# HANDOVER

#### Foreword

There are 3 models I researched on within the last 6 months of my work at GM&T.

All 3 models are contained in the library GMT-FOM-GAS-RESEARCH which can be found at GM&T tfs repository https://tfs/tfs/Shared/FOM/\_git/gmt-fom-gas-research.

Please see this page: GMT-FOM-GAS-RESEARCH for more information about the software you need to have installed on your PC in order to use this library.

The library employs a Python-based task-scheduler which runs the below mentioned models on a schedule.

An example of how to run the scheduler can look like this:

C: set PATH=%PATH%;C:\Users\ashubert\Anaconda3\condabin call conda activate gas\_research set PYTHONPATH=C:\gir\gmt-fom-gas-research cd C:\gir\gmt-fom-gas-research python -m main pause

All three models have corresponding folders at \\trading1\Common\\gasmodels

Some of them also have the corresponding tables at GasMongo Mongo Db, pls. see further: GasMongo Database collections | GasModels Mongodb Connection Details

Notes on the library code structure:

- gmt-fom-gas-research\jupyter -this folder contains exemplar Jupyter Notebooks which demonstrate the way the models can be called
- gmt-fom-gas-research\logs this folder contains txt-based logs
- gmt-fom-gas-reseach\sandbox this folder contains a number of subfolders, the idea here was for every analyst to have a dedicated subfolder where he/she can keep all the research-based notebooks/scripts

#### Git Storage Model

This model is concerned with the optimal amounts of storage injections/withdrawals and was borrowed from https://github.com/cmdty/storage and https://pypi.org/project/cmdty-storage/0.1.0/

The two Python packages, that need to be installed are cmdty\_storage and curves (using pip).

All numerical computations are done in C# with Python code wrapped around it.

The documentation/papers describing the model are stored at: \trading1\Common\gasmodels\git\_storage\papers

The model has not yet been put into production, rather it's in a pre-production (job-based) mode. As such we've organized to put all the files needed at \\trading1\Common\gasmodels\git\_storage\production.

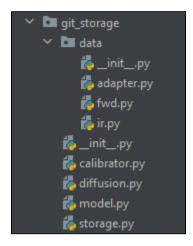
In there

- calibrator.json contains all the necessary parameters for doing the LSMC regressions (most likely it won't require any changes)
- diffusion.json contains the values for mean reversion, and 3 volatilities, which defined the SDE of the forward diffusion (this needs to be either calibrated to Big Bird model or being expressed as traders' view)
- storages this folder contain all the definition of the storages. Each storage needs to have a separate file and contain exactly the same columns as the other files. The storages can be of simple or ratchet type

The Python code in the GMT-FOM-GAS-RESEARCH (see above) library is organized as follows:

- gmt-fom-gas-research\gmt\fom\gas\shared\models\git\_storage contain all the model related scripts
- gmt-fom-gas-research\gmt\fom\gas\shared\jobs\git\_storage.py contains the job (an example of how the model can be called), which is subsequently run in main.py on a schedule.

The code structure is as follows:



Please also have a look at the gmt-fom-gas-research\jupyter\Git Storage Model.ipynb Jupyter Notebook for an example of how the git-storage model can be used in the notebook, you can also see there a method for how the results can be retrieved from MongoDB. Short Term Power Burn Model

The Short Term Power Burn Model is an attempt to forecast the amount of gas is used to produce electricity (in the UK). Please have a look at this page Short term power burn forecasting for more details about the project.

As an outcome the research phase we've decided to use SARIMAX (1,1,1) x (1,0,1,7) model with the external covariates to be weather (temperature, precipitation, wind) and dark spreads.

All the research-related code can be found in the GMT-FOM-GAS-RESEARCH library (see the foreword above) in the folder sandbox/alexs /short\_term\_power\_burn. In there you can also find various candidate models that we've tried.

The model has a dedicated server folder: \\trading1\Common\gasmodels\short\_term\_power\_burn

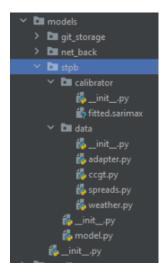
#### Data

There are two regimes you can run the model: calibration or forecasting.

In the calibration regime you need to provide actual (historical) values for all the covariates and (!) dependent variable. In the forecasting regime you need to provide forecasted values for all the covariates.

