

Modelling trade-offs between building energy and health

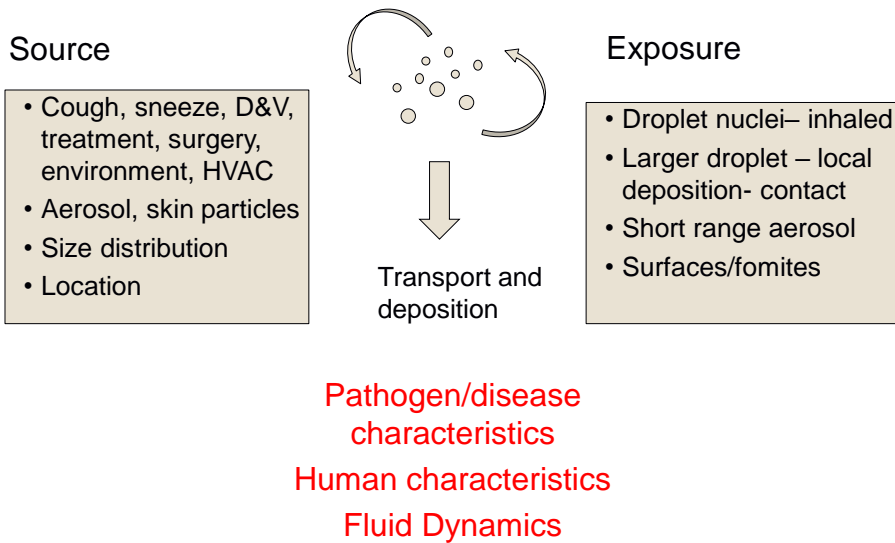
The challenge of airborne infection

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Outline

- ❖ Airborne infection
- ❖ Ventilation control – evidence, guidance, methods
- ❖ Methods/metrics
 - Air distribution
 - UV control
 - Risk –energy models
- ❖ Future needs



Break the chain of transmission – reduce/prevent source-susceptible route

Ventilation: Dilution, Air distribution, Pressure controls, Extraction, temperature and humidity

Technology: Filtration, air cleaning, decontamination, surface technology

Human behaviour: SMART buildings, prompt actions, prevent access

Guidance

- ❖ Country specific – e.g. UK and US are quite different
- ❖ Specification:
 - Nat or Mech vent
 - Air change rate
 - Pressure
 - Distribution – high risk only
- ❖ Why are these values recommended?
- ❖ What are the energy – comfort – risk trade-offs?

Application	Ventilation	AC/hr	Pressure (Pascals)
General ward	S/N	6	–
Communal ward toilet	E	6	–ve
Single room	S/E/N	6	0 or –ve
Single room WC	E	3	–ve
Clean utility	S	6	+ve
Dirty utility	E	6	–ve
Ward isolation room	–	–	–
Infectious diseases isolation room	E	10	–5
Neutropenic patient ward	S	10	+10
Critical care areas	S	10	+10
Birthing room	S & E	15	–ve
SCBU	S	6	+ve
Preparation room (lay-up)	S	>25	35
Preparation room/bay (sterile pack store)	S	10	25
Operating theatre	S	25	25
UCV operating theatre	S	25*	25

HTM03-01 Specialised Ventilation for Healthcare Premises, Department of Health (2007), UK

Evidence – and problems



- ❖ Ventilation rate
 - Primary metric for control/design
 - 2ACH for TB shown by infection data (Menzies et al 2000)
- ❖ Energy - 6ACH suggested as “too high” (Lomas & Giridharan 2012)
- ❖ Local concentrations are higher – risk to healthcare staff?
- ❖ Variations in mixing, particularly in large multi-bed rooms
- ❖ Displacement ventilation suggested problematic (Li et al 2011)
- ❖ Pressures compromised by activities
- ❖ UCV theatres validated empty – compromised by heat
 - New evidence suggests they may not be right (Allegranzi et al 2016)

What are we missing?

