Testing theoretical minimal genomes using whole-cell models

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Abstract

The minimal gene set for life has often been theorised, with at least ten produced for

Mycoplasma genitalium (*M. genitalium*). Due to the difficulty of using *M. genitalium* in the lab,

combined with its long replication time of 12 - 15 hours, none of these theoretical minimal

genomes have been tested, even with modern techniques. The publication of the *M. genitalium*

whole-cell model provided the first opportunity to test them, simulating the genome edits

in-silico. We simulated eight minimal gene sets from the literature, finding that they produced

in-silico cells that did not divide. Using knowledge from previous research, we reintroduced

specific essential and low essential genes *in-silico*; enabling cellular division. This reinforces the

need to identify species-specific low essential genes and their interactions. Any genome

designs created using the currently incomplete and fragmented gene essentiality information,

will very likely require *in-vivo* reintroductions to correct issues and produce dividing cells.

Introduction

Genome engineering builds on historical gene essentiality research. The sequencing of small bacterial genomes (Fleischmann et al., 1995; Fraser et al., 1995) led to comparative genomics; then, as genome sequences completed increased, minimal gene sets (Forster & Church, 2006; R. Gil et al., 2004; Rosario Gil, 2015; J. I. Glass et al., 2006; Huang et al., 2013; Hutchison et al., 1999; Karr et al., 2012; Mushegian & Koonin, 1996; Shuler et al., 2012; Tomita et al., 1999) were hypothesised. Minimal genomes are reduced genomes where no single gene can be removed without loss of reproduction (John I. Glass et al., 2017), given an appropriately rich medium and no external stresses, and focusing solely on protein-coding genes. For a recent review of gene essentiality, see Rancati *et al.* (Rancati et al., 2018).

M. genitalium is the focal point of minimal gene set creation due its naturally small genome size (0.58mb and 525 genes) and early sequenced genome (Fraser et al., 1995). Minimal gene sets are designed using three different approaches: protocells, comparative genomics, and single gene knockouts. Protocell designs (Dzieciol & Mann, 2012) are not expected to function as true biological cells, instead functioning as a self-replicating, membrane-encapsulated collection of biomolecules (Forster & Church, 2006). Comparative genomics (Koonin, 2003) compares multiple species to identify common genes. This is complicated (John I. Glass et al., 2017; Lagesen et al., 2010) by non-orthologous gene displacements, i.e. independently evolved or diverged proteins that perform the same function but are not recognisably related (John I. Glass et al., 2017; Mushegian & Koonin, 1996), which can result in the removal of a large number of genes essential to one species. Design using single gene essentiality classifications should, in theory, not remove any essential genes; but if transposon mutagenesis is used, variance from different transposon variants, antibiotic resistance genes, and growth periods can result in differing essentiality classifications (J. I. Glass et al., 2006; Juhas et al., 2011).

Ten minimal gene sets were found in the literature that were designed with *M. genitalium* genes

(Forster & Church, 2006; R. Gil et al., 2004; Rosario Gil, 2015; J. I. Glass et al., 2006; Huang et

al., 2013; Hutchison et al., 1999; Karr et al., 2012; Mushegian & Koonin, 1996; Shuler et al.,

2012; Tomita et al., 1999). Due to the difficulty of using *M. genitalium* in the lab (Reich, 2000),

combined with its long replication time of 12 - 15 hours (Benders et al., 2010; John I. Glass et

al., 2017; Hutchison et al., 2016), none of these theoretical minimal genomes have been tested,

even with modern techniques (Benders et al., 2010). The publication of the *M. genitalium*

whole-cell model (Karr et al., 2012) provided the first opportunity to test them, simulating the

genome edits *in-silico*.

The *Mycoplasma genitalium* whole-cell model (Karr et al., 2012) was the first existing model of a

cell's individual molecules that includes the function of every known gene product (401 of the

525 *M.genitalium* genes), making it capable of modelling genes in their genomic context.

Previously, it has been used to investigate discrepancies between the model and real-world

measurements (Karr et al., 2012; Sanghvi et al., 2013), design synthetic genetic circuits in the

context of the cell (Purcell et al., 2013), make predictions about the use of existing antibiotics

against new targets (Kazakiewicz et al., 2015), and produce *in-silico* minimal genomes

(Rees-Garbutt et al., 2020).

