaline phosphatase conjugated anti-monkey (1:1,000) (MilliporeSigma, St. Louis, MO). Assays using peroxide labeled antibodies were developed with TMB microwell peroxidase substrate (Sera Care) at an absorbance of 450 nm; assays using alkaline phosphatase conjugated antibodies were developed with p-nitrophenyl phosphate (PNPP) (ThermoFisher Scientific, Waltham, MA) at 405 nm. A sample was considered positive if its peak optical density (OD) value was greater than either 0.025 or the background value (the average of three negative controls + 3 times the standard deviation), whichever was higher. The specific OD sum is the summation of all values greater than background and represents the area under the titer curve.

PRNT assays were performed as previously described with minor modifications [35-37]. HTNV-infected monolayers were fixed 7 days post-infection, while PUUV-infected monolayers were fixed 10 days post infection by 2 mL of 10% formalin per well. Immunostaining was performed as previously described [38]. All sera samples were assayed in duplicate beginning at a 1:20 final dilution. PRNT 50 values represent the reciprocal dilution at which the serum neutralizes 50% of the virus.

Approximately 200 mg of organ tissue was homogenized in 1.0 mL of TRIzol (ThermoFisher

r, Waltham, MA) reagent using M tubes on the gentleMACS (Miltenyi Biotec, Auburn, CA) dissociation system on the RNA setting. RNA was extracted from TRIzol per manufacturer's protocol. A Nanodrop 8000 was used to determine RNA concentration, which was then raised to either 100 ng/μL or 1,000 ng/μL in UltraPure distilled water (ThermoFisher). Real-time PCR was conducted on a BioRad CFX thermal cycler using either an Invitrogen Power SYBR Green RNA-to-Ct one-step kit (ThermoFisher) or Brilliant II QRT-PCR 1-Step Master Mix (Agilent, Santa Clara, CA) according to the manufacturer's protocol. For spiked assays the master mix was spiked with either HTNV or PUUV viral RNA prior to addition to the samples. For HTNV, primer sequences were 594F 5'-AAG CAT GAA GGC AGA AGA GAT -3' and

T-25 flasks of one week old Vero E6 cells were infected with 50 μL of urine plus an additional 450 μL of cMEM media. After a 1 hr adsorption at 37°C with 5% CO₂, the volume was raised to 3.5 mL. On Day 4 post infection supernatant was collected and frozen down, 500 μL of which was used to infect fresh Vero E6 cells at a later time point. After a 1 hr adsorption at 37°C with 5% CO₂, the volume was raised to 3.5 mL. On days 7, 11, 14, 17, 21 and 28 1.2 mL of culture supernatant was collected and frozen down. The volume of cMEM in the flask was raised to 3.5 mL with fresh media.

Approximately 200 μg of organ tissue were homogenized in 1 mL of cMEM media using M tubes on the gentleMACS dissociation system on the RNA setting. Plaque assays using urine, sera, or organ homogenate were then performed beginning at the 1:10 dilution as described in [6, 41] with minor modifications. For spiked plaque assays the protocol was identical except for equivalent amounts of virus being spiked into either media alone (control), or the 1:10-1:1,000 dilution of organ homogenate. HTNV-infected monolayers were fixed 7 days postinfection, while PUUV-infected monolayers were fixed 10 days post infection by 2 mL of 10% formalin per well. Immunostaining was performed as previously described [38].

Following euthanasia, necropsies were performed. Samples were collected aseptically for the virology studies described above. For the hamsters and ferrets, samples of the following were collected: heart, lung, liver, spleen, kidney, brain, and urine. In addition, ferrets had samples of intestine, adrenal gland, pituitary gland, and eye were collected. Samples of the following were collected from the marmosets: heart, lung, liver, spleen, kidney, intestine, and brain. After the virology samples had been collected, all major internal organs in each animal were also sampled for histology.

Tissues were fixed in 10% neutral-buffered formalin for 21 days. Tissues were then trimmed, processed under vacuum through increasing concentrations of alcohols, and embedded in paraffin. Paraffin embedded tissue sections of 5-6 μm were cut and mounted on glass slides, stained with hematoxylin and eosin (H&E), and mounted under a glass coverslip for routine histologic evaluation. The paraffin-embedded tissues used for producing the H&E-stained histology slides were also utilized for immunohistochemistry (IHC) studies in the hamsters. Immunolocalization of HTNV in tissues was performed with an immunoperoxidase procedure (horseradish peroxidase EnVision system; Dako) according to the manufacturer's directions. The primary antibody was an anti-HTNV nucleocapsid rabbit polyclonal antibody diluted 1:3,500 (ferret) or 1:5,000 (hamster) (BEI Resources, Manassas, VA). After deparaffinization and peroxidase blocking, tissue sections were pretreated with proteinase K for 6 min at room temperature, rinsed, and then covered with primary antibody and incubated at room temperature for 30 min. They were rinsed, and then the peroxidase-labeled polymer (secondary antibody) was applied for 30 min. Slides were rinsed, and a substrate-chromogen solution (3,3'-diaminobenzidine; Dako, Santa Clara, CA) was applied for 5 min. The substrate-chromogen solution was rinsed off the slides, and the slides were stained with hematoxylin and rinsed. The sections were dehydrated and cleared with xylene, and then a coverslip was placed.

On the indicated days, anesthetized ferrets were injected i.p. with water soluble Cyp (Baxter Health Care Corporation, Deerfield, IL) with the indicated dosages per body weight of drug diluted in sterile phosphate-buffered saline (PBS), pH 7.4. In the first experiment ferrets were administered a loading dose of 30 mg/kg on Day 41 post infection.

ction, with maintenance doses of 30 mg/kg administered every other day until euthanasia. In the second experiment, ferrets were administered a loading dose of 30 mg/kg on Day -1, and a maintenance dose of 10 mg/kg on Day 1, 3, 11, and 13. Administration of Cyp was discontinued between days 3 and 11 due to secondary infection. To combat the infection (rapid onset of fever and weight loss), ferrets were treated with 5mg/kg i.m. enrofloxacin (Norbrook Laboratories, Overland Park, KS) twice daily per veterinarian instructions starting on Day 4. Beginning on Day 11 with the resumption of immunosuppression ferrets were treated prophylactically with 10 mg/kg i.m. enrofloxacin once daily).

Blood samples of 0.5 mL were collected in lithium heparin capillary blood collection tubes and analyzed using an Advia 120 hematology analyzer (Software version 3.1.8.0-MS). Per manufacturer's recommendation, the dog setting was used for the complete blood count (CBC) and the guinea pig setting was used for the white blood cell differential (WBC) in hamsters. For ferrets, the dog setting was used for both the CBC and WBC.

The PsVNA used to detect neutralizing antibodies in sera was described previously [42, 43]. This is a replication-restricted, recombinant vesicular stomatitis virus (rVSV Δ G) expressing luciferase, which is pseudotyped with the Hantaan glycoprotein. First, heat-inactivated sera was diluted 1:10, followed by five-fold serial dilutions that were mixed with an equal volume of Eagle's minimum essential medium with Earle's salts and 10% fetal bovine sera containing 4000 fluorescent focus units of Hantaan pseudovirions. This mixture was incubated overnight at 4°C. Following this incubation, 50 μ l was inoculated onto Vero cell monolayers in a clear bottom, black-walled 96-well plate in duplicate. Plates were incubated at 37°C for 18-24 hr. The media was discarded and cells were lysed according to the luciferase kit protocol (Promega, Madison, WI). A Tecan M200 Pro was used to acquire luciferase data. The values were graphed using GraphPad Prism (version 7) and used to calculate the percent neutralization normalized to cells alone and pseudovirions alone as the minimum and maximum signals, respectively. The percent neutralization values for duplicate serial dilutions were plotted. Fifty percent PsVNA (PsVNA 50) titers were interpolated from 4-parameter curves, and geometric mean titers were calculated.

Blood was collected in serum separator tube, and spun at 500x g to isolate sera. Ferret sera was analyzed on the Piccolo comprehensive metabolic panel, and marmoset sera was analyzed on the Piccolo general chemistry 13 panel per manufacturer's instructions (Abaxis Global Diagnostics, Union City, CA).

Urine was expressed from anesthetized ferrets and analyzed by urinalysis reagent strips (VWR, Radnor, PA).

A Bayesian probit model was used to estimate 95% highest posterior density intervals for a 50% and 95% infectious dose calculation. Student's t-test and Mann-Whitney tests were used to compare white blood cell levels pre- and post-Cyp administration. P-values of <0.05 were considered significant. Standard deviation of data was assessed to ensure data was normally distributed prior to use of Student's t-test. Analyses were conducted using GraphPad Prism (version 7); Bayesian analyses were performed using SAS.

We have previously demonstrated that Syrian hamsters are capable of being infected by HFRS-causing hantaviruses, but they do not develop any signs of clinical disease [9]. To develop standard models of HTNV and PUUV infection for future evaluation of vaccines and medical countermeasures, groups of between 10 and 20 hamsters were exposed to serial ten-fold dilutions of either HTNV or PUUV by either the i.m. or i.n. route (from 2-20,000 PFU HTNV or 0.2-20,000 PFU PUUV). Between 28-35 days post infection, hamsters were terminally bled with infection status monitored by N-ELISA titers (Fig 1). From these data the doses required to infect 50% (ID 50) and 99% (ID 99) were calculated (Table 1).

