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Evolutionary forecasting of phenotypic and genetic outcomes of experimental evolution in *Pseudomonas*

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prediction of short-term evolution in *P. protegens* Pf-5 and lays the foundation for
evolutionary forecasting in other *Pseudomonas*.

Abstract

Experimental evolution is often highly repeatable, but the underlying causes are

generally unknown, which prevents extension of evolutionary forecasts to related

species. Data on adaptive phenotypes, mutation rates and targets from the

Pseudomonas fluorescens SBW25 Wrinkly Spreader system combined with

mathematical models of the genotype-to-phenotype map allowed evolutionary

forecasts to be made for several related *Pseudomonas* species. Predicted outcomes of

experimental evolution in terms of phenotype, types of mutations, relative rates of

pathways and mutational targets were then tested in *Pseudomonas protegens* Pf-5. As

predicted, most mutations were found in three specific regulatory pathways resulting

in increased production of Pel exopolysaccharide. Mutations were, as predicted,

mainly found to disrupt negative regulation with a smaller number in upstream

promoter regions. Mutated regions in proteins could also be predicted, but most

mutations were not identical to those previously found. This study demonstrates the

potential of short-term evolutionary forecasting in experimental populations.

Introduction

An increasing number of experimental evolution studies, primarily using microbes, have provided insights into many fundamental questions in evolutionary biology including the repeatability of evolutionary processes (Barrick and Lenski 2013; Jerison and Desai 2015; Long, et al. 2015; Orgogozo 2015). Given the ability to control environmental conditions as well as population size and the use of a single asexual organism, such studies could provide an ideal test of our ability to predict evolutionary outcomes in simplified model systems. High repeatability on both phenotypic and genetic level have been observed in a large number of experimental evolution studies (Wichman, et al. 1999; Conrad, et al. 2009; Lee and Marx 2012; Tenaillon, et al. 2012; Barrick and Lenski 2013; Ferguson, et al. 2013; Herron and Doebeli 2013; Blank, et al. 2014; McElroy, et al. 2014; Fraebel, et al. 2017; Kram, et al. 2017; Knoppel, et al. 2018), but it has become clear that high repeatability alone is not sufficient for testing evolutionary predictability beyond the prediction that under identical conditions the same evolutionary outcome is likely. The difficulties of moving from repeatability to predictability are largely a result of the lack of knowledge of the genotype-phenotype-fitness map (Figure 1).

Figure 1. Prediction of the adaptive outcomes of experimental evolution requires understanding of how genotypes map to phenotypes and fitness. The differing capacity of genes to translate genotypic variation into phenotypic variation and

differences in mutation rates can introduce biases in the production of phenotypic

variation. Natural selection, genetic architecture and mutational biases can both

increase and decrease the predictability of evolution depending on if they can be

recognized beforehand and included into evolutionary forecasting models. (**A**)

Fitness space. Mutants that increase to high frequencies in the population are all

expected to have increased fitness as drift is negligible at population sizes typically

used for adaptation experiments with microbes. Much effort has been put into

characterizing the distribution of fitness effects of beneficial mutations, but the shapes

of the distribution and magnitudes of the fittest mutations appear to be highly context

dependent. Many different phenotypes are typically adaptive during experimental

evolution, but in most cases they are not known beforehand and their relative fitness

cannot be predicted. Relative fitness is, in most cases, also expected to be highly

dependent on external environment including the frequency of other adaptive mutants,

which means that even small changes to experimental protocols can lead to difference

in outcomes. (**B) Phenotype space**. Each of the adaptive phenotypes can usually be

realized by mutations in different positions and in different genes, but distinct

phenotypes are expected to have similar fitness regardless of genetic foundations,

which can simplify predictions. Depending on the genetic architecture underlying

each trait, which is often unknown, adaptive phenotypes are produced at different

rates. (**C) Parameter space**. The adaptive phenotypes are caused by changes in the

molecular networks of cells, which are also influenced by the external environment. If

the wiring of a molecular network underpinning an adaptive phenotype is well

understood, parameterization of the system is possible and predictive models can be

formulated. Mutations can cause functional effects on gene products, but mutations in

some genes are more likely to lead to phenotypic variation. This can for example be