- weather:
  - historical values were taken from an external api and saved into the csv files. These files contain temperature, precipitation, wind
    values for a number of the cities in the UK
  - forecasted values were implemented in ANGARA and need to be retrieved from there (this hasn't been yet implemented)
- spreads
  - historical values are supposed to come from ARC, however we didn't manage to reconcile the output of the the underlying SQL query and the results retrieved using ARC Python api, thus saved the SQL query results into the csv file
  - forecasted values are also supposed to come from ARC (this hasn't been yet implemented)
- ccgt (dependent variable)
  - historical values were provided by Chris Arnold. This time series is only needed for the calibration regime

### Code structure:



The calibrator folder contains fitted.sarimax file which is the file produced by the `statsmodels` Python package. The idea is in case the user doesn't need re-calibrate the model, she can re-instantiate the model from this file and produce forecasts. Please bear in mind however, that that `statsmodels` package is using `pikle` Python package underneath to create such a file, and it might not work correctly once run on a new PC. Therefore you might need to re-calibrate the model first prior to running forecasts.

The model is linked to a Python-based scheduler. The job, written in the gmt-fom-gas-research\gmt\fom\gas\shared\jobs\stpb.py file provides a clear an concise demonstration of how to run the model (both in the calibration and forecasting regimes).

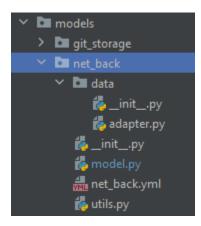
NetBack script

The netback script is not a statistical model per-se, rather an attempt to productionalize the Python script developed by Chris Arnold. Please have a look on Netback model (LNG) for a bit more info around the project.

The script takes two sets of data: costs and variables. They come in a csv file and were provided by Chris Arnold. Subsequently they were transferred into MongoDB collections.

The code body of the model is the set of transformations of these two data-sets(costs and variables) enriched by the data retrieved from ARC (a number of curves).

Code structure:



The adapter.py file contains DataAdapter class which handles all operations with in and out MongoDB.

The model.py file contains main NetBackModel class as well as NetBackSymbols class.

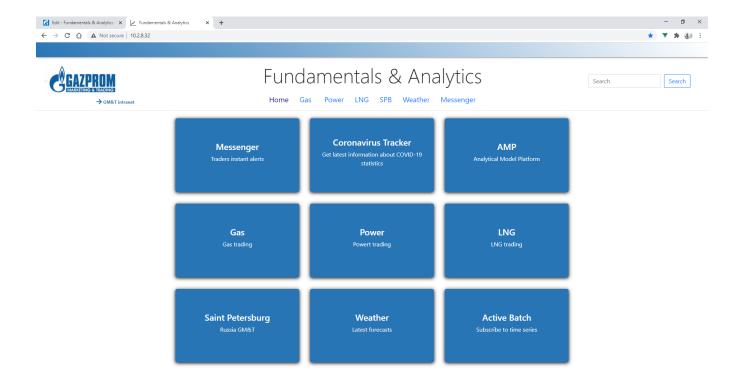
Please note the Mapping class, which is embedded into NetBackModel class. Given a vast amount of constants used in the script, we've decided to collect them all in this Mapping class.

The script is also linked to a Python-based scheduler. The job, written in the gmt-fom-gas-research\gmt\fom\gas\shared\jobs\net\_back.py file provides a clear an concise demonstration of how to run the model. The results are stored in the MongoDB (GasMongo) and can be easily retrieved by using this Jupyter Notebook: gmt-fom-gas-research\jupyter\NetBackModel.ipynb.

FOM Web-portal And Trading Alert Messenger

DECOMISSIONED: FOR THE CENTRAL IT TO DEVELOP

Interface



#### Structure

The portal is powered by the Python (Fast API) backend with the UI written in Vue.js (2x).

