Emerging Technologies to Modernize AstraZeneca Pharmaceutical and Bio-processing Development & Manufacture for both small and Large molecules Intelligent Factory - Concept

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# The Development of Process Understanding and Control Strategy



Current Model Manual and Automated Experiments Some Automatic Data Capture Manual data entry to ELN

Data Systems: DATA MART D360 –IMed Data System e.g.Times Temps Mixing etc. e.g. Purity Polymorph Crystallinity

e.g. Dissolution Disintegration Stability etc.



### **Process Intelligence and CPV**



Data Captured in Discoverant, Engineering Systems & QA LIMS Systems



### **The Opportunity**

Can we digitise all our Development Data in order to be able to produce a " Digital Design Space"?

Can we create a virtual model of the manufacturing Process?

Can we use the process model to define the data capture requirements from the manufacturing information?

Can we combine this process data with Input Material and Product quality data to drive process optimisation?

Can we combine manufacturing & development data to strengthen the model?

Can we use A.I. to allow the process to learn?

Can we do this continuously, dynamically, and autonomously?



#### **API Control Strategy**



Can we digitise all our Development Data in order to be able to produce a "Digital Design Space"?

Hyper Link





Can we use the process model to define the data capture requirements from the manufacturing information?

Late Stage Development driven by risk assessment, experiments and clinical manufacture which build process understanding including relationships between inputs and outputs (cause and effect)

Single Risk assessment tool starting with DORAS, QRAS, SURE etc and leading to a completed CPV risk assessment matrix









#### **Innovate UK Project Overview**



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# Advanced Process Control/Model Predictive Control What does it mean?



Model Predictive Control (MPC) understands process constraints and complex process interactions:

- o Build multivariate correlation between variables and actuators, causes and effects
- Predict impact of known disturbances on operation
- Predict, Advise, Make co-ordinated moves on multiple actuators
- Exploit **all** opportunities to push quality / throughput close to constraint / consent







#### **Key Achievements**

- Proven ability to run for 5 days and 26% higher yield (vs batch equivalent) in continuous crystallisation => improved yield
- ✓ 2 fold reduction in span of PSD vs stirred tank reactor => consistent & higher quality
- Reduction in crystallisation time from 16 hour to 5 hours => intensified process
- ✓ Reduced manpower and waste requirements
- ✓ 'Dial a Particle' capability achieved
- ✓ Batch to continuous methodology developed for crystallisation
- ✓ Advanced Process Control capable of controlling both Crystallisation platforms



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# SELF OPTIMISING FLOW REACTOR (SOFR)

## An example from the lab

**Mubina Mohamed and Graeme Clemens** 



#### **SOFR Instrument**



- Fully autonomous experimentation ideal for DOE and kinetic profiling
- Large amount of experimentation at low starting materials/reagents cost
- Large data sets from multiple analytical techniques to give bigger picture of the reaction
- SOFR is a *tool within the flow toolbox* to facilitate quick reaction parameter screen and reaction understanding



#### **SOFR Instrument - What can it do**



#### **Concentrations Ramps**

To produce rapid kinetic profiles of reactions

#### DOEs

Perform DOE experiments in sequence exploring reaction space to find optimum reaction parameters

#### SNOBFIT

Online analysis with a feedback control loop, which uses the optimising SNOBFIT (Stable Noisy Optimisation by Branch and Fit) algorithm to keep generating new conditions until an optimum is reached



NB: The code is not limited to the above, and is able to accommodate different structured designs

#### Where do we stand with kit and code MATLAB code - requires a User Interface lot of manipulation to ✓ Connection of equipment accommodate reaction ✓ Monitor flow rates, system parameters temperature and pressure \_\_\_\_ Advion MS (single Quad) Liquid chromatography MS 60 nl LC 60 nl sample loop sample loop AstraZeneca inter Steel ----Abla Mara Dat Nes ----Knaur LC pump ----(Quench eluent) SyrDos AstraZene pumps (1,2 and 3) Jasco LC pump (MS eluent pump) React IR with ATR diamond accessory ligh temp micro treactor Nace 💌 TIC Base Peak Chromatogram Analog InputLC Signal Equipment set-up essential for success 2016 6 21 11 11 50 daty # 2016 6 30 14 41 28 daty # 2016 6 20 17 27 41 daty # 2016 6 20 14 51 43 daty Vaste Chromatogram 278.9 - 200 2016\_6\_23\_16\_36\_33.date 2016.06.23.15.35.35 RichB MubM 02 (MLR Data formatting in a visual manner for projects 2016 6 23 11 11 58.datx × 2016 6 30 14 41 28.datx × 2016 6 29 17 27 41.datx × 2016 6 29 14 51 43.datx × 2016 6 23 15 35 33.datx × CEPANAS 1111 HEHRS 2016\_6\_23\_16\_36\_33.datk 2016.06.23.15.35 Inline MS analysis (1 ÷. N=22, R2=0.747, RSD=12.8, DF=18, Q2=0.519, Confidence=0.9 chromatogram per minute) MS software requires further Inline PAT analysis (1 spectra understand and manual data every 30 secs) analysis at present PAT data (IR) requires expertise to analyse

#### **SOFR and Manufacturing – Where can we take it?**

Successful Manufactures Every Time

**FEDBACK** 

Good predictions are based on robust models, which are built on variable data sets. Models are updated each time more data is available.

More data = Better models

Robust Models

# Data

- Data from all manufacturing processes recorded
- Data about variable input materials
- Data from reaction screening experiments
- All the above can be accomplished in the lab using the SOFR

### The Intelligent Factory Vision: The Big Hairy Audacious Goal The BHAG.

We understand all the critical quality attributes of all our products

We have a virtual Process model for all our processes – a Digital Design Space

We combine development and commercial data to strengthen the model

The model is used to control manufacture and has the capability to learn

The quality of our products is controlled automatically and the processes are continuously, dynamically and autonomously öptimised.



### How do we deliver this

- This is a 5 year vision but we have some of these things in place already
  - Process Models
  - Continuous Autonomous optimisation
  - Continuous Processing Options
- We start with One product and maybe one set of CQAs
- We have some systems in place to build on