To further characterize a low-dose standard hamster infection model for HFRS-causing

hantaviruses a hamster serial sacrifice study was performed. Hamsters were challenged with either 10 PFU (~3 ID₅₀) HTNV i.m., 500 PFU (~1.5 ID₅₀) HTNV i.n., 1,000 PFU (~1 ID₅₀) PUUV i.m., or 1,000 PFU (~1.5 ID₅₀) PUUV i.n. On various days post infection, groups of three hamsters were euthanized to monitor viral and serological parameters.

Seroconversion occurred, at least partially, by Day 17 post HTNV infection and Day 24 post PUUV infection. Seroconversion on Day 28 post HTNV i.m. infection was incomplete, though viral genome was recovered from all hamsters euthanized that day indicating a productive HTNV infection occurred (Fig 2). To confirm seroconversion, all hamsters euthanized on days 17, 24, and 28 were assayed for neutralizing antibodies by PRNT regardless of N-ELISA seroconversion status. All HTNV infected hamsters with N-ELISA titers had neutralizing antibodies as measured by PRNT 50, with all but one having full neutralization of virus at a 1:20 dilution of sera. (S1A and S1B Fig). Similarly, all PUUV hamsters with N-ELISA titers had neutralizing antibodies as measured by PRNT 50, though three of the five PUUV i.n. challenged hamsters did not have complete neutralization of the virus at a 1:20 dilution of sera (S1C and S1D Fig). Two of three PUUV i.m. challenged hamsters that had not seroconverted on Day 17 post infection had low levels of neutralizing antibodies, while none of the PUUV i.n. challenged hamsters that were seronegative by N-ELISA had neutralizing antibodies. For both HTNV and PUUV infected hamsters, infection via the i.m. route lead to a more robust neutralizing antibody response than the i.n. route.

The kinetics of HTNV and PUUV infection in the heart, lung, liver, spleen, kidney, and brain were monitored by both RT-PCR and plaque assay (Figs 3 and 4). HTNV infected by either the i.m. or i.n. route resulted in a persistent infection. High levels of viral genome were detected in the heart, lung, kidney and brain of HTNV infected hamsters beginning on either Day 11 or 17 post infection (Fig 3A, 3B, 3E and 3F). Infection of the kidney and brain was found in all examined hamsters beginning on either Day 17 or 24 post infection, while the high titers of viral genome detected in the heart and lung were present in only one or two hamsters at each time point. Low levels of viral genome were detected between days 11 and 24 post infection in the liver of HTNV i.m. but not i.n. infected hamsters (Fig 3C). Hardly any viral genome was detected in the spleen (Fig 3D). To confirm the lack of viral genome in the spleen was not due to the presence of inhibitors all spleen samples from i.m. and i.n. infected hamsters were spiked with HTNV prior to RT-PCR. No significant inhibition of the spiked RNA was noted, indicating that HTNV infection does not result in virus dissemination to the spleen (S2A and S2B Fig).

The detection of PUUV in organs was transient after i.m. infection, occurring between Day 11-17 for the heart, liver, kidney and brain, with no virus detected in the spleen (Fig 3A, 3C, 3E and 3F). Viral genome was more persistent in the lung where it was detected in at least one of three hamsters on/after Day 11 post infection (Fig 3B). A small amount of viral genome detected in the brain of a PUUV i.n. infected hamster, 28 days post challenge, is the only viral genome detected in any organ at any time point post PUUV i.n. infection (Fig 3F). Serum viremia was detected in hamsters challenged with HTNV i.m. between 11 and 28 days post infection, though the presence of virus was sporadic except for Day 24. Serum viremia was only detected in two hamsters challenged with HTNV i.n., one on Day 11 and one on Day 17. No serum viremia was detected in hamsters challenged with PUUV by either route (Fig 3G). Similarly, the presence of viral genome was sporadically detected in the urine of Three HFRS-associated hantavirus infection models HTNV infected hamsters between days 17-28, but was not detected in PUUV challenged hamsters (Fig 3H).

In HTNV i.m. infected hamsters, infectious virus was first detected in the liver beginning at Day 11, and in the lung, liver, spleen, and kidney at Day 17 post infection (Fig 4B, 4C, 4D and 4E). With the exception of the kidney, in which infectious virus was recovered from every hamster after 17 days post infection, infectious virus was recovered from the lung, liver, brain and spleen in only a portion of hamsters at each time point. No infectious virus was detected in the heart at any time point (Fig 4A).

. Virus recovery from HTNV i.n. infected hamsters was markedly lower, with infectious virus being detected only in the kidney on Day17 and 28 post Three HFRS-associated hantavirus infection models infection (Fig 4E) . No infectious virus was recovered from the organ of any PUUV infected hamster (Fig 4) .

To confirm that the lack of infectious virus in the spleen was not due to the presence of inhibitors select samples were spiked with HTNV virus prior to a plaque assay. No significant inhibition of the spiked RNA was noted, confirming the RT-PCR results that HTNV infection does not result in viral dissemination to the spleen (S2C and S2D Fig) .

No infectious virus was detected in the urine for any hamster tested, even those that were RT-PCR positive for HTNV viral genome. In human ANDV infected patients with ANDV antigen positive urine, samples had to be cultured in Vero E6 cells for between 16-22 days post infection before infectious virus was detected [44] . Patients infected with HFRS hantaviruses exhibit leukocytosis and thrombocytopenia during infection [45, 46] . At every time point post infection, EDTA-treated whole blood from HTNV and PUUV exposed hamsters was evaluated to determine if changes in white blood cell count or platelets occurred. No changes were observed (S4 Fig). To further characterize disease, tissue sections from HTNV i.m. infected hamsters were analyzed by IHC and H&E to assess viral localization and any pathologic changes (S5 Fig). No significant histopathological findings were noted in the kidney or brain. Splenic follicular lymphoid hyperplasia and hepatic extramedullary hematopoiesis in the liver were each noted in three hamsters. Both are seen in animals from later time points in the study (days 17-28) and likely represent a reaction to HTNV infection, though other unidentified antigenic stimuli cannot be ruled out. Five hamsters between days 4 and 17 post infection exhibited a minimal to mild inflammation of the pericardium, characterized by a mixed lymphoplasmacytic histiocytic and neutrophilic infiltrate. Occasional macrophages in the inflammatory infiltrate within the pericardium are immunopositive suggesting a possible association with HTNV; however, no evidence of cardiac tissue injury is associated with the presence of HTNV antigen. Beginning on Day 4 post infection 66% (12/18) hamsters exhibit minimal (likely subclinical) respiratory lesions consisting of interstitial neutrophilic and histiocytic infiltrates in the lungs, with 44% (8/18) also exhibiting minimal amounts of alveolar edema. An additional hamster had minimal alveolar edema but not pulmonary infiltrates. Such findings suggest a response to antigenic stimulus and the presence of immunopositive endothelial cells, pneumocytes and alveolar macrophages suggest a response to HTNV infection.

Immunohistochemistry staining was uniformly negative in all tissues on Day 1 post challenge. Between Day 4 and 11 post challenge minimal antigen was sporadically detected in macrophages, pneumocytes and endothelial cells. By Day 17 post challenge mild to moderate levels of antigen were observed in endothelial cells and choroid plexus cells in the brain, in endothelial cells, macrophages and pneumocytes in the lung, and endothelial cells in the kidney. Minimal to mild levels of immunopositivity were found in endothelial cells and macrophages in the heart, and in endothelial cells, hepatocytes and Kupffer cells in the liver. Surprisingly, given the lack of viral genome and infectious virus recovered, mild to moderate amounts of immunopositive macrophages, lymphocytes and endothelial cells were found in the spleen (Table 2,

No published studies detail if ferrets are susceptible to hantavirus infection. To examine this, four ferrets were exposed to either 2,000 PFU HTNV or PUUV K27 i.n. No seroconversion occurred within 35 days post infection. The same animals were re-exposed to either 200,000 PFU HTNV, 94,000 PFU PUUV Beaumont (a human PUUV isolate) or 164,000 PFU PUUV Seloignes (a vole PUUV isolate) by i.m. Prior to the re-exposure one of the seronegative ferrets in the HTNV group was removed for health concerns (rapid weight loss) unrelated to the study, and its cage mate was subsequently removed for behavioral health reasons before the completion of the study. Data from those two ferrets are not shown. As soon as three days post infection ferrets began to lose weight with HTNV infected ferrets losing between 5-12% of peak body weight as did PUUV infected ferrets (Fig 5A-5C). By Day 35 post infection all animals had developed antibodies ag

against all strains of the virus as measured by N-ELISA assay (Fig 5D-5F). Neutralizing antibody development began as early as Day 14 post infection and all ferrets developed neutralizing antibodies by Day 28 post infection (Fig 5G-5L). The organs of infected ferrets were analyzed for viral load by RT-PCR and for the presence of infectious virus by plaque assay. The lung, liver, spleen, intestine and urine of HTNV infected ferrets were negative for viral genome, with small amounts ($<2 \log_{10}$) detected in the heart and spleen of a single animal (S7B Fig). The heart, lung, kidney, intestine and urine of PUUV infected ferrets were negative for viral genome, though small amounts were detected in the liver (2/4) and spleen (1/4) of PUUV infected ferrets (S7C and S7D Fig). Viral genome was spiked into the assay to confirm that the lack of signal was not due to the presence of inhibitors. All organs except for the intestine (4/6) had no inhibition of RT-PCR product. Similarly, no infectious virus was found in the organ of any ferret by plaque assay, despite spiked-in virus exhibiting no significant inhibition (S9 Fig). To confirm that immunosuppression of uninfected ferrets did not result in rapid weight loss, four healthy ferrets were immunosuppressed with a loading dose of 30 mg/kg Cyp followed by 10 mg/kg Cyp maintenance doses every other day (S10A Fig). Five days post immunosuppression ferrets exhibited rapid weight loss, fever, and lethargy due to secondary infection (S10B & S10C Fig). Immunosuppression was discontinued and 5 mg/kg enrofloxacin (a broad spectrum antibiotic) was given twice daily for a week. During this time, ferrets began to gain weight and their fever diminished. On Day 11 immunosuppression resumed for two doses with prophylactic enrofloxacin given once daily. Even with prophylactic antibiotics two ferrets spiked fevers within a few days post the second round of immunosuppressive treatment, though they did not lose weight (S10B & S10C Fig). Based on these results the rapid weight loss and clinical signs observed upon immunosuppression of HTNV and PUUV infected ferrets was most likely due to secondary infection. Due to the inability to completely manage secondary infection with prophylactic antibiotic treatment, no further immunosuppression studies were carried out in ferrets.