Repository: https://tfs/tfs/Shared/FOM/\_git/gmt-fom-gas-dashboard

# Folder structure:

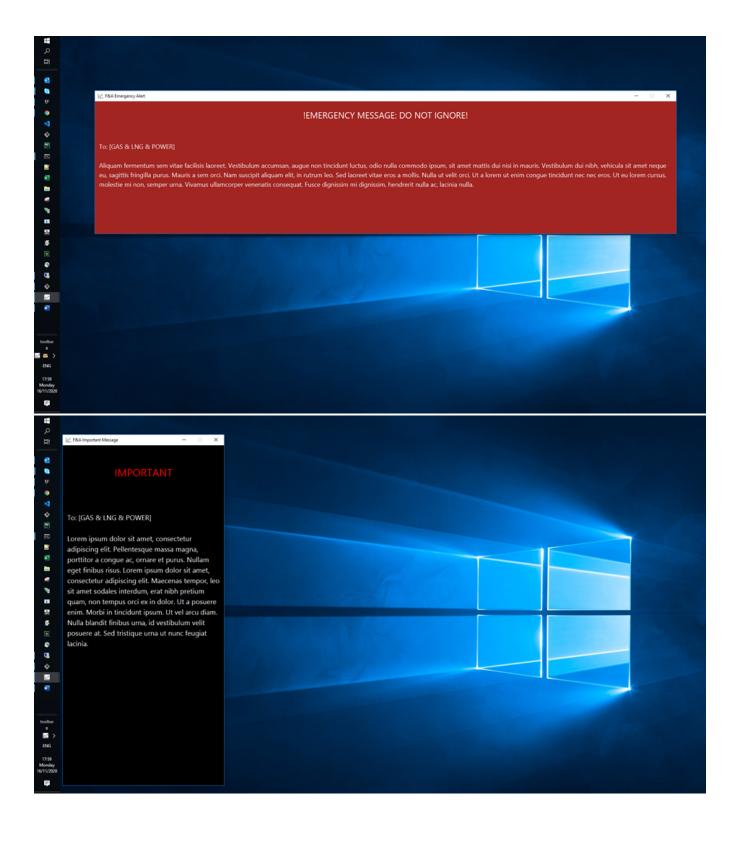
- gmt-fom-gas-dashboard
  - gmt
    - fa
- lib (for auxiliary python functionality)
- ui (web-portal)
  - fstapi (web-portal python backend)
  - vue (web-portal Vue frontend)
- java
  - dashboards
    - messenger (java-based messenger application)
    - Deployment (java classes to be deployed)
    - Install (files to produce setup.exe 7z self-extracted archive)

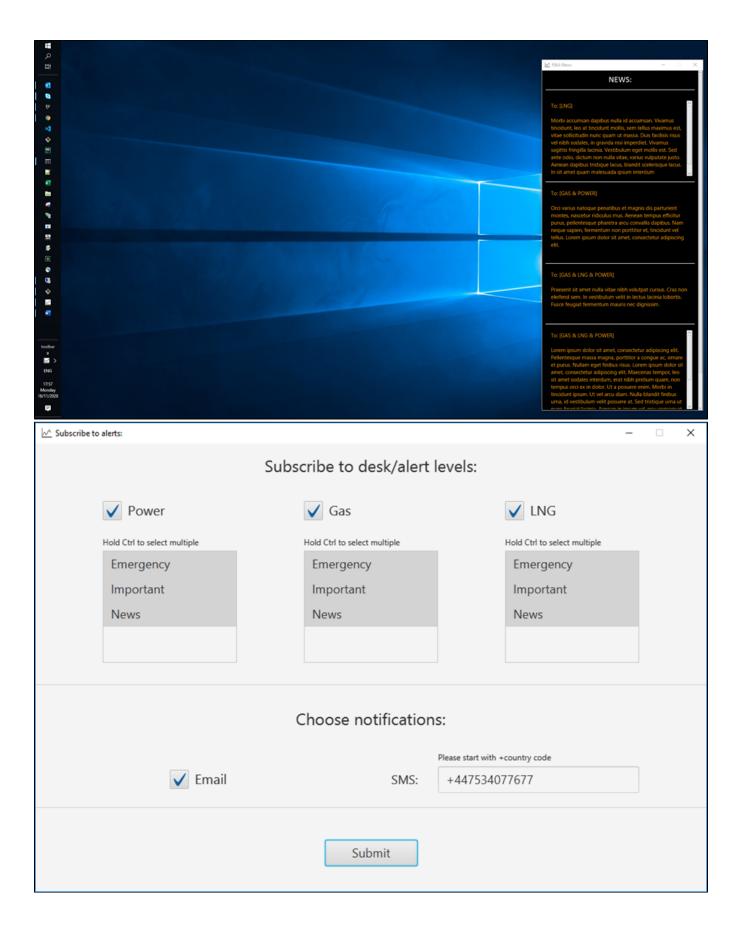
# Server structure:

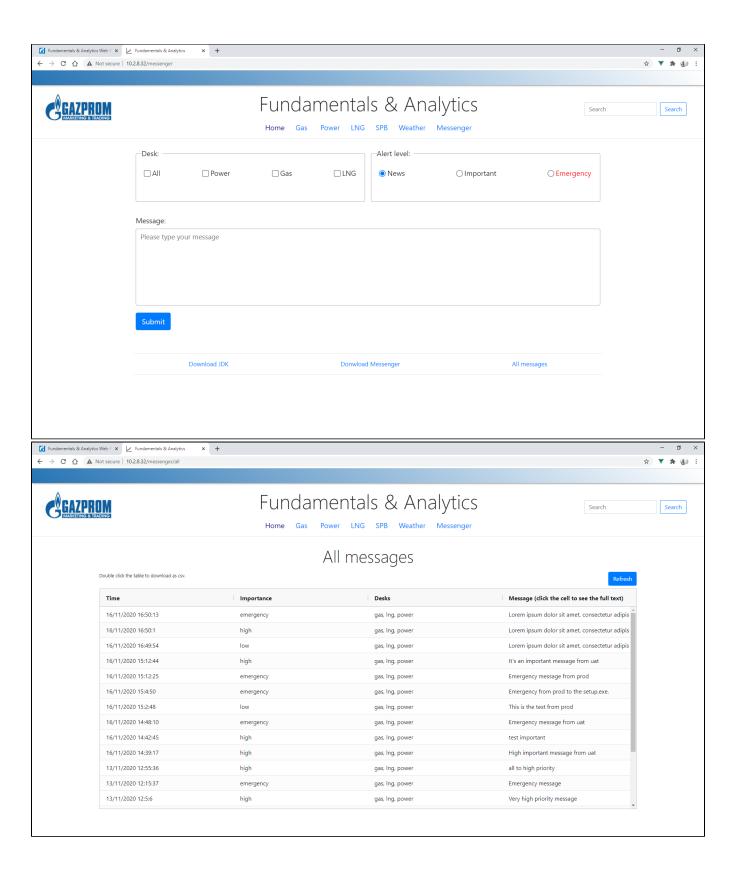
The project is still under development and resides at the gmt00526.gazpromuk.intra

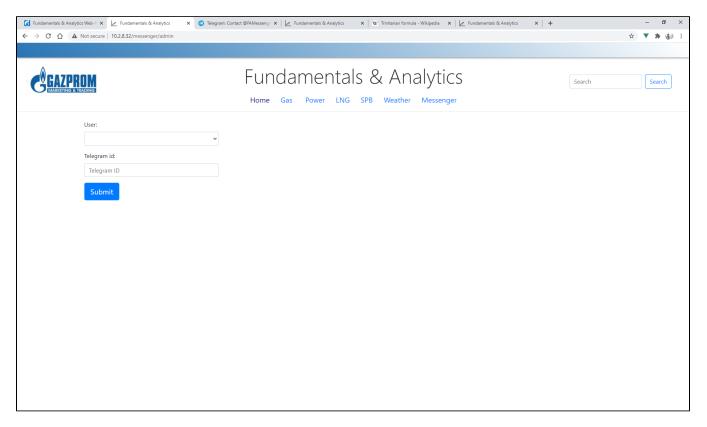
- DEV\_PROD: http://10.2.8.32/
- DEV\_UAT: http://10.2.8.32:8081/

Trader Messenger App Interface

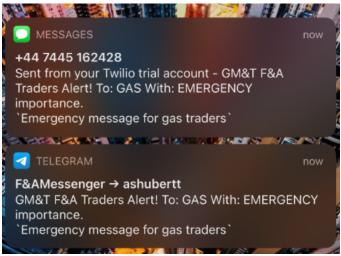








# SMS and Telegram Notification:



Send message All messages Subscribe Show Logs

### Functionality

When launched for the first time, the user is presented with the subscription window, where she can subscribe to different alert levels for different desks. Subsequently she can subscribe to receive email and/or SMS. In the present configuration, the email and SMS are sent for Important and Emergency alert levels. The user can also opt to receive messages in the Telegram bot.