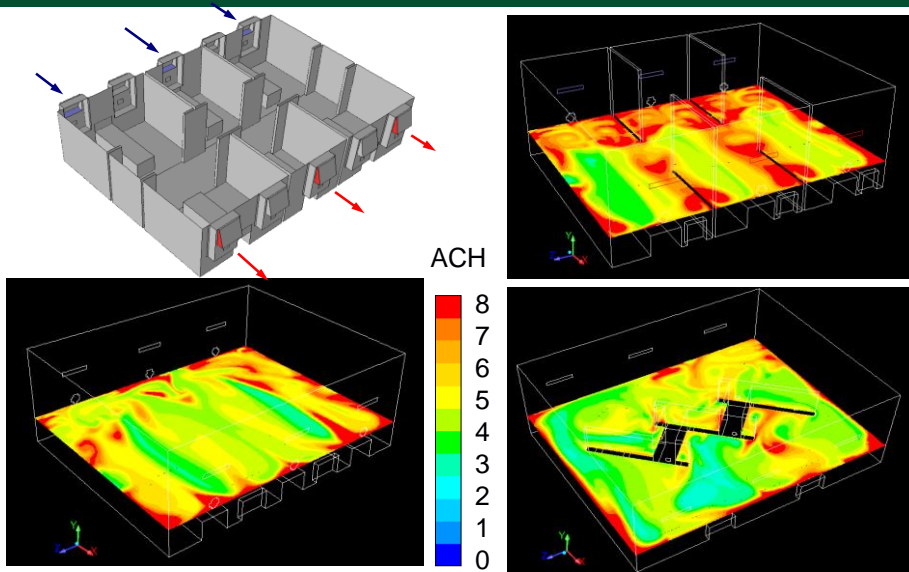
- ❖ Human influence – link between design and user
- ❖ Variability – heat load, season, source, movement, occupancy.....
- ❖ Connections – how one room affects another
- ❖ Risk assessment that deals with local needs – patient cohort, particular pathogen
- ❖ Ability to make evidence based decisions
- ❖ Ability to respond to outbreaks

Multi-occupant spaces

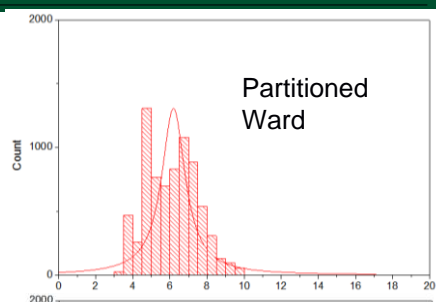
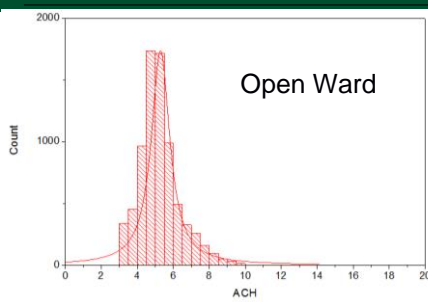


- Many hospital wards in UK are naturally ventilated with complex airflows and multiple openings
- Multi-bed wards still common – 2 to 8 people
- Infection risks are complex, variable, multi-route
- Challenge of privacy, comfort, energy

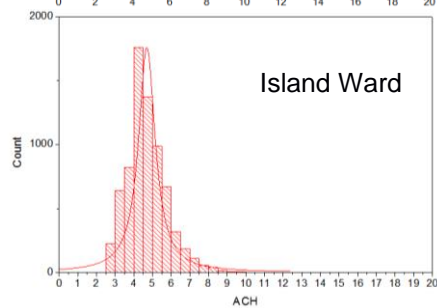
Air Change Rate



Air Distribution – 6 ACH supply



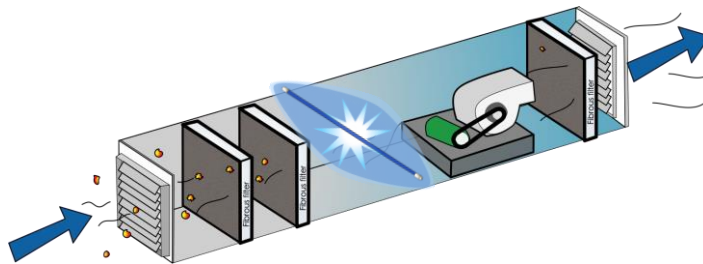
- Local ACH distribution on breathing plane
- Peak – above or below supply
- Spread – how much under or over ventilated