Of the ten sets, two of the sets (Gil *et al.* (2004)(R. Gil et al., 2004)and Shuler *et al.(Shuler et al.,*

2012)) were excluded as they were considered derivative of the Gil (2014) set (Rosario Gil,

2015); four genes differ in the Shuler *et al.* set (MG_056, MG_146, MG_388, MG_391) and four

genes are absent in the Gil *et al.* (2004) set (MG_009, MG_091, MG_132, MG_460). Of the

other eight sets, two (Tomita *et al.* (Tomita et al., 1999) and Church *et al.* (Forster & Church,

2006)) were designed as protocells, three (Mushegian and Koonin (Mushegian & Koonin, 1996),

Huang *et al.* (Huang et al., 2013), and Gil (R. Gil et al., 2004; Rosario Gil, 2015)) from

comparative genomics, and three (Hutchison *et al.* (Hutchison et al., 1999), Glass *et al.* (J. I.

Glass et al., 2006), and Karr *et al.* (Karr et al., 2012)) from single gene essentiality experiments.

We simulated eight minimal gene sets from the literature (Forster & Church, 2006; Rosario Gil,

2015; J. I. Glass et al., 2006; Huang et al., 2013; Hutchison et al., 1999; Karr et al., 2012;

Mushegian & Koonin, 1996; Tomita et al., 1999), testing them for the first time. This was to

assess their functionality as extant and untested genome designs, but also to produce new

data for designing genomes to evaluate the differing approaches of focusing on

species-specific data or focusing on identifying core genes and common functions across

species.

Results

We adapted eight minimal gene sets for simulation within the *M. genitalium* whole-cell model.

To increase clarity, we renamed the sets after the main location where the set was constructed

(Table 1). The Bethesda set directly compared *M. genitalium* and *Haemophilus influenzae*

genomes (gram-positive and gram-negative bacteria, respectively) (Mushegian & Koonin, 1996).

The Rockville set applied global transposon mutagenesis to *M. genitalium in-vivo* to identify

non-essential genes (Hutchison et al., 1999). The Fujisawa set constructed an *in-silico*

hypothetical cell from 127 *M. genitalium* genes using the E-Cell software (Tomita et al., 1999).

The Rockville 2 set reapplied global transposon mutagenesis *in-vivo*, isolating and

characterising pure clonal populations (J. I. Glass et al., 2006). The Nashville set listed 151

E.coli genes (compared to *M. genitalium* genes within the paper) theorised to produce a

chemical system capable of replication and evolution (Forster & Church, 2006). The Stanford

set was the result of *in-silico* single gene knockouts using the *M. genitalium* whole-cell model

(Karr et al., 2012). The Guelph set compared 186 bacterial genomes (Huang et al., 2013),

whereas the Valencia set compared *M. genitalium* with genetic data of five insect

endosymbionts (Rosario Gil, 2015). The Nashville, Fujisawa, and Stanford sets were

unchanged, but the others had between 6 and 44 genes removed (Table 1) either because the

genes were unmodelled (the genes' function were unknown (Karr et al., 2012)) or genes had

duplicate entries.

Table 1. *Minimal Gene Sets from the literature.*

Minimal gene set	Code name	Design approach	*In-vivo* genome design size*	Unmodelled genes^ in genome design	Unmodelled genes^ in gene deletions	Predicted *in-silico* genome size
Forster and Church 2006 (Forster & Church, 2006)	Nashville	Protocell	89	0	-	131
Tomita *et al.* 1999 (Tomita et al., 1999)	Fujisawa	Protocell	98	0	-	140
Huang *et al.* 2013 (Huang et al., 2013)	Guelph	Comparative Genomics	123	5	-	160
Gil 2014 (Rosario Gil, 2015)	Valencia	Comparative Genomics	180	6	-	216
Mushegian and Koonin 1996 (Mushegian & Koonin, 1996)	Bethesda	Comparative Genomics	253	12	-	283
Karr *et al.* 2012 (Karr et al., 2012)	Stanford	Single Gene Deletions	242	-	0	284
Glass *et al.* 2006 (J. I. Glass et al., 2006)	Rockville 2	Single Gene Deletions	258	-	44	344
Hutchison *et al.* 1999 (Hutchison et al., 1999)	Rockville	Single Gene Deletions	265	-	41	348
-	*M. genitalium* whole-cell model*	-	359	124	-	401
-	*M. genitalium in-vivo* *	-	483	-	-	525

Comparing minimal gene sets with M. genitalium in-vivo and the M. genitalium whole-cell

*model. * = protein-coding genes. M. genitalium has 42 RNA-coding genes that are not included*

in this column, to enable comparison across genome designs. ^ = due to unknown function.