When the new message arrives it's either shown in Emergency window (which is displayed on top of all windows on *all monitors*), or Important window (which is displayed on top of all windows on the *main* monitor), or pushed as a stack onto the News window (the news window is not going to steal focus, but the standard window alert will be displayed inviting the user to check his news feed). If the user tries to close News window it will be minimized to the taskbar. In order to close News window, the user has to unsubscribe from News alerts for all desks.

From the application tray icon context menu the user can open the web-page to send the message or the web-page to see all messages sent so far. She can also see the logs (to send to the support in case the program malfunction) and amend/create her subscription settings.

Python

Python backend: gmt-fom-gas-dashboard\gmt\fa\ui\fstapi\dashboard\messenger

UI (Web)



From within the app context menu

the user can be forwarded to the following pages on the F&A Web-portal:

- send message form
- all messages table

There is also a dedicated admin page to update user's telegram ID in case she opts to receive the messages also in the Telegram bot. UI (JAVA)

The source code consists of an Icon Tray class, 4 JavaFX windows (Emergency, Important, News, Subscribe); Config, Utility, SubscribeStore auxiliary classes and Dispatcher class which governs the flows in the application.

The applications establishes web-socket connection the Python back-end, which in turn broadcasts messages to the users (given their subscription settings) upon the message been sent from the web-page.

UAT Installation

Currently the JAVA 15 runtime is required, which can be downloaded as part of the JDK here.

The messenger setup.exe can be downloaded here.