Quantifying UV disinfection

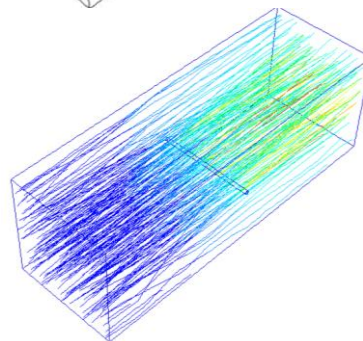
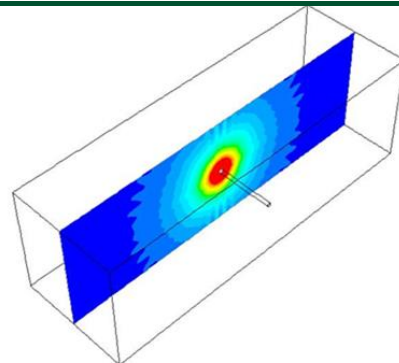
Depends on:

- Lamps – number, location, intensity
- Airflow – determines duration of UV exposure
- Microorganism susceptibility



CFD models

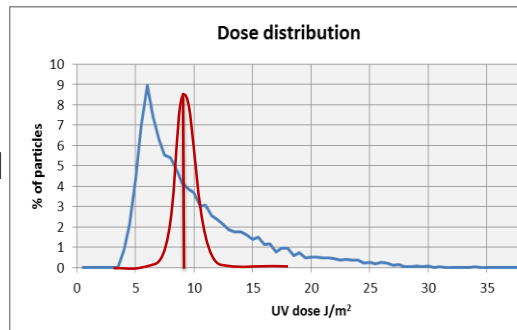
- ❖ Turbulent flow field representing duct
- ❖ Discrete ordinates UV field
- ❖ Custom models using Scalar/species or Lagrangian particle tracks
 - Dose = Irradiation*time – calculate from cumulative track through 3D UV field
 - Survival – empirical model based on experimental pathogen decay with time in UV field



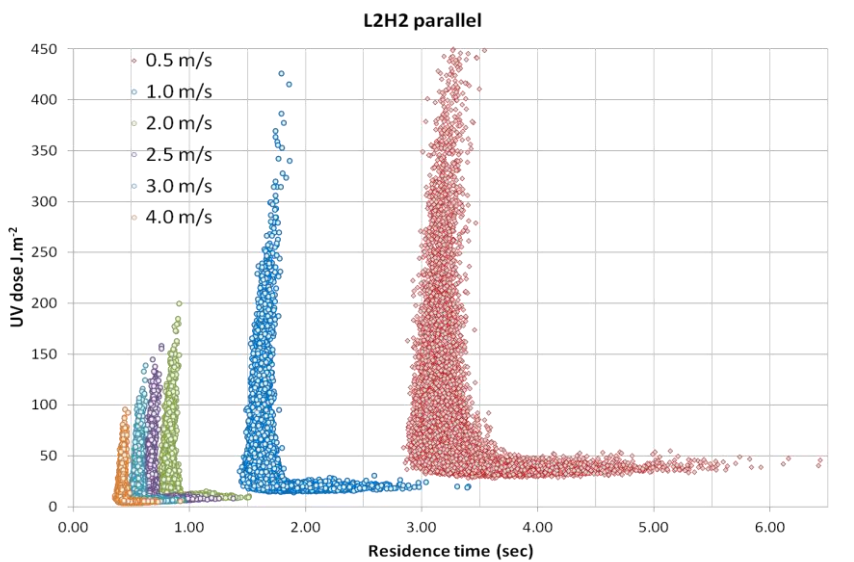
CFD outputs

- ❖ Mean CFD compares well to experiment
- ❖ CFD shows distribution
 - Some pathogens over irradiated
 - Many are under irradiated
- ❖ Inefficient and potential risk

Microorganism	EPA 600/R06/050 1 lamp 9.73 J/m ²	
	EPA	CFD
S. Marcescens	99%	99.46%
MS2	39%	34%
B. Atrophaeus	4%	8.72%