To refine the HFRS-causing hantavirus ferret infection model four ferrets were challenged with 94,000 PFU PUUV Beaumont i.m. on Day 0. Weight and temperature were recorded daily, while twice weekly blood draws and urine collection was used to monitor kidney function. As with the pilot experiment, ferrets slowly lost between 7-11% of peak body weight, recapitulating our previous findings (Fig 6A). No elevated temperatures were observed (Fig 6B). Ferrets developed a robust antibody response beginning on Day 10 post infection (Fig 6C and 6D). Neutralizing antibodies developed early as Day 14 post infection, and were present in all ferrets by Day 28 post infection, as measured by PsVNA and PRNT (Fig 6E and 6F).

Despite the slow weight loss no signs of renal impairment were observed. Proteinuria and hematuria are hallmarks of PUUV infection occurring in between 94-100% (proteinuria) and 58-85% (hematuria) of human clinical cases [46]. No prolonged proteinuria or hematuria was observed (Fig 7A and 7B) in infected ferrets. Similarly, blood urea nitrogen and creatinine levels in the sera, both of which are elevated due to kidney failure in PUUV patients [47] remained unchanged in PUUV infected ferrets (Fig 7C and 7D) [48, 49]. No infectious virus or viral genome was detected in the brain, heart, lung, liver, spleen, kidney, intestine, or eye (S11 and S12 Figs). No changes in other serologic or urologic parameters were noted (S13 and S14 Figs).

No gross pathological changes or significant lesions associated with PUUV infection were noted in the ferrets (S15 Fig). In the lungs, one ferret had mild neutrophilic and histiocytic inflammation centered on the bronchioles and expanded alveolar septa. Given that the inflammation was centered around bronchioles and not the vasculature, it is unlikely to be in response to PUUV infection. A number of other common or age-associated lesions were observed in the ferrets. Two of four ferrets had proliferative cortical cells in either the adrenal capsule or adrenal cortex that were likely clinically silent as they lack clinical signs consistent with adrenal-associated endocrinopathy. Additionally, alveolar mineralization was noted in all four ferrets as was eosinophilic and lymphoplasmacytic enteritis, and hepatitis. Two ferrets had fibromyxomatous degeneration of the atrioventricular valve. The spleen, brain, kidney, and pituitary

tary gland were normal in all ferrets examined.

As with ferrets, there is no literature on the susceptibility of marmosets to hantavirus infection. To test this, three male marmosets were exposed to 1,000 PFU HTNV i.m. Blood was collected weekly to measure seroconversion, serum viremia, as well as serum parameters relating to renal function. All three animals seroconverted by Day 21 post infection (Fig 8A and 8B) despite displaying no clinical signs of illness. Neutralizing antibody production began around the same time, and was robust by Day 30 post infection (14, 557 by PsVNA 50 and 10,240-20,480 by PRNT 50) (Fig 8C and 8D). Due to low volumes of blood drawn at each time point serum viremia could not be examined for each individual animal, however a pool of sera from all three animals was evaluated at each time point post infection. Low levels of serum viremia were detected between days 14 and 28 post infection by RT-PCR (Fig 8G) . As with the ferret infection model of HFRS-causing hantavirus, no renal injury as measured by changes noted in blood urea nitrogen or serum creatinine were observed over the course of infection, nor were changes in any of the other serum parameters monitored (S16 Fig). Animals were euthanized on Day 30 post infection to examine organs for viral dissemination. No infectious virus or viral genome was detected in the heart, lung, liver, spleen, kidney, intestine or brain (S17 and S18 Figs). Lymphoid hyperplasia was noted in all three animals by histology, though the location and intensity varied between the spleen, various lymph nodes and gut associated lymphoid tissue. Such hyperplasia is indicative of a response to antigenic stimulation and was most likely caused by the viral challenge. Mild to moderate congestion Three HFRS-associated hantavirus infection models was also seen in the lungs of each marmoset; however, this was an acute change and was most likely associated with terminal anesthesia and euthanasia. No other significant histologic lesions were noted in any of the three animals (S19 Fig). Due to the negative virology results and the lack of significant histologic changes, IHC to detect the presence of HTNV antigen in tissues was not performed.

To date, the use of adult animal models to evaluate anti-HTNV and anti-PUUV medical counter measures has been limited. Recombinant protein, vaccinia virus-vectored, and DNA vaccines have been tested in the high dose HTNV hamster model [9, 37, 50] . Additionally, the ability of passively administered neutralizing antibodies has been evaluated in the high-dose hamster model of HTNV and PUUV and in PUUV challenged cynomolgus macaques PUUV [50, 51] . Both of these models have limitations; the size of the macaque requires large quantities of passive transfer material, and the high dose of the hamster model, with challenge doses of ~650 ID₅₀ , could require prohibitively large volumes of test article to neutralize the high initial dose. Suckling mice, which present with a disseminated disease not reminiscent of HFRS, have been used to evaluate HTNV therapeutics as well [52] [53] [54] [55] . In this paper we present three adult animal models of HFRS-causing hantavirus infection that can be used for future evaluation of therapeutics, biologics and vaccines.

The ID₅₀ 's for HTNV and PUUV determined in this report are similar to the lethal dose 50% (LD₅₀) calculated for SNV and ANDV, <3 PFU via the i.m. route [6, 7, 41] . Also similar is that the challenge dose of ANDV required to infect/kill 50% of hamsters by the i.n. or intragastric route is ~10-30 fold higher [56] . Previous reports also demonstrate that PUUV is capable of infecting hamsters by the intragastric route, with an ID₅₀ of >10,000, making it a much less effective route of infection [57] . Despite similarities in their ID₅₀ 's, the ID₉₉ 's of HTNV and PUUV greatly diverge, with ~200 times less HTNV required to infect hamsters via the i.m. route, and ~2 fold less required to infect via the i.n. route (Table 1). The lower ID₉₉ doses, coupled with viral persistence in HTNV infected animals as opposed to viral clearance (Fig 3) , suggest that HTNV is more infectious than PUUV in the hamster. The mechanism for this difference between these closely related hantaviruses remains unknown.

Lethal infection of hamsters with ANDV leads to extensive organ dissemination, with infectious virus recovered in the lung, liver, kidney, spleen and heart [6] . Asymptomatic infection of hamsters with SNV repeatedly passaged through hamsters has a similar organ distribution [58] . The organ distribution of both those viruses is similar to

o the low dose HTNV hamster model, with two notable exceptions. First, the HTNV model has low, transient levels of virus in the liver and hardly any virus in the spleen (Figs 3 and 4). This dissimilarity between the models led us to confirm the lack of virus in the spleen was not due to the presence of inhibitors by spiking either infectious virus or viral RNA into samples prior to evaluation (S2 Fig). Second, in the ANDV model, the presence of virus was determined by plaque assay, indicating the virus was infectious and replication competent. In the HTNV low dose hamster model, while there is detection of high levels of viral genome (and in the case of the HTNV i.m. model, by pathology) by RT-PCR, recovery of infectious virus is sporadic, typically occurring at low levels, in only a few hamsters per time point (Figs 3 and 4, Table 1). The discrepancy between RT-PCR/pathology and plaque assay is notable. The hamsters in this study were not perfused, and given the appearance of neutralizing antibodies as early as day XX post infection, the presence of these antibodies could be impairing our ability to recover live virus via the plaque assay. It is also possible that viral packaging is somehow impaired in the hamster, leading to a larger amount of nucleocapsid protein and viral genome than live virus. Further studies need to be undertaken to elucidate.

The low dose HTNV hamster model also mimics the infection pattern of the virus in its host species, the striped field mouse (*Apodemus agrarius*), with viral genome being detected in the lung, liver and kidney but not the spleen (the heart and brain were not examined) [32, [59] [60] [61].

In contrast to the HTNV low dose model, the organ distribution of virus in the PUUV low dose model is transient. In PUUV i.m. infected hamsters virus is detectable by RT-PCR around Day 11 post infection, with the virus being cleared from all organs except the lung by Day 24. No infectious virus was recovered at any time point examined. Even less virus was detected in the PUUV i.n. model; only a small amount in the brain of one hamster on Day 28 post infection. In neither case was serum viremia observed (Figs 3 and 4). This is most similar to the SNV models involving low passage virus in immunocompetent hamsters: the virus is transiently detected in the lung by PCR, and then sporadically found in organs 12 to 14 days post infection using immunohistochemistry [41, 58]. The distribution of PUUV in the hamster differs somewhat from its host species the bank vole, where it is found to persist in the lung, spleen and kidney, and was not detected in the heart or the brain [62].