Unmodelled genes in genome design reduces the size of the in-vivo genome design.

Unmodelled genes in gene deletions increases the size of the in-vivo genome design.

The protocell designs (Nashville, Fujisawa) predicted the smallest *in-silico* genome. Guelph

contained substantially fewer genes than Valencia and Bethesda, due to comparing 186

species (Huang et al., 2013). Stanford, Rockville, Rockville 2 had similar numbers of *in-vivo*

deletions, but Rockville and Rockville 2 had the highest numbers of unmodelled genes (as

genes can be disrupted *in-vivo* without knowing the gene's function). The *in-silico* genomes

(and associated gene deletions) of the minimal gene sets are listed in Supplementary Data 1

and 2.

Prior to simulations, we analysed the sets produced by comparative genomics and single gene

essentiality and found 96 genes they had in common (Supplementary Data 1 (col L) and 3). We

excluded the protocell designs from this comparison due to their much reduced size. 87 of the

shared genes were classified as essential (genes that when removed *in-silico* stopped the cell

from dividing, classified *in-silico* previously (Karr et al., 2012; Rees-Garbutt et al., 2020)), eight

were non-essential (removal did not prevent successful cell division), and one gene was

unmodelled.

The 87 essential genes affect a range of cellular functions including: DNA (repair, supercoiling,

chromosome replication, nucleotide synthesis/modification, sigma factors, ligation, transcription

termination, and DNA polymerase); RNA (ribosome proteins, translation initiation factors, tRNA

modification, ribonucleases, and RNA polymerase); and cellular processes (protein

folding/modification, protein shuttling, protein membrane transport, metabolic substrates

production/recycling, redox signalling, oxidation stress response, and the pyruvate

dehydrogenase complex). Of the eight non-essential genes, four (MG_048, MG_072, MG_170,

MG_297) are associated with the SecYEG complex (du Plessis et al., 2011) (protein transport

across or into the cell membrane), while MG_172 removes protein synthesis targets from

synthesised proteins, MG_305 and MG_392 assist in late protein folding, and MG_425

processes ribosomal RNA precursors. Although these eight genes are singly non-essential (by

single gene deletion *in-silico* (Karr et al., 2012) and *in-vivo* (J. I. Glass et al., 2006)) they all play

a part in essential functions, hence their inclusion.

We also identified 14 genes deleted by all eight minimal gene sets (Supplementary Data 2 (col

L) and 4). The functions of these genes include: fructose import, host immune response

activation, chromosomal partition, amino acid transport, antibody binding, phosphonate

transport, external DNA uptake, DNA repair, rRNA modification, membrane breakdown, toxin

transport, quorum sensing, and a restriction enzyme. These had been previously classified as

non-essential by single gene deletion *in-silico* (Karr et al., 2012) and *in-vivo* (J. I. Glass et al.,

2006). We placed these 14 common genes in an 'Agreed set' and a genome with these genes

removed was also simulated.

We simulated each minimal gene set in the *M. genitalium* whole-cell model and found that

every set, including the Agreed set, produced a non-dividing *in-silico* cell (30 *in-silico* replicates,

Supplementary Data 5).

Analysis found that every one of the sets deleted essential genes (classified *in-silico* previously

(Karr et al., 2012; Rees-Garbutt et al., 2020)): Nashville deleted 121, Fujisawa deleted 112,

Guelph deleted 107, Valencia deleted 69, Bethesda deleted 34, Rockville and Rockville 2 both

deleted 9, and Stanford deleted 3 (Supplementary Data 6-14). This is especially surprising for

the single gene essentiality minimal gene sets. For the Rockville sets, this is likely due to

transposon mutagenesis issues. Rockville labelled six genes as non-essential in 1999,

subsequently labelled essential in 2006. Additionally, Rockville grew cells in mixed pools with

DNA isolated from these mixtures rather than from isolated pure colonies of cells (J. I. Glass et

al., 2006). Whereas for Rockville 2, the engineers of *JCVI-Syn3.0* used the same transposon

mutagenesis protocol (Hutchison et al., 2016), but found that they had to improve upon it due

to incorrect identification of essentiality. For the Stanford set, the removal of MG_203, MG_250,

and MG_470 is likely due to averaging multiple simulation's data together before computational

assessment, genes later found to be essential (Supplementary Table 3(Rees-Garbutt et al.,

2020)).