Design analysis



Wells-Riley Model

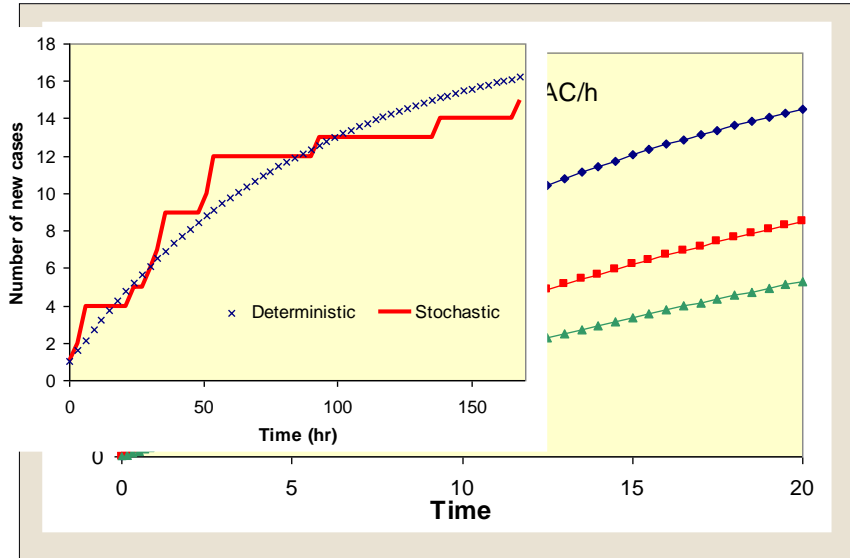
$$N_C = S \left(1 - e^{-\left(\frac{Iqpt}{Q} \right)} \right)$$

- ❖ Used in wide range of risk analysis studies and assessment of disease outbreaks
- ❖ Assumes well mixed airflow within a zone
- ❖ Relates new infections (N_C) with time (t) to disease, occupant and ventilation characteristics
 - S – number of susceptibles, I – number of infectors
 - Q – room ventilation rate
 - P – occupant breathing rate
 - q - Quanta- number of infectious doses generated per unit time

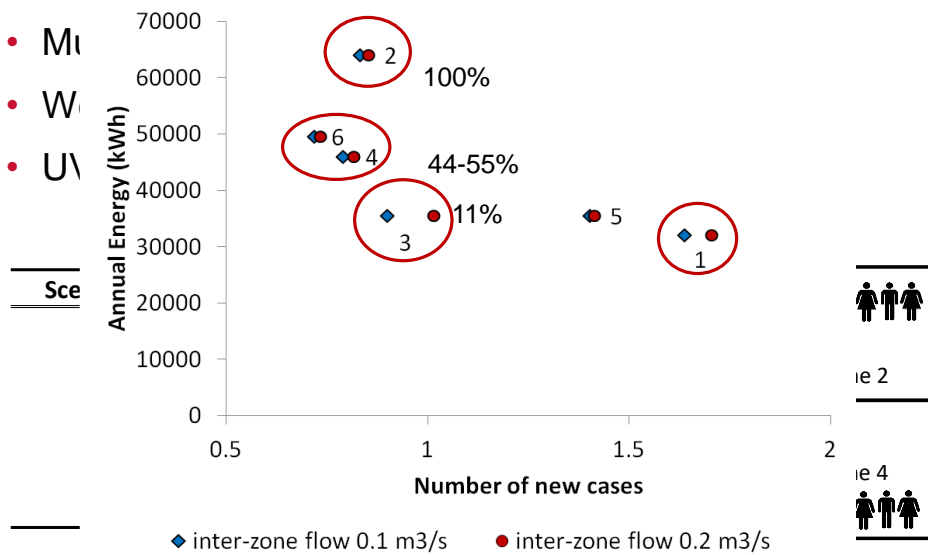
Infectious Dose (quanta)

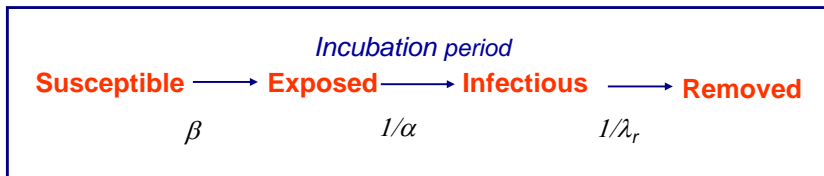
Disease	Case	Quanta/h	Reported by
TB	Average TB patient	1.25	Nardell <i>et al</i> (1991)
TB	Outbreak in office building	12.7	Nardell <i>et al</i> (1991)
TB	Human to guinea pig transmission	0.3-44	Escombe <i>et al</i> (2007)
MDR TB	Human to guinea pig transmission (highest infectors)	40,52,226	Escombe <i>et al</i> (2008)
Measles	Outbreak in a school	570	Rudnick & Milton (2003)
Influenza	School cases in Taiwan	66.91 (LN*)	Liao <i>et al</i> (2005)
Influenza	Aircraft outbreak	79-128	Rudnick & Milton (2003)
SARs	Taipei Hospital outbreak	28.77 (LN*)	Liao <i>et al</i> (2005)
Rhinovirus	Experimental data of Dick <i>et al</i> 1987	1-10	Rudnick & Milton (2003)