Python Workshop

Foreword

In what follows I will be writing the sequence of commands using the arrow

#### For example:

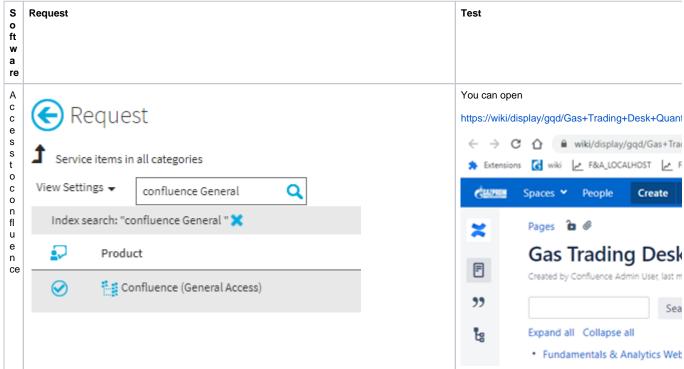


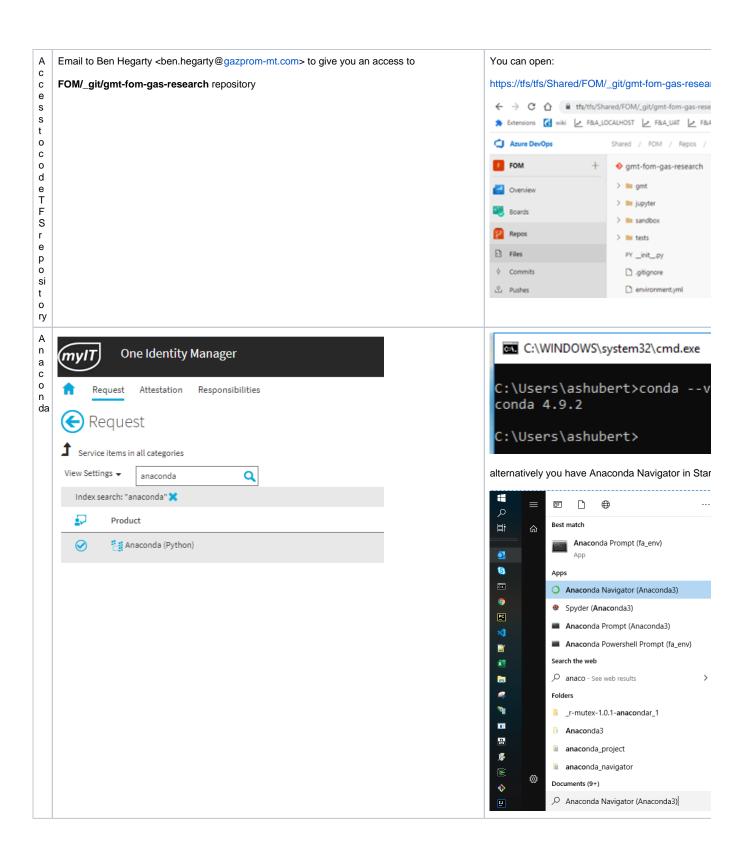
+ R cmd conda --version

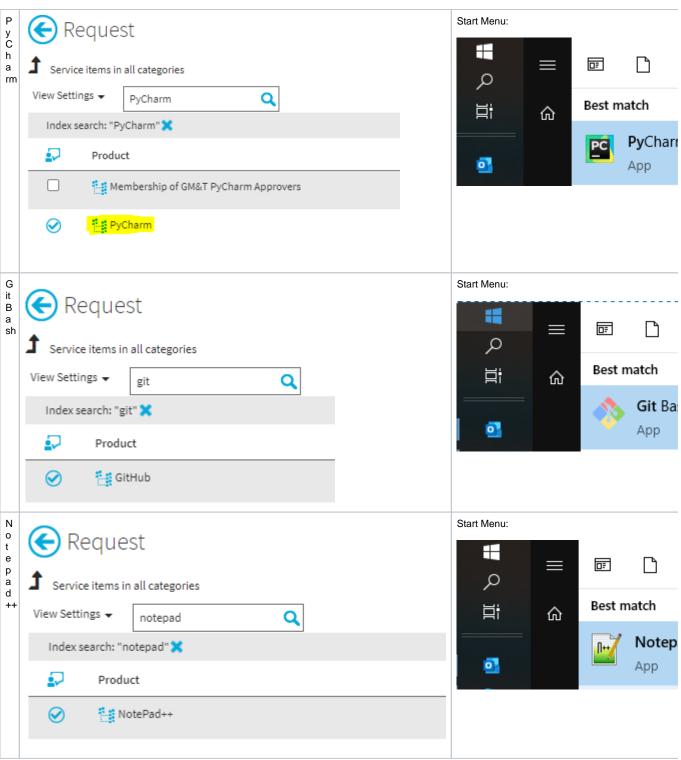
#### Means:

- Press button and R button (to open Run window)
- Type cmd (to open cmd window)
- Type conda --version

# Software To Install







Running Python

The standard Python distribution consists of an interpreter and a number of packages (standard library). However, for scientific computations the most popular Python distributive is Anaconda.

It might be already the case that you have Python installed at GM&T. To check it:

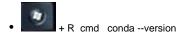
+ R cmd where python



In case it's not installed do the following:

- go to MyIT requests and start a new request
- in the search box type Anaconda
- once request is completed, go to Windows Start menu and find Software Center
- in Software Center go to Applications, find Anaconda and proceed with the installation

Once Anaconda is installed:



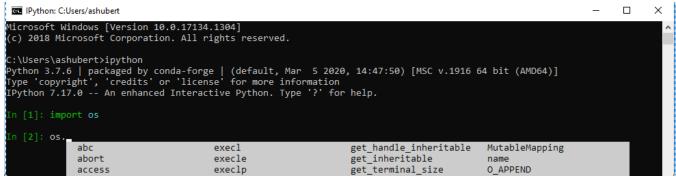


## Programming IDEs

There are a number of programming IDEs (Integrated Developer Environment) to make programming in Python as comfortable as possible:

# CMD:

- you can type your program directly in the **Python interpeter:** 
  - + R cmd python print("Hello World")
  - type exit() to exit Python
- a bit more user-friendly command line interface is **IPython** (interactive Python):
  - + R cmd ipython
  - now type: import os press enter, then type os. (mind the dot) and press Tab button you'll see IPython suggests further alternative



# Notepad:

You can use any simple text editor (but not MS Word) to type your Python code, including Windows Notepad

- in the folder let's say C:\temp open notepad ( + R notepad), type print("Hello, world!") and save as "myscipt.py"
- + R cd C:/temp python myscript.py



PyCharm:

- professional IDE developed by JetBrains. There are two versions: Professional and Community (free).
- to request go to MyIT requests and start a new request
- in the search box type *PyCharm*
- once request is completed, go to Windows Start menu and find Software Center
- in Software Center go to Applications, find Anaconda and proceed with the installation

#### Jupyter

- this is a web-based IDE which I would highly recommend for learning Python and prototyping
- it is a part of Anaconda Python distribution
- to test that you have jupyter installed run the following:



• to run jupyter notebook:



#### Anaconda

To be able to "preserve" the state of packages you used in your current application you need to create a virtual environment. Physically it's just a folder on your PC where all the packages are going to be saved.