In an attempt to restore *in-silico* division, we reintroduced essential genes to the minimal gene

sets (Figure 1). Based on previous research (Rees-Garbutt et al., 2020), we also reintroduced

low essential genes (i.e. genes dispensable in some contexts, such as redundant essential

genes and gene complexes (Rancati et al., 2018)). We did this by comparing the gene content

of the individual minimal gene sets with a complete list of the *M. genitalium in-silico* genes and

their essentiality classifications (Rees-Garbutt et al., 2020) (Supplementary Data 6-14). For

example, the original Agreed set removed low essential genes MG_291 (phosphonate

transport) and MG_412 (phosphate transport); by disrupting both these processes, the *in-silico*

cell has no functioning source of phosphate, which has been established previously

(Rees-Garbutt et al., 2020). By reintroducing set-specific genes (Table 2, Supplementary Data

15), each modified set, including the Agreed set, was able to produce a dividing cell *in-silico* (30

in-silico replicates, Supplementary Data 5).

Figure 1. *Testing and restoring the minimal gene sets.*

124 genes were unmodeled in the M. genitalium whole-cell model due to unknown function.

Essential genes are required by the cell to enable survival until successful division. Low

essential genes are required by the cell in certain genomic and environmental contexts to

enable survival to successful division. The restored minimal gene sets produced dividing

in-silico cells. Greater detail on the minimal gene sets is provided in Table 1 and 2.

Table 2. *Reintroducing genes to produce division*

Code name	*In-silico* gene deletions (cell did not divide)	*In-silico* gene deletions (cell divided)	Essential genes reintroduced	Low essential genes reintroduced	Size of *in-silico* genome *
Nashville	270	142	121	7	259
Fujisawa	261	142	112	7	259
Guelph	241	128	107	6	273
Valencia	185	110	69	6	291
Stanford	117	109	3	5	292
Bethesda	118	82	34	2	319
Rockville 2	57	45	9	3	356
Rockville	53	43	9	1	358
Agreed	14	13	0	1	388

*Minimal gene sets before and after the reintroduction of essential and low essential genes dividing in-silico cells. *Size of in-silico genome includes both the 359 protein-coding genes and the 42 RNA-coding genes in the M. genitalium whole-cell model, to provide a better comparison for in-vivo experiments and future work.*

In an attempt to gain further understanding, we investigated what processes the nine repaired minimal gene sets removed using gene ontology (GO) biological process terms, and we compared the repaired minimal gene sets to the *in-silico M. genitalium* minimal genomes we produced previously (Rees-Garbutt et al., 2020). The smallest repaired *in-silico* genomes (Nashville and Fujisawa, 259 genes) were larger than the prior novel *in-silico* minimal genomes (256 and 237 genes), and both removed the low essential genes related to phosphonate transport, relying on inorganic phosphate being transported into the cell (Supplementary Table 6(Rees-Garbutt et al., 2020)). The other sets had different designs that, due to not systematically targeting non-essential genes, resulted in non-essential genes remaining in the genome, making them subsets of the repaired protocell sets and the prior *in-silico* minimal

genomes (Supplementary Data 16). As such, the GO terms were also subsets and did not

deviate from what we would expect to produce a dividing *in-silico* cell (Supplementary Data

17-25).

Analysis of the repaired sets found that 31 genes were reintroduced into five or more of the

minimal gene sets (Supplementary Data 26). 26 were essential and 5 were low essential

(Supplementary Table 5(Rees-Garbutt et al., 2020)). The corresponding cellular functions

included: DNA (polymerase subunits, thymidine insertion, recycling of pyrimidine, chromosome

segregation); RNA (polymerase subunit, tRNA modification, the 50S and 30S ribosomal

subunits); transporters (cobalt, phosphonate, potassium); production (NAD, flavin, NADP, fatty

acid/phospholipids); and dehydrogenation (glycerol and alpha-keto acids). Of the 26

reintroduced essential genes, 19 were already present in the single gene deletion minimal gene

sets (Stanford, Rockville, Rockville 2). MG_137 (mycobacteria cell wall production) and MG_517

(plasma membrane stability) are genes specifically essential for *Mycoplasma* species, which

were only identified as essential by the single gene deletion minimal gene sets. A further five

genes were involved in cobalt transport, which increases the rates of DNA synthesis, fatty acid

metabolism, and amino acid metabolism, and were also not identified as essential by the other

design methodologies. Of the five reintroduced low essential genes, Bethesda did not delete

four, likely due to the direct comparison of two genomes resulting in low essentiality genes

being conserved to a greater degree.