Seroconversion of hamsters post viral exposure remains the best way to measure infection, and should be considered the primary endpoint for efficacy studies. Though the PRNT assay was slightly more sensitive than the N-ELISA assay, detecting neutralizing antibodies in all animals with N-ELISA titers, and in two PUUV animals that did not have N-ELISA titers, the increased time, sample, and biosafety requirements necessary for a PRNT assay make the N-ELISA a better choice (Fig 2, S1 Fig). For a 10 PFU HTNV i.m. challenge, given lower titer and specific OD sum values as compared with higher challenge doses, and the fact that one hamster with significant viral genome in its organs at Day 28 post infection (7.1 log₁₀ in the brain, and 6.8 log₁₀ in the kidney) did not seroconvert, waiting until Day 35 post infection to monitor seroconversion is advisable (Figs 1-3). Viral load in the brain, kidney, and lungs as measured by RT-PCR need to be evaluated at Day 35 post low-dose challenge to determine their usefulness as secondary endpoints. Recovery of infectious virus in any organ, and viremia are too sporadic to serve as proxy markers for infection.

The ferret has been used as an experimental model for numerous hemorrhagic fever viruses, and respiratory viruses [12] [13] [14] [15] [16] [17], though no published reports exist examining its susceptibility to hantavirus infection. In comparison to the hamster, the ferret is far more resistant to HTNV and PUUV infection. Exposure of ferrets to 2,000 PFU i.n. (greater than the ID₅₀ for both viruses in hamsters), failed to result in a productive infection and seroconversion. Instead i.m. challenge doses of ~100,000-200,000 PFU were needed (Fig 5). Initially ferrets were exposed to PUUV K27, a commonly used laboratory isolate that has been in cell culture for over 15 years. Repeated passaging of hantavirus is known to cause mutations [63] [64] [65]. In c

contrast PUUV Beaumont and Seloignes are relatively recent isolates, with no more than 3 passages in cell culture post isolation. These strains were used for all subsequent experiments to maximize the likelihood of PUUV to cause disease by eliminating possible attenuation of the laboratory strain of the virus. Despite the high challenge dose and use of recent isolates, no elevation in white blood cells was observed over the course of the experiment, no pathology or organ burden was detected at the conclusion of the experiment, and the N-ELISA specific OD sum, PRNT 50, and PsVNA 50 titers remained low (Figs 6 and 8 and S9 Fig). This outcome is almost identical to that of Marburg and Ravn virus infection in ferrets, where the development of neutralizing antibody titers was the only sign that productive infection occurred [66]. Previous experiments in hamsters, have demonstrated that 2×10^5 PFU of gamma-irradiated ANDV and SNV are not sufficient to cause seroconversion, and neither is 1×10^4 gamma-irradiated PFU PUUV [6, 57]. Thus the seroconversion observed in ferrets, though low, is not likely to be due to a reaction to the large quantity of antigen administered, but to a productive infection.

PUUV antigen, viral genome, or infectious virus has been found in the brain, pituitary gland, lung, heart, liver, kidney, spleen, cerebrospinal fluid, and gastrointestinal tract of human patients with clinical symptoms of NE, though the pattern of viral dissemination varies between individuals [67] [68] [69] [70]. Acute kidney injury and vision disturbances including blurred vision, myopic shift, and lens thickening, while pulmonary involvement including pleural effusion and vascular congestion, and renal failure occurred less frequently [47, [71] [72] [73] [74]. Given the lack of high neutralizing antibody titers, which could have aided in viral clearance, the lack of viral genome in any of the ferret organs examined is rather surprising (S6, S7, S11 and S12 Figs). Furthermore, the lack of viral antigen and pathology in the organs tested suggest either a transient infection cleared prior to Day 35 post infection, levels of virus so low as to be undetectable by the tests used, or a viral reservoir outside of the organs tested.

The lack of detectable virus is most surprising given the gradual weight loss infected animals exhibit (Figs 5 and 6). The animals gained weight until ~3 days post challenge, at which point a gradual weight loss occurs, regardless of if HTNV or PUUV was the challenge virus. While weight loss is a feature of other ferret models of infectious diseases, the pattern we observed was unique: ferrets infected with morbillivirus, avian influenza and filoviruses rapidly lose weight during the first week to two weeks post infection, while infection with severe acute respiratory syndrome virus results in no significant weight loss [16, 17, [75] [76] [77].

In the hamster, infection with SNV is asymptomatic unless the hamster is immunosuppressed. When ferrets were immunosuppressed on Day 42 post infection, rapid weight loss and lethargy ensued (Fig 5). Given that these clinical signs were also observed in unchallenged control animals (S10 Fig), this could likely be the result of a secondary infection or drug toxicity. The use of Cyp is well documented in ferrets, primarily given at a high dose as an emetic, and no dosage for long term immunosuppression was found [78]. The dosages used in this study (between 10-30 mg/kg) successfully reduced white blood cell levels, in ferrets and demonstrate that Cyp can be used to induce long-term immunosuppression, if antibiotics are given to control for secondary infection (S8 Fig).

Despite not being able to detect infectious virus or viral genome in the kidney, we hypothesized that the weight loss we observed could be due to kidney failure. Individuals with HFRS exhibit proteinuria and hematuria, both of which can indicate kidney damage [46, 47]. Additionally, serum blood urea nitrogen and creatinine levels are both elevated in HFRS patients and provide a second way to measure kidney function [46, 47, 73]. Decreased platelet levels also characterize clinical HFRS in humans, impairing coagulation [46, 47, 73]. In a second experiment designed to monitor kidney parameters, PUUV infected ferrets exhibited the same gradual weight loss that characterized the first experiment. However, no prolonged signs of clinical kidney failure were observed: blood urea nitrogen and creatinine levels did not dramatically increase over

r the five week study period. Only one animal exhibited proteinuria (day 35), and two exhibited hematuria (one on Day 4, and one on Day 21) (Fig 7) . Similarly to parameters monitoring renal failure, no thrombocytopenia occurred in PUUV infected ferrets (S9 and S13 Figs). The cause of the weight loss remains undetermined.

Though susceptible to both HTNV and PUUV, the ferret has limited usefulness for studies involving medical countermeasure efficacy testing. The ferret's large mass, even as an adolescent, makes the amount of test article needed also prohibitively large. The large challenge dose required for infection could potentially obscure the protective effect of drugs or vaccines, due to the overwhelming amount of virus administered. Moreover, although the animals are infected the resultant neutralizing antibody titers are small, resulting in potential sensitivity issues with the model. Husbandry and handling of the animals under ABSL-3 procedures is also substantially more difficult than hamsters, and they lack the genetic similarity to humans that marmosets possess.

In this study we have demonstrated the susceptibility of marmosets to HTNV infection. Marmosets represent an attractive model for testing vaccines and therapeutics against HFRS-causing hantaviruses due to genetic similarities to humans and small size. Also, the model has a simple read-out of infection, i.e. robust antibody production as measured by N-ELISA and PsVNA, making the determination of protection by vaccine or passive transfer material, straightforward. Further optimization of the model, namely to determine the ID₉₉, could prove important as a 1,000 PFU challenge dose could be excessively high and prohibit therapeutic effects of candidate medical countermeasures.

Overall the marmoset model is more similar to the ferret HFRS-causing hantavirus infection model than the hamster, though there are key differences. Like the ferret, no significant pathological abnormalities were noted, and no signs of renal failure were observed (S16 and S19 Figs). Serum chemistry values do not differ from the normal range with the exception of albumin, total bilirubin, and amylase. While these values fell outside the normal range, they did not change over the course of infection indicating the problem may lie in the reference values used. The Piccolo general chemistry panel used to evaluate the parameters is optimized for human testing, and therefore may be less than optimal for evaluating the marmoset, especially those parameters. Additionally, no infectious virus or viral genome was recovered from any organ at Day 30 post infection [79, 80] . This is not surprising, given the high levels of neutralizing antibodies present as early as 21 days post infection (Fig 8) . Unlike the ferret, however, marmosets develop exceptionally high neutralizing antibody titers (10,240-20,480 by PRNT 50 and 14,866-221,557 by PsVNA 50), and display low-level serum viremia between two and four weeks post infection (Fig 8) . The serum viremia is significantly lower than in hamsters infected with HTNV, where some animals displayed RT-PCR titers of $>7 \log_{10}$, and in hamsters infected with ANDV, where infectious virus titers prior to death are $> 6 \log_{10}$ [41].

Despite not exhibiting clinical signs of disease, the model's robust antibody response (as measured by PRNT, PsVNA and N-ELISA) make it a useful tool for evaluating vaccines and pre-or post-exposure therapeutics.