Wells-Riley Results

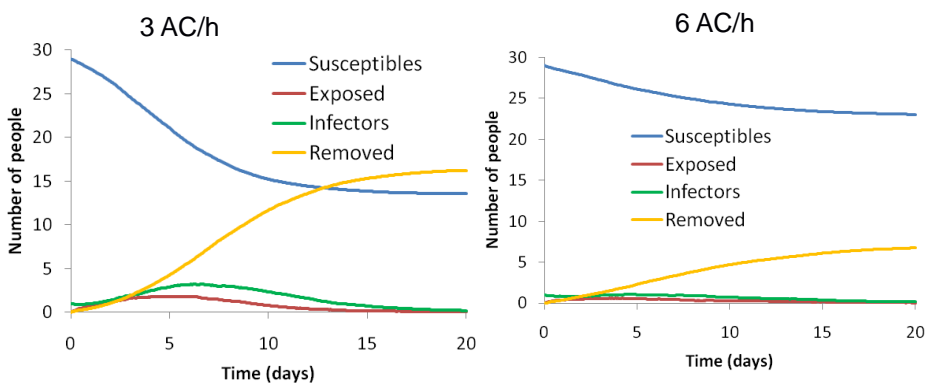


Coupling and connectivity





- ❖ SEIR model – well known epidemic model
- ❖ β determined by environment, infectious dose and breathing – Wells Riley model
- ❖ α and λ_r are disease parameters
- ❖ Examine effect of intervention and disease on progression of outbreak



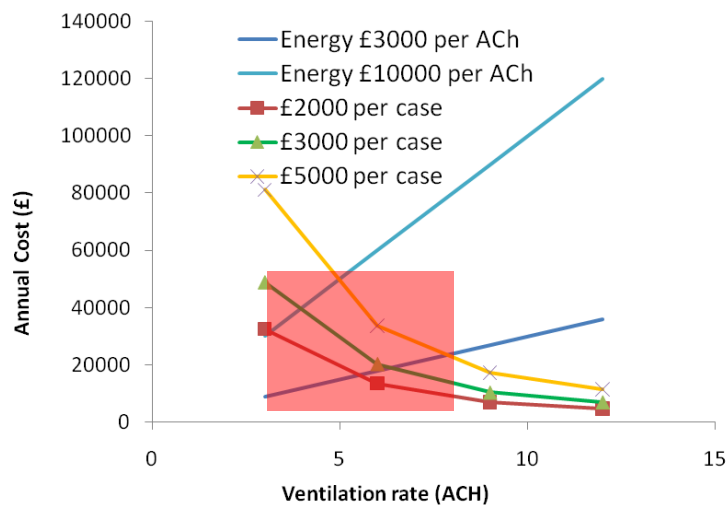
- ❖ Mean result from 500 simulations – deterministic
- ❖ Can determine expected number of cases for a given scenario

❖ Theory:

- Put a cost to providing ventilation
- Put a cost to dealing with cases of infection
- Assess the trade-off to establish the most cost effective approach

❖ Example case:

- 8 wards, each 1000 m³
- Average outbreak behaviour – one outbreak in a year
- Infection £2000 per case
- Energy £3000 per year per air change rate



Needs

- ❖ Probability based approaches – monte-carlo (e.g. QMRA) to deal with uncertainty
- ❖ Methods to directly relate risk and value/cost
- ❖ Flexibility to address local challenges
- ❖ Evidence from healthcare environments
 - Data from real environments – not just proxy/idealised models
 - Direct data on infections – very little of this
 - Cross-disciplinary challenge – funding!!
- ❖ Complex challenge that needs to enable robust engineering and operational decisions.

The future?

- ❖ HECOIRA –Hospital Environment Control, Optimisation and Infection Risk Assessment
- ❖ EPSRC funded Healthcare Impact Partnership grant
- ❖ Starts summer 2017 for 4 years
- ❖ Working with clinicians, healthcare estates, industry
- ❖ Tools for hospital design with a health/infection focus
 - Real-time sensor driven tools to adapt hospital environments
 - Quantitative pathogen exposure models
- ❖ Decision making, scenario testing, training

- ❖ Leeds academics, researchers, PhD students
 - Louise Fletcher, Andy Sleight
 - Marco-Felipe King, Carl Gilkeson, Amir Khan
 - Azael Capetillo

Thank You
Questions?