We also looked at reintroductions to the protocell sets, as they could outline additional cellular

requirements for the successful future unification of independent protocell systems. The genes

reintroduced to the Nashville set repaired functions that had been reduced (translation,

glycolytic process, protein folding) and restored functions that had been removed including: cell

(division, cycle, transport, redox homeostasis), DNA (topological change, transcription), rRNA

processing (including pseudouridine synthesis), protein transport, and cellular processes

(carbohydrate metabolic, glycerol metabolic, fatty acid biosynthesis, UMP salvage)

(Supplementary Data 27). The glycolytic process had the most change, with 10 of 11 genes

being reintroduced, and DNA repair had the least, with only one gene being reintroduced

(MG_254), however, this did reintroduce single strand DNA break repair. The genes additionally

reintroduced to the Fujisawa set included tRNA processing and protein folding (Supplementary

Data 28), with 8 out of 10 DNA replication genes reintroduced.

Discussion

In conclusion, the repaired protocell minimal gene sets (Nashville and Fujisawa) produced the

smallest genomes *in-silico* (Table 2), differing by 6 genes (Supplementary Data 29), but required

the most gene reintroductions. The repaired comparative genomics minimal gene sets (Guelph,

Valencia, Bethesda) required fewer gene reintroductions, presumably because fewer genomes

were compared during their original design. Interestingly, Stanford (a single gene essentiality

set) produced a smaller *in-silico* genome than Bethesda (a comparative genomics set), as it did

not target unmodelled gene deletions and only required eight genes to be reintroduced.

This research has limitations associated with the use of the *M. genitalium* whole-cell model.

Through necessity the *M. genitalium* whole-cell model bases some of its parameters on data

from other bacteria [8] and is only capable of modelling a single generation, missing

subgenerational gene expression and subsequent essentiality which affects >50% of the genes

in some species (Macklin et al., 2020). Additional uncertainty exists around the unknown impact

of the unmodelled genes on *in-vivo* experiments, as stated previously (Rees-Garbutt et al.,

2020).

The computational predictions we have produced need to be tested in living cells, but with the

advancement of gene synthesis and genome transplantation in other *Mycoplasma* species

(Benders et al., 2010; Gibson, Benders, Andrews-Pfannkoch, et al., 2008; Gibson, Benders,

Axelrod, et al., 2008; Gibson et al., 2010; Karas et al., 2013; Tsarmpopoulos et al., 2016) this is

becoming a more realistic proposition for *Mycoplasma* researchers.

Without the ability to identify species-specific low essential genes, any minimal gene sets

designed with the current incomplete and fragmented gene essentiality information will require

gene reintroductions to produce dividing cells. New research is required to generate the

knowledge required to fill these gaps. Currently, genome-scale engineering has not combined

in silico and *in vivo* research (Jiang et al., 2013; Karr et al., 2012; Rees-Garbutt et al., 2020).

With the recent publication of an *E.coli* whole-cell model (Macklin et al., 2020) and the

introduction of a synthetic, recoded *E.coli* genome into a cell (Fredens et al., 2019), this will

quickly change. The current nature of species-specific gene essentiality information is,

potentially, the only remaining barrier to producing custom genomes.

Methods

These methods have been published previously [27].

Experimental Design and Statistics

Each simulation was replicated 30 times. Prior publications have completed five repetitions [8], ten or one hundred repetitions [27], depending on data storage capacity and supercomputer usage. These were the main factors in our choice for the number of replications.

The initial conditions for each simulation are randomised within a natural range for *M. genitalium* cells, introducing variation in the behaviour of simulated cells [8], as previously shown [27]. Statistical tests were not used in this research.

Code Availability

All code created as part of this paper will be made available on Github under a GNU General Public License v3.0 (gpl-3.0). For more information see choosealicense.com/licenses/lgpl-3.0/.

Data Availability

The databases used to design the *in-silico* experiments, and compare the results to, includes Karr *et al.* (Karr et al., 2012) and Glass *et al.* (J. I. Glass et al., 2006) Supplementary Tables, and Fraser *et al. M. genitalium* G37 genome (Fraser et al., 1995) interpreted by KEGG (Kanehisa & Goto, 2000) and UniProt (Apweiler et al., 2004) as strain ATCC 33530/NCTC 10195. The output .fig files for all simulations referenced will be made available at the group's Research Data Repository (data-bris) at the University of Bristol.