This paper has explored the use of three laboratory animal species as possible infection and disease models for HFRS-causing hantaviruses: the hamster, the ferret, and the marmoset. These models, especially the hamster model and marmoset model, will be useful for evaluating medical countermeasures with the potential to induce sterile immunity. The marmosets should be particularly useful for the evaluation of passively transferred protective human antibodies because of the relative genetic similarities between species in the Order Primates, and the small size of marmosets, allowing testing with smaller volumes of material than would be required for larger species such as macaques. . Heart, lung, liver, spleen, kidney, and intestine were collected and assayed by plaque assay for the presence of infectious virus (A). To confirm lack of virus recovered was not due to inhibitors, virus was spiked into serial dilutions of organ homogenate to confirm no inhibitor was present (B). For a standard plaque assay the limit of detection, $1.7 \log_{10}$, is depicted as a dashed line in (A). In (B) the dashed

d line is amount of HTNV plaques obtained when spiked into media rather than organ homogenate. (TIF) S10 Fig. Immunosuppression of uninfected ferrets leads to rapid weight loss and secondary bacterial infection. Uninfected ferrets were administered 10mg/kg Cyp, and the antibiotic enrofloxacin, according to the schedule in (A). Weight (B) and temperature (C) are shown. b.i.d indicates antibody was administered twice daily, and q.d indicates antibiotic was administered daily. (TIF) S11 Fig. PUUV infected ferrets had no infectious virus in the organs. Ferrets were infected with 94,000 PFU PUUV Beaumont i.m. Heart, lung, liver, spleen, kidney, intestine, brain, eye, and adrenal gland were collected on day 35 post infection and assayed for infectious virus by plaque assay (A). Virus was spiked in to confirm no inhibitor was present (B). For a standard plaque assay the limit of detection, $1.7 \log_{10}$, is depicted as a dashed line in (A). In (B) the mean \pm SEM is depicted in all spiked groups and the dashed line is amount of HTNV plaques obtained when spiked into media rather than organ homogenate. To confirm no inhibitors were present, virus was spiked into samples (B). For a standard plaque assay the limit of detection, $1.7 \log_{10}$, is depicted as a dashed line in (A). In (B) the mean \pm SEM is displayed for all spiked groups, and the dashed line is amount of HTNV plaques obtained when spiked into media rather than organ homogenate. (TIF)

Three marmosets were infected with 1,000 PFU HTNV i.m. On Day 30 organs were harvested, and the presence of viral genome was determined by RT-PCR. To confirm no inhibitors were present, viral genome was spiked into samples. Heart (A), lung (B), liver (C), spleen (D), kidney (E), intestine (F), and brain (G) were collected. The mean \pm SEM is shown for the not spiked (NS) and spiked (S) groups and the limit of detection, $1 \log_{10}$, is depicted as a dashed line.



Stop words

Here's a quote from Stanford's NLP team that will provide some context on stop words and their intended usage:

"Sometimes, some extremely common words which would appear to be of little value in helping select documents matching a user need are excluded from the vocabulary entirely. These words are called stop words. The general strategy for determining a stop list is to sort the terms by collection frequency (the total number of times each term appears in the document collection), and then to take the most frequent terms, often hand-filtered for their semantic content relative to the domain of the documents being indexed, as a stop list, the members of which are then discarded during indexing."

Let's investigate the stop words list that we will use to clean our data. Note that apart from the standard stopwords, we will also remove any punctuation and also any occurrences of *et al.*, as they are often found in academic articles.

```
In [12]: stop_tokens = nltk.corpus.stopwords.words('english') + list(punctuation) + ['et', 'al.'].
print(stop_tokens)
```

['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', 'you're', 'you've', 'e', 'you'll', 'you'd', 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', 'most', 'other', 'some', 'such', 'no', 'nor', 'not', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", 'won', "won't", 'wouldn', "wouldn't", '!', '"', '#', '\$', '%', '&', "'", '(', ')', '*', '+', ',', '-', '.', '/', ':', ';', '<', '=', '>', '?', '@', '[', '\\', ']', '^', '_', '`', '{', '|', '}', '~', 'et', 'al.']

The code below will be used to read the text associated with a series of articles, remove stop words from their text, and reduce) inflected words to their base form (stemming).

NOTE:

If you are not familiar with Python code, just execute the following cell and continue with the notebook. Understanding the code below is not required for understanding and following the overall flow of the notebook.

```
In [13]: class Reader:
        """Class used to read the files associated with an article"""

        def __init__(self):
            self.stemmer = SnowballStemmer('english')

        def read_file_to_json(self, filepath):
            try:
                with open(filepath, 'r') as f:
                    data = json.load(f)
            except FileNotFoundError as e:
                mount.start()
                with open(filepath, 'r') as f:
                    data = json.load(f)

            return data

        def parse_document(self, document_index):
            document = metadata_with_docs.iloc[document_index]

            # One article can have multiple associated documents
            words = []
            for filename in document['pdf_json_files'].split('; '):
                filepath = '{0}/{1}'.format(covid_dirpath, filename)
                pdf_json_files = os.listdir(covid_dirpath + '/comm_use_subset/pdf_json')
                filepath = covid_dirpath + '/comm_use_subset/pdf_json/' + pdf_json_files[0]

                if document.index % 50 == 0:
```

```

        print(filepath)

        data = self.read_file_to_json(filepath)

        # Split each paragraph into multiple sentences first, in order to improve tokenization
        text = data['body_text']
        for paragraph in text:
            p_sentences = sent_tokenize(paragraph['text'])

            # Split each sentence into words, while making sure to remove the stopwords
            for p_sentence in p_sentences:
                sentence = [ self.stemmer.stem(word) for word in word_tokenize(p_sentence)]
                words.extend(sentence)

        return (words, document['cord_uid'])

class Corpus:
    """An iterator that reads all sentences from the first N documents"""

    def __init__(self, n_documents):
        self.n_documents = n_documents
        self.stemmer = SnowballStemmer('english')
        self.reader = Reader()

    def __iter__(self):
        for document_index in range(0, self.n_documents):
            words, document_id = self.reader.parse_document(document_index)
            yield TaggedDocument(words, document_id)

    def plain_iter(self):
        for document_index in range(0, self.n_documents):
            words, document_id = self.reader.parse_document(document_index)
            yield (words, document_id)

```

Encoding documents as vectors

In this lab, we're using a subset of 1000 articles to train a Machine Learning model that encodes text documents into numerical vectors (a document embedding model).

Training a document embedding model takes a significant amount of time, and for this reason we already provide a trained model. We also provide the code below in case you want to get more details about the process. Running the next two cells will result in loading the already existing model.

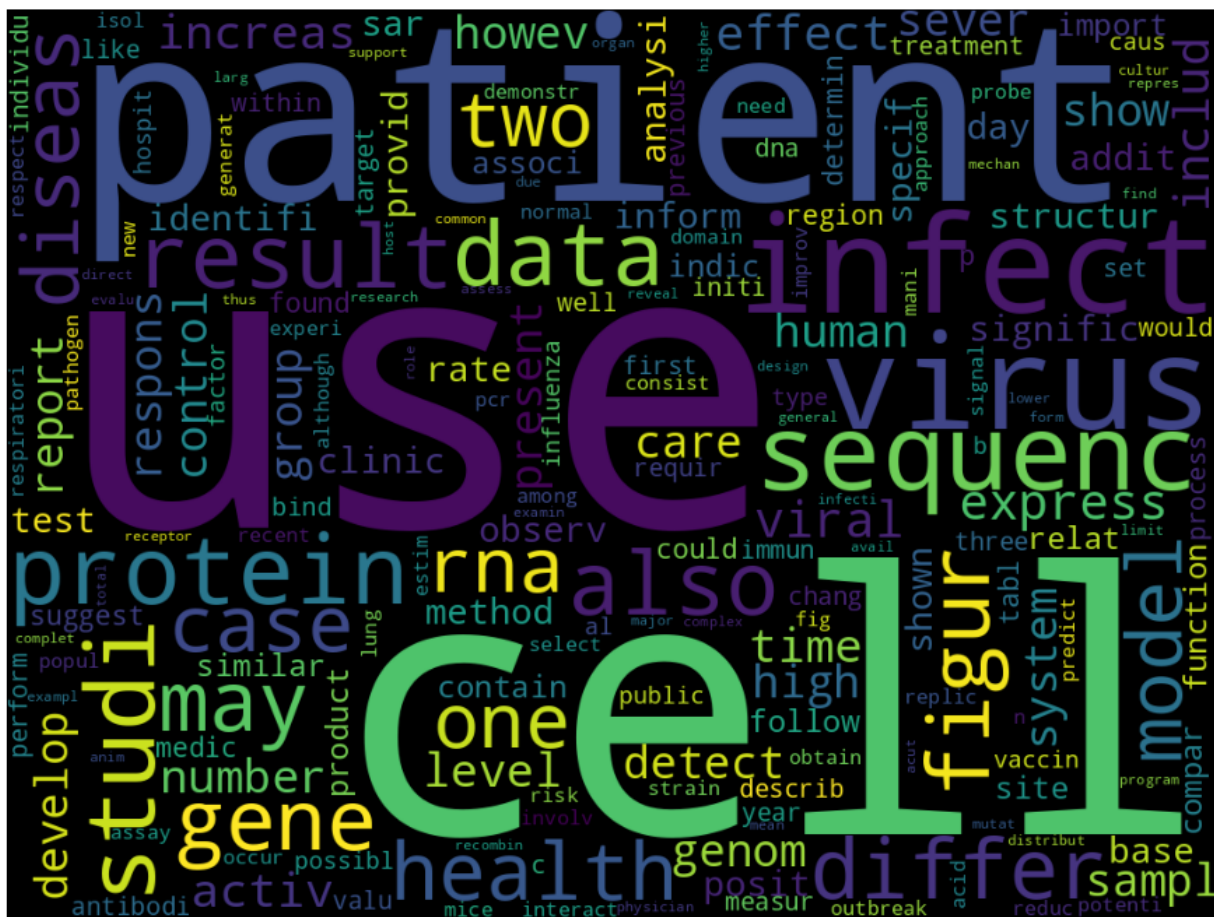
```
In [14]: N_DOCUMENTS = 500
```

```
In [15]: %%time
```

```
Done, loaded word2vec model with 10751 words.
CPU times: user 123 ms, sys: 11.1 ms, total: 134 ms
Wall time: 459 ms
```