Anaconda has it's own package manager `conda` which allows you to easily create a new virtual environment. In order to to do it type the following in



• conda create --name <name of the environment> python=3.7

Once it's done type:

· conda activate <name of the environment>

To see all the existing environments on your machine type:

conda info --envs

To deactivate the environment, type:

conda deactivate

To install a new package inside your environment, once the environment is activated, type (for e.g. pandas package):

· conda install pandas

To learn mode conda commands read Conda Cheat Sheet. Jupyter Notebook

- Keyboard Shortcuts
- Markdown Cheat Sheet
- Simple LaTex commands
- YouTube shortTutorial | longTutorial

Software Development Life Cycle (SDLC)

GIT

Introduction to git

Azure DevOps Server (TFS)

Microsoft Documenation

**AMP** 

Documentation WikiPage: AMP (Analytics Modelling Platform)

Python Lessons Internal Workshop:

Python Workshop: jupyter | html An example of a modelling class: jupyter | html

Exercises: html • Solutions: html

# **External resources:**

- YouTube Python Course
- Simply Python Tutorial

Python package-specific tutorials

- numpy
- scipy
- pandas
- matplotlib

- seabornstatsmodels
- scikit-learn
- prophettensorflowkeras

	Description	Files	Data Set	conda environment	Comment
	General time-series modelling framework in Python (based on ARIMAX model):				
	data retrieval				
	exploratory analysis and visualisations				
	data pre-processing     sample size				
	• imputations				
	• stationarity				
	transformations     design matrix				
	data-set partitions				
	• training				
	validation     testing				
	model selection				
	model parameters specification				
	heuristic     grid methods				
	forecasting				
	model performance metrics				
	cross-validation error     final model calibration				
	out-of-time performance				
	Short Term Power Burn Model		\\trading1\Common\gasmod els\short_term_power_burn		
	an example of a newly developed model for Gas Analytics		\data		
		Data pre-			
		processing notebooks			
	• data:	<ul> <li>Candidate</li> </ul>			
	dark, clean spark spreads and weather covariates	Models			
	<ul> <li>is coming from ARC DB and external api</li> <li>pre-processed and copied to Gas MongoDB</li> </ul>				
	, , , , , , , , , , , , , , , , , , , ,				
	• models:				
	Linear Regression     Support Vector (kernel) Regression				
	SARIMAX				
	• LSTM RNN				
	Additional nation				
	Additional notes:				
	This article clearly defines causality   correlation   forecasting, for it's essential to				
	clearly define the purpose of the research prior to conducting one.				
	Trying to formulate the ultimate target of any statistical research we can try (in the				
	first approach) to reduce it to the following two main tasks:  1. Choose the loss function we'd like to minimize				
	<ul> <li>This gives rise, among others, to least ordinary squares loss function,</li> </ul>				
	but also to a variety of penalized (regularized) loss functions such as				
	(pls see this article for explanation)  • ridge				
	• lasso				
	lasso     elastic net  2. Capture the functional relationship between the dependent variable and				
	lasso     elastic net  2. Capture the functional relationship between the dependent variable and covariates				
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	Iasso elastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model				
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	Isso Ielastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest:				
	Isso Ielastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest: Local regression (LOESS   LOWESS)				
	Isso Ielastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest:				
	I lasso elastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest: Local regression (LOESS   LOWESS) Polynomial regression Basis functions / splines regression Support Vector Regression				
	I lasso I elastic net  2. Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest: Local regression (LOESS   LOWESS) Polynomial regression Support Vector Regression Support Vector Regression Long Short Term Memory Recurrent Neural Network				
	I lasso elastic net  Capture the functional relationship between the dependent variable and covariates  This gives rise to the following cascade of modelling frameworks: Simple linear model General linear model Generalized linear model Generalized additive model  The following also might be of interest: Local regression (LOESS   LOWESS) Polynomial regression Basis functions / splines regression Support Vector Regression				

• Get Data from ARC database: jupyter | html