Model Availability

The *M. genitalium* whole-cell model is freely available: github.com/CovertLab/WholeCell. The

model requires a single CPU and can be run with 8GB of RAM. We run the *M. genitalium*

whole-cell model on Bristol's supercomputers using MATLAB R2013b, with the model's

standard settings. However, we use our own version of the SimulationRunner.m.

MGGRunner.m

(github.com/GriersonMarucciLab/Analysis_Code_for_Mycoplasma_genitalium_whole-cell_mod

el) is designed for use with supercomputers that start hundreds of simulations simultaneously. It

artificially increments the starting time-date value for each simulation, as this value is

subsequently used to create the initial conditions of the simulation. Our research copy of the

whole-cell model was downloaded 10th January 2017.

M. genitalium in-silico Environmental Conditions

M. genitalium is grown *in-vivo* on SP4 media. The *in-silico* media composition is based on the

experimentally characterized composition, with additional essential molecules added

(nucleobases, gases, polyamines, vitamins, and ions) in reported amounts to support *in-silico*

cellular growth. Additionally, the *M. genitalium* whole-cell model represents 10 external stimuli

including temperature, several types of radiation, and three stress conditions. For more

information see Karr *et al.* Supplementary Tables S3F, S3H, S3R (Karr et al., 2012).

Equipment

For the *M. genitalium* whole-cell model we used the University of Bristol Advanced Computing

Research Centres's BlueGem, a 900-core supercomputer, which uses the Slurm queuing

system, to run whole-cell model simulations.

We used a standard office desktop computer, with 8GB of ram, to write new code, and interact

with the supercomputer. We used the following GUI software on Windows 7: Notepad++ for

code editing, Putty (ssh software) for terminal access to the supercomputer, FileZilla (ftp

software) to move files in bulk to and from the supercomputer, and PyCharm (IDE software) as

an inbuilt desktop terminal and for python debugging. The command line software used

included: VIM for code editing, and SSH, Rsync, and Bash for communication and file transfer

with the supercomputers.

Data Format

For the *M. genitalium* whole-cell model the majority of output files are state-NNN.mat files,

which are logs of the simulation split into 100-second segments. The data within a

state-NNN.mat file is organised into the 16 cellular variables. These are typically arranged as

3-dimensional matrices or time series, which are flattened to conduct analysis. The other file

types contain summaries of data spanning the simulation. Each gene manipulated simulation

can consist of up to 500 files requiring between 0.4GB and 0.9GB. Each simulation takes 5 to

12 hours to complete in real time, 7 - 13.89 hours in simulated time.

Data Analysis Process

For the *M. genitalium* whole-cell model, the raw data is automatically processed as the

simulation ends. runGraphs.m carries out the initial analysis, while compareGraphs.m overlays

the output on collated graphs of 200 unmodified *M. genitalium* simulations. Both outputs are

saved as MATLAB .fig and .pdfs, though the .pdf files were the sole files analysed. The raw

.mat files were stored in case of further investigation.

The GO biological process terms used for further analysis were downloaded from Uniprot

(Apweiler et al., 2004) (strain ATCC 33530/NCTC 10195), processed by a created script

(github.com/squishybinary/Gene_Ontology_Comparison_for_Mycoplasma_genitalium_whole-ce

ll_model) in combination with lists of genes, organised manually into tables of GO terms that

were unaffected, reduced, or removed entirely by gene deletions, and then analysed.

Modelling Scripts

There are six scripts used to run the *M. genitalium* whole-cell model. Three are the experimental

files created with each new experiment (the bash script, gene list, experiment list), and three are

stored within the whole-cell model and are updated only upon improvement (MGGrunner.m,

runGraphs.m, and compareGraphs.m). The bash script is a list of commands for the

supercomputer(s) to carry out. Each bash script determines how many simulations to run,

where to store the output, and where to store the results of the analysis. The gene list is a text

file containing rows of gene codes (in the format 'MG_XXX',). Each row corresponds to a single

simulation and determines which genes that simulation should knockout. The experiment list is

a text file containing rows of simulation names. Each row corresponds to a single simulation

and determines the final location of the simulation output and analysis results.

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Competing Interests

The authors declare no competing interests.

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Author Contributions

C.G., L.M., O.C., J.R-G were involved in ideation. O.C. conducted simulations comparing the Stanford and Rockville 2 sets, a prototype version of the research presented here. J.R-G. collated, simulated, analysed and repaired the nine sets, and wrote the paper and supplementary data 1 - 16, 26 - 29. J.R. analysed the nine sets and wrote supplementary data 17 - 25. C.G. and L.M supervised the project. C.G., L.M., J.R. edited the paper.

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