Let's analyze the relative frequencies of words in the corpus of articles. We will display a word cloud to provide a visual representation of these relative frequencies.

```
Out[17]: (-0.5, 1023.5, 767.5, -0.5)
```



Embedding documents

Below is an example on how we embed (transform from text to numerical vector) one of the documents.

```
In [18]: words, doc_id = Reader().parse_document(DOCUMENT_INDEX)
         model.infer_vector(words)

/tmp/tmpemabxly/covid19temp/comm_use_subset/pdf_json/02a009e42054081b441d0f4b203679c
4b0cae38d.json
Out[18]: array([-0.36768144, -1.1106528 ,  2.4751897 , -0.8354529 , -1.1695763 ,
                0.80843717, -0.39670926,  0.48410633, -0.7084072 , -1.5064273 ,
                0.88584656, -2.4492157 ,  3.2096512 ,  0.11121531,  2.2204032 ,
               -1.1593775 ,  0.21179079,  5.6468253 , -0.7457158 , -1.2919372 ,
               -0.50090504, -1.0035561 ,  1.0373933 ,  0.00688864,  1.0333565 ,
                0.8773138 , -0.56150395,  0.18981826, -0.81829715, -1.4064279 ,
               -0.86820185, -1.885929 , -0.07239459,  1.0694723 ,  0.6629561 ,
               -2.0599358 ,  1.4812938 , -0.26281548,  0.20125028,  0.85074055,
               -2.4990096 ,  0.26456544, -0.6714662 , -0.4843073 ,  1.0774498 ,
                1.5166879 , -0.35177374,  0.8190227 , -1.7486702 , -2.1246812 ,
               -1.4046166 , -0.70962495, -1.4102095 , -0.2974848 , -1.55897 ,
                0.19682325, -0.15490614,  2.4801288 ,  1.0956861 ,  0.27181962,
                0.9518106 , -1.2184901 ,  0.8317703 ,  1.8769274 , -2.0623431 ,
               -1.1901475 , -2.263474 ,  0.3310059 ,  0.56674606, -0.11053646,
                2.2180562 , -0.28195322,  0.8319942 ,  1.0140984 ,  2.1507628 ,
                0.9072955 , -1.1474484 ,  2.2471466 ,  1.7143121 , -1.4232329 ,
               -0.41604537, -2.7342613 , -1.030752 ,  0.6367958 ,  0.53506535,
                1.7867123 ,  0.23827484, -0.82510537,  0.691904 , -0.8767283 ,
               -2.7645059 , -0.28387508,  1.8029312 ,  1.5281334 , -0.25860503,
               -0.56525475, -1.5408717 , -1.1052479 ,  0.76068234,  1.4443927 ,
                0.85816324,  0.23737751,  0.60747755, -0.88963866,  0.64770806,
                0.2448809 ,  1.3057928 ,  1.5467266 ,  1.1140411 , -2.052351 ,
               -0.33170623, -0.26855245,  0.45625284,  1.8268589 ,  0.6162205 ,
                1.8192254 , -0.9949126 ,  1.683619 , -2.219154 , -0.38159177,
               -0.4732724 , -1.9200023 , -0.04358702, -2.1741493 , -1.1506021 ,
                1.4797405 ,  0.08616818,  1.4481105 ], dtype=float32)
```

And this is an example of a trivial "document" (containing a single, trivial sentence) going through the same process. Notice how, regardless of the length of the sentence, the result vector is always the same size - the `vector_size` argument used while training the `Doc2Vec` model.

This is very important in the following stages of the process, when we are working with multiple documents.

```
In [19]: model.infer_vector(['human', 'love', 'cat', 'dog'])
```

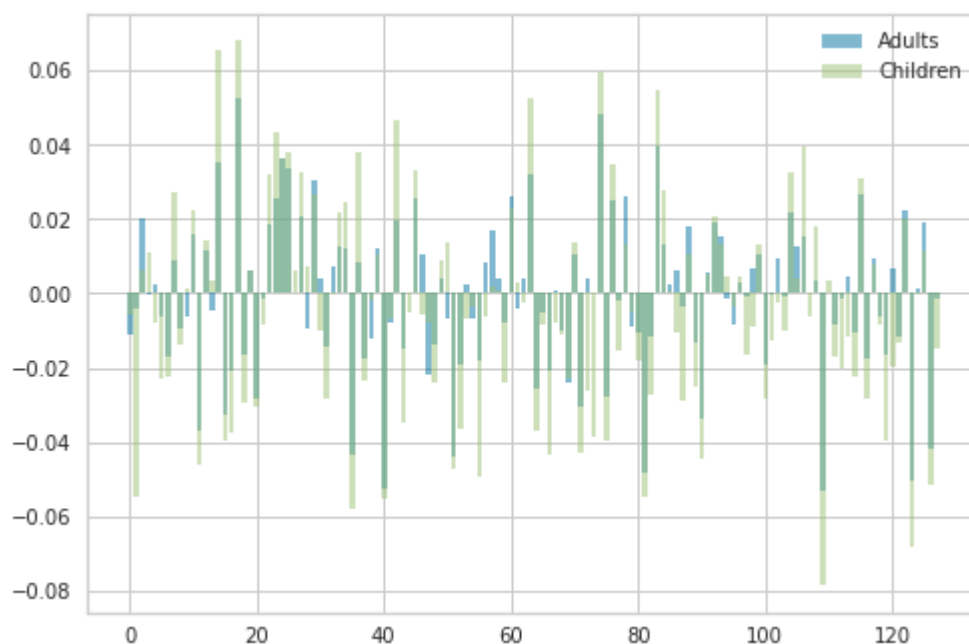


```
Out[19]: array([ 0.01187267, -0.07222479,  0.0009857 ,  0.02489351, -0.00371165,
        -0.05244752, -0.04042441,  0.00981789, -0.00546434, -0.0114641 ,
         0.00382181, -0.0549628 , -0.00109327,  0.02235473,  0.06640013,
        -0.02146467, -0.00610835,  0.07812942, -0.01487299, -0.00225763,
        -0.02316215, -0.01193333,  0.01044642,  0.03996939,  0.02644479,
         0.0227175 ,  0.04123925,  0.01465284,  0.00910605,  0.02305634,
        -0.00566173, -0.03546144, -0.02495514,  0.04078628,  0.01669642,
        -0.03630543,  0.03745678, -0.01681023,  0.01524001,  0.02174 ,
        -0.04845124, -0.02859627,  0.02559633, -0.01294257,  0.01237459,
         0.05609317, -0.01783988,  0.00306992, -0.02992033, -0.01554093,
         0.01374504, -0.03704116, -0.01849834,  0.00342414, -0.00394119,
        -0.01736315,  0.00061388,  0.00300337,  0.02583033, -0.02169578,
         0.03127309, -0.00914276, -0.00147238,  0.06283462, -0.02120893,
         0.00339467, -0.03140209, -0.0407921 , -0.01630779, -0.01234598,
         0.03513828, -0.05091879, -0.03363749, -0.04252905,  0.03870467,
        -0.02403057,  0.03826586, -0.00683732, -0.005172 , -0.03065385,
        -0.03811062, -0.02149642, -0.02606765,  0.02589219,  0.01650087,
        -0.00108351,  0.00658174, -0.03633691,  0.01417964, -0.01604395,
        -0.03558038,  0.02062096,  0.00863639, -0.00796049, -0.00531631,
         0.00715494,  0.03192762, -0.00160919, -0.00166048,  0.00513896,
        -0.02604181, -0.01912436,  0.02104215, -0.01204367,  0.01967912,
         0.00946659, -0.00244938, -0.01942666,  0.00923178, -0.07924119,
        -0.00391558, -0.00874832, -0.02192896,  0.01491943,  0.00303279,
         0.02053712, -0.04007722,  0.01599913,  0.00882744, -0.04477157,
        -0.02722152,  0.00240392,  0.02242075, -0.06093423, -0.03261027,
         0.02414383, -0.03612995, -0.00615444], dtype=float32)
```

The resulting vectors will look more or less similar, depending on how different the contents of the articles are themselves. See below the differences resulting from a single word change - some of the values significantly overlap, while others are quite different if not opposite.

```
In [20]: adult_vector = model.infer_vector(['adult', 'love', 'cat', 'dog'])
        child_vector = model.infer_vector(['child', 'love', 'cat', 'dog'])
        labels = range(0, 128)

        plt.bar(labels, adult_vector, align='center', alpha=0.5)
        plt.bar(labels, child_vector, align='center', alpha=0.5)
        plt.legend(['Adults', 'Children'])
        plt.show()
```



Let's now do the same for the same for all the documents we're focusing on.

```
In [21]: %%time

word_vectors = []
ids = []

for (words, doc_id) in Corpus(N_DOCUMENTS).plain_iter():
    ids.append(doc_id)
    word_vector = model.infer_vector(words)
    word_vectors.append(word_vector)
    if len(word_vectors) % 100 == 0:
        print(f'Processed {len(word_vectors)} documents.')
```

```
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/02a009e42054081b441d0f4b203679c4b0cae38d.json
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/0b180c5c5edf329811114548e19a708303e7c1c2.json
Processed 100 documents.
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/0ca8648d40bee3056b3f9840de6d34b57ed121d0.json
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/0df2a9766ade17a0d9a625ef02722fc167ee0526.json
Processed 200 documents.
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/0f7bb2b30b0eba1a065a6dfc88dbbd99053ff1ba.json
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/1b02698e082376846f59c99c449161a7a7eb737f.json
Processed 300 documents.
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/1c8afcb822dac8e8e5b33a85c2c2e9c7cc24a25a.json
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/1dd28ef7fb600a0ca94980bcf6e5cbc52a77fe.json
Processed 400 documents.
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/1f2bbc79b56c51c0ae26ced2d7d8ccf10360fa72.json
/tmp/tmpemaxbzly/covid19temp/comm_use_subset/pdf_json/2a8a6ddd84f0c80ce33fed8960c3201c08e56854.json
Processed 500 documents.
CPU times: user 47.1 s, sys: 279 ms, total: 47.4 s
Wall time: 1min 4s
```

Now that we've finished reading the articles, we can dismount the dataset in order to free up resources

```
In [22]: mount.stop()
```

Covid-19 documents prepared for Machine Learning

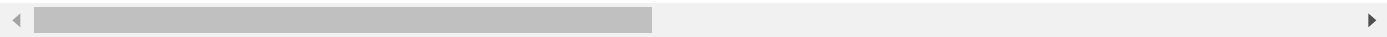
We'll create a new DataFrame using the word vectors we've just calculated, this is the numerical form of the documents which is ready for Machine Learning workloads.

```
In [23]: wv_df = pd.DataFrame(word_vectors, index=ids)
wv_df
```

Out[23]:

	0	1	2	3	4	5	6	7	
ug7v899j	-0.360649	-1.138998	2.476880	-0.788588	-1.130553	0.846280	-0.359760	0.609766	-0.69
02tnwd4m	0.054707	-2.481819	0.917430	1.064308	-0.671468	-1.027804	-1.930289	0.263879	-0.00
ejv2xln0	-0.072066	-0.641108	0.855729	-0.032554	-0.280952	-1.078268	-1.413609	0.141413	0.58
2b73a28n	-1.238587	-0.418109	0.669005	-0.239547	-1.073317	0.036139	-1.333731	0.813179	-0.35
9785vg6d	0.396538	-1.957209	0.616441	-0.054515	-0.672714	-0.731759	-0.851979	-0.648167	-1.44
...
v6jyplcn	-0.991498	-1.217878	1.090420	0.815862	-0.347704	-0.737506	-1.035312	1.022310	-0.02
fae3sczm	-0.779463	0.317488	1.283767	0.217249	0.233826	-0.167178	-1.575740	-0.245862	-0.37
1c4m2fym	1.698243	-3.183822	0.957596	0.766598	0.914241	-3.589298	-2.758309	-0.146896	-0.77
1ldebnq8	-0.771220	-1.271643	0.470235	0.511493	-0.262360	-0.795250	-1.670549	-0.151086	-0.05
x7wva1ax	-0.376713	-2.424713	1.729154	1.284823	-0.440978	-2.996548	-3.534384	0.270851	-0.44

500 rows × 128 columns



We'll join the DataFrame containing the numerical embeddings with the original dataset.

```
In [24]: indexed_metadata = metadata_with_docs.set_index('cord_uid')
metadata_with_embeddings = pd.concat([indexed_metadata.iloc[:N_DOCUMENTS], wv_df], axis=1)
metadata_with_embeddings
```

Out[24]:

	sha	source_x	title	
ug7v899j	d1aafb70c066a2068b02786f8929fd9c900897fb	PMC	Clinical features of culture-proven Mycoplasma...	10.1186/1471-22
02tnwd4m	6b0567729c2143a66d737eb0a2f63f2dce2e5a7d	PMC	Nitric oxide: a pro-inflammatory mediator in l...	10.11
ejv2xln0	06ced00a5fc04215949aa72528f2eeaae1d58927	PMC	Surfactant protein-D and pulmonary host defense	10.11
2b73a28n	348055649b6b8cf2b9a376498df9bf41f7123605	PMC	Role of endothelin-1 in lung disease	10.11
9785vg6d	5f48792a5fa08bed9f56016f4981ae2ca6031b32	PMC	Gene expression in epithelial cells in respons...	10.11
...
v6jyplcn	ba1a74766e0d96756105e85ce014b4bbe86fe819	PMC	Immune responses against severe acute respirat...	10.1111, 2567.2007.
fae3sczm	00acd3fd31ed0cde8df286697caefc5298e54df1	PMC	Distinguishing Molecular Features and Clinical...	10.1371/journal.pone.0
1c4m2fym	808dfc4c59f3e2e9150aa5542ea227718741388b	PMC	Towards a Coronavirus-Based HIV Multigene Vaccine	10.1080/17402520600
1ldebnq8	9cf43ad7346d9099015ee0545e0b3498c10213b4	PMC	Epithelial Cell Apoptosis and Neutrophil Recru...	10.2119/2008-000
x7wva1ax	3ccc18758132523e4b94a6cd392053a5b4b542cb	PMC	The cucumovirus 2b gene drives selection of in...	10.1093/nar/gl

500 rows × 146 columns

Preparing for clustering documents

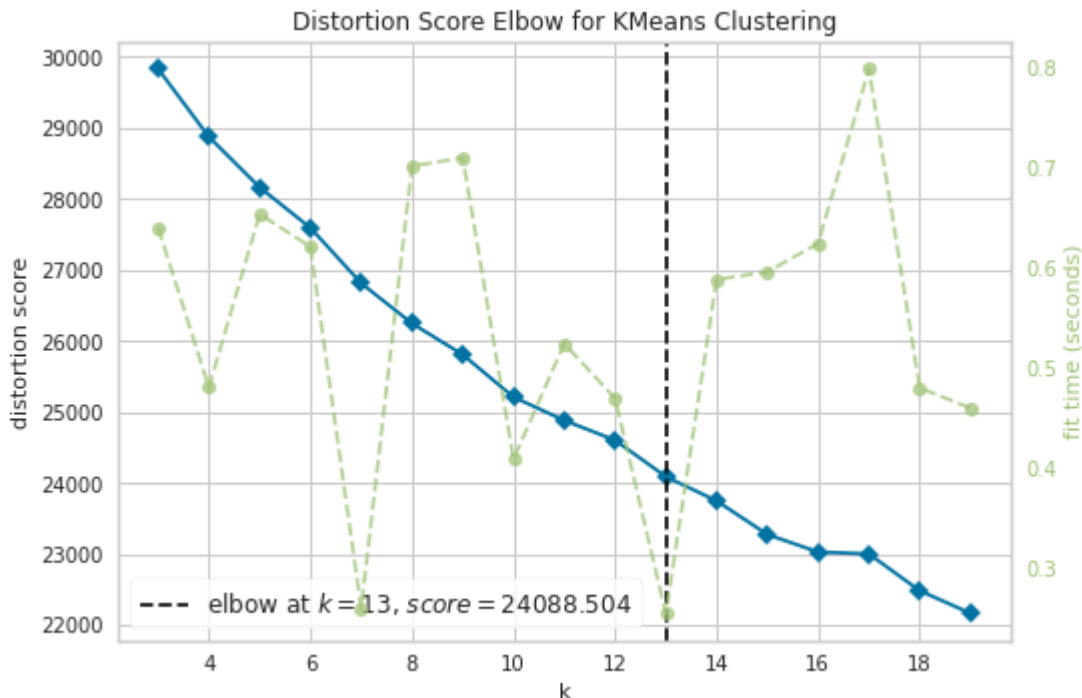
One of the challenges with clustering is to find the ideal number of clusters to look for. The elbow method is one of the most common approaches.

We're visualizing an elbow metric (the "distortion" score) and trying to find a point where it stops decreasing with the number of clusters.

```
In [25]: visualizer = KElbowVisualizer(KMeans(), k=(3,20))
visualizer.fit(wv_df)

visualizer.show()
```

findfont: Font family ['sans-serif'] not found. Falling back to DejaVu Sans.
findfont: Generic family 'sans-serif' not found because none of the following families were found: Arial, Liberation Sans, Bitstream Vera Sans, sans-serif



```
Out[25]: <AxesSubplot:title={'center':'Distortion Score Elbow for KMeans Clustering'}, xlabel='k', ylabel='distortion score'>
```

Clustering documents

We've determined the acceptable value for the clusters, so let's use Machine Learning to determine those clusters. We'll use the classic KMeans algorithm to do this.

```
In [26]: clusterer = KMeans(12 if visualizer.elbow_value_ > 12 else visualizer.elbow_value_)
clusterer.fit(wv_df)
clusters = clusterer.labels_
```

We'll perform a quick visual check on the clusters. In order to be able to visualize 128

dimensions (which is the size of the word vectors) in a 2-D space, we'll use the PCA (Principal Component Analysis) dimensionality reduction technique. This will transform our 128-dimensional vectors into 2-dimensional ones that we can display.

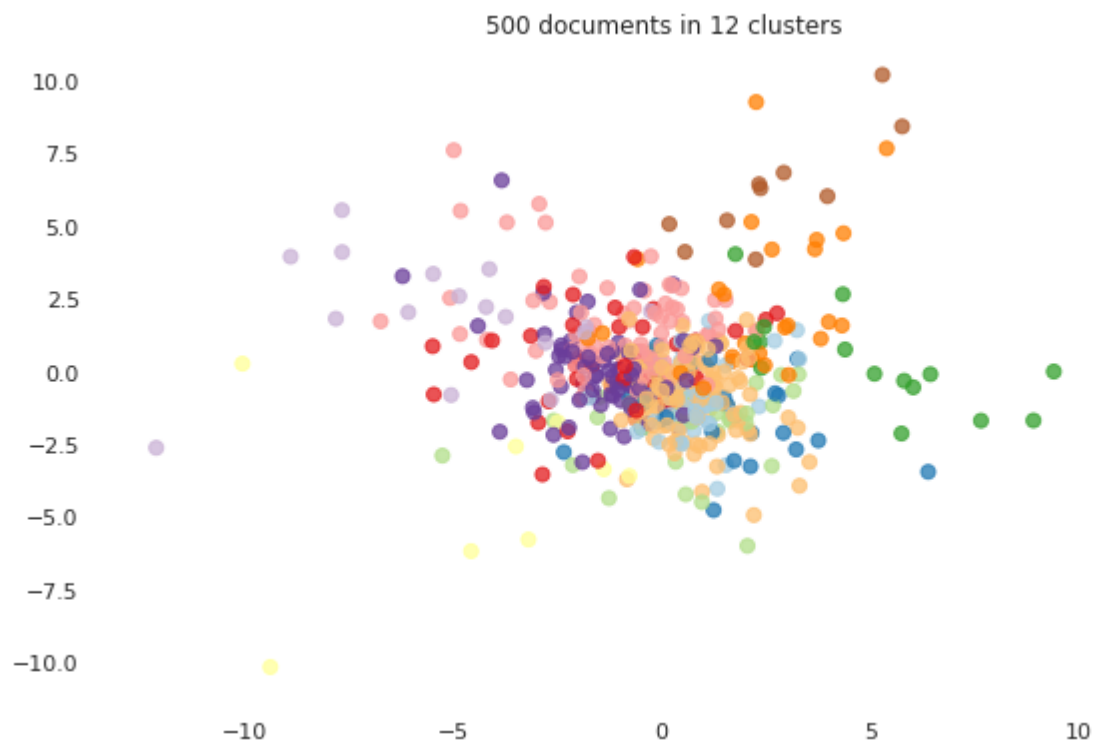
```
In [27]: pca = PCA(n_components=2)
pca.fit(wv_df)
result = pca.transform(wv_df)
```

Afterwards, we can plot the documents in a simple 2-D chart, and color each one according to their cluster

```
In [28]: sns.set(rc={'figure.figsize':(10, 6), 'figure.facecolor':'white', 'axes.facecolor':'white'})

color_palette = sns.color_palette('Paired')
# Each cluster gets its own color from the palette
cluster_colors = [color_palette[x] if x >= 0 else (0.5, 0.5, 0.5) for x in clusterer.labels_]
plt.scatter(result[:,0], result[:,1], s=50, c=cluster_colors, alpha=0.75)

plt.title(f'{N_DOCUMENTS} documents in {clusterer.n_clusters} clusters')
plt.show()
```



We'll add each article's cluster as new column to our combined dataset

```
In [29]: metadata_with_clusters = metadata_with_embeddings
metadata_with_clusters['cluster'] = clusters
metadata_with_clusters
```

Out[29]:

	sha	source_x	title	
ug7v899j	d1aafb70c066a2068b02786f8929fd9c900897fb	PMC	Clinical features of culture-proven Mycoplasma...	10.1186/1471-22
02tnwd4m	6b0567729c2143a66d737eb0a2f63f2dce2e5a7d	PMC	Nitric oxide: a pro-inflammatory mediator in l...	10.11
ejv2xln0	06ced00a5fc04215949aa72528f2eeaae1d58927	PMC	Surfactant protein-D and pulmonary host defense	10.11
2b73a28n	348055649b6b8cf2b9a376498df9bf41f7123605	PMC	Role of endothelin-1 in lung disease	10.11
9785vg6d	5f48792a5fa08bed9f56016f4981ae2ca6031b32	PMC	Gene expression in epithelial cells in respons...	10.11
...	
v6jyplcn	ba1a74766e0d96756105e85ce014b4bbe86fe819	PMC	Immune responses against severe acute respirat...	10.1111, 2567.2007.
fae3sczm	00acd3fd31ed0cde8df286697caefc5298e54df1	PMC	Distinguishing Molecular Features and Clinical...	10.1371/journal.pone.0
1c4m2fym	808dfc4c59f3e2e9150aa5542ea227718741388b	PMC	Towards a Coronavirus-Based HIV Multigene Vaccine	10.1080/17402520600
1ldebnq8	9cf43ad7346d9099015ee0545e0b3498c10213b4	PMC	Epithelial Cell Apoptosis and Neutrophil Recru...	10.2119/2008-000
x7wva1ax	3ccc18758132523e4b94a6cd392053a5b4b542cb	PMC	The cucumovirus 2b gene drives selection of in...	10.1093/nar/gl

We can now split our data into two datasets - a **training** one that will be used to train a Machine Learning model, able to determine the cluster that should be assigned to an article, and a **test** one that we'll use to test this classifier.

We will allocate 80% of the articles to training the Machine Learning model, and the remaining 20% to testing it.

```
In [30]: train, test = train_test_split(metadata_with_clusters, train_size=0.8)
train
```

Out[30]:

	sha	source_x	title	
ujhgb3b0	85c04886b69b27eeddd4a72c74e2e1727a80598f	PMC	CoVDB: a comprehensive database for comparativ...	10.1093/na
apiz60hu	9aa51dd5b5755afa670398027272f2e6b4f3d83b	PMC	Adaptive evolution of the spike gene of SARS c...	10.1186/1471-
icj261ls	6a3c7595bc8fa30e3f74fd3d350d7cb7e3a1668b	PMC	The three transfer RNAs occupying the A, P and...	10.1093/na
1fu1blu0	4b43f61d164be997a34343c11c70c42edd91898b	PMC	Distributed data processing for public health ...	10.1186/1471-2
wrrhr237o	275d7a25357ea46699072167d4ca109108132068	PMC	Surveillance recommendations based on an explo...	10.1186/1471-2
...
6lvn10f4	14e0cac6e86d62859e6c9f1351ab67466d89e5b3	PMC	Heterogeneous nuclear ribonucleoprotein A1 reg...	10.1093/emboj/1
zzkkm496	c4cbc868c6aa292decf1296da0f74eb496cca7ac	PMC	Outcome of paediatric intensive care survivors	10.1007/s00431-0
wutnzzhg	017ca5bdac589a37196df7b8e077c4c371ab32da	PMC	Pro/con clinical debate: Isolation precautions...	10.11
p34ezktf	d77233dd9085c9c76787c84ee9a1e249424b283a	PMC	Australian public health policy in 2003 – 2004	10.1186/1743
mqnqjn0c	3899c3b15d80afef1944752dafbebaa99ee43187	PMC	A Mouse-Adapted SARS-Coronavirus Causes Diseases...	10.1371/journal.ppa

400 rows × 147 columns

To speed up training, we'll ignore all columns except the word vectors calculated using Doc2Vec. For this reason, we will create a separate dataset just with the vectors.

```
In [31]: columns_to_ignore = ['sha', 'source_x', 'title', 'doi', 'pmcid', 'pubmed_id', 'license',
                             'who_covidence_id', 'arxiv_id', 'pdf_json_files', 'pmc_json_files']
train_data_vectors = train.drop(columns_to_ignore, axis=1)
test_data_vectors = test.drop(columns_to_ignore, axis=1)
```

Register the training and testing datasets for AutoML availability

We're registering the training and testing datasets with the Azure Machine Learning datastore to make them available inside Azure Machine Learning Studio and Automated ML.

```
In [32]: # Retrieve your ML workspace
ws = Workspace.from_config()
# Retrieve the workspace's default datastore
datastore = ws.get_default_datastore()

Dataset.Tabular.register_pandas_dataframe(train_data_vectors, datastore, 'COVID19Articles_Train_Vectors')
Dataset.Tabular.register_pandas_dataframe(test_data_vectors, datastore, 'COVID19Articles_Test_Vectors')
```

```
Validating arguments.
Arguments validated.
Successfully obtained datastore reference and path.
Uploading file to managed-dataset/4569b189-49fa-43e9-91d9-63be7d5b4c30/
Successfully uploaded file to datastore.
Creating and registering a new dataset.
Successfully created and registered a new dataset.
Validating arguments.
Arguments validated.
Successfully obtained datastore reference and path.
Uploading file to managed-dataset/51c93458-f386-42af-83f8-1da164289585/
Successfully uploaded file to datastore.
Creating and registering a new dataset.
Successfully created and registered a new dataset.
```

```
Out[32]: {
  "source": [
    "('workspaceblobstore', 'managed-dataset/51c93458-f386-42af-83f8-1da164289585/')'
  ],
  "definition": [
    "GetDatastoreFiles",
    "ReadParquetFile",
    "DropColumns"
  ],
  "registration": {
    "id": "d164de0e-93b2-49c1-b860-9b1ff0e72546",
    "name": "COVID19Articles_Test_Vectors",
    "version": 1,
    "workspace": "Workspace.create(name='ai-in-a-day-800633', subscription_id='b8fce007-8472-482f-92d6-2278c8aab6a9', resource_group='ai-in-a-day')"
```

Open Azure Machine Learning Studio

Return to the GitHub repo and follow the instructions from there. You will use Automated ML in Azure Machine Learning Studio to train a classification model that predicts the document cluster for new research articles.