due to differences in mutational robustness of the gene products themselves or their

functions in regulatory networks. (**D) Genotype space**. Mutation rates are not

uniform across the genome and mutational hot spots can lead to bias in the number of

mutants producing relevant changes in parameter space and causing new phenotypes

that are presented for natural selection to act upon. The current understanding of the

distribution of mutation rates is limited and computational predictions have not yet

been described. Adding information about well-characterized mutational hot spots,

including indels in homonucleotide tracts or deletion and duplication between

sequence homologies, could possibly improve prediction compared to a null model

with uniform rates. Alternatively experimental data might be incorporated into models

of mutation rates to improve predictions.

There are several problems that need to be solved to develop a model system for true

testing of predictive ability. In some cases adaptive mutations are highly strain

specific, so that for example adaptation of different strains to a specific environment

will produce different results. In some cases this is due to the long history of sub-

culturing under laboratory conditions combined with rounds of mutagenesis that has

caused, for example, many *E. coli* and *Salmonella* strains to accumulate diverse

mutations, some of which are rapidly compensated by secondary mutations restoring

fitness (Barrick, et al. 2009; Tenaillon, et al. 2012; Knoppel, et al. 2018). Thus, in

many cases conclusions from one strain cannot be extended to another because of

differences in their genotype-to-phenotype maps (Figure 1C).

Another problem for testing predictability is that in many cases it is not possible to

design an experiment where one specific selective pressure is dominant. For example

experiments with intended adaptation to high temperature (Tenaillon, et al. 2012) or

freeze-thaw-growth cycles (Sleight, et al. 2008) result in similar mutations in *uspA*,

which may indicate adaptation to the medium used (Knoppel, et al. 2018) or generally

stressful conditions. Relatively minor changes in environmental conditions can also

results in divergent mutational patterns (Deatherage, et al. 2017). This means that the

range of possible adaptive phenotypes cannot be defined beforehand (Figure 1B) and

that in many cases the phenotypes that solve the intended selective problem are

outcompeted by other phenotypes with increased fitness (Figure 1A).

A highly specific selective pressure can be applied by selection for antitbiotic

resistance and mutation targets are often highly conserved between different strains

and species and between laboratory and natural populations (O'Neill, et al. 2006;

Schenk, et al. 2012; Brandis, et al. 2015; Jahn, et al. 2017; Lukacisinova and

Bollenbach 2017; Sommer, et al. 2017). However resistance phenotypes are often

explained solely by the molecular phenotype of a single protein and no alternative

pathways to resistance are known resulting in a relatively simple parameter and

genotype space (Figure 1C, Figure 1D). Thus the prediction will be identical for all

species and it cannot provide a test of prediction from general principles. In many

cases mutants isolated after selection for high-level antibiotic resistance also lacks the

complexity that is inherent to many phenotypic traits where the genotype-to-

phenotype map involves a large number of functional interactions and complex

regulation (Figure 1C). This complexity comes to light in that adaptive mutations to

new environments are commonly found in global regulators of gene expression such

as genes involved in the stringent response, DNA binding proteins, supercoiling and

core genes for RNA and protein synthesis (Barrick, et al. 2009; Conrad, et al. 2009;

Kishimoto, et al. 2010; Tenaillon, et al. 2012; Herron and Doebeli 2013; Sandberg, et

al. 2014; LaCroix, et al. 2015; Deatherage, et al. 2017). The physiological effects of

these mutations are diverse, sometimes affecting the expression of hundreds of genes

making the elucidation of the molecular underpinnings of the adaptive phenotype

(Figure 1C, 1D) extremely complex and thus difficult to use for predictive modeling.

The wrinkly spreader model in *P. fluorescens* SBW25 (hereafter SBW25) is one of

the best-characterized experimental evolution systems and has several properties that

could make it possible to extend knowledge and principles from this species to related

species (Rainey and Travisano 1998; Spiers, et al. 2002; Spiers, et al. 2003; Spiers

and Rainey 2005; Goymer, et al. 2006; Bantinaki, et al. 2007; McDonald, et al. 2009;

Silby, et al. 2009; Ferguson, et al. 2013; Lind, et al. 2015, 2017b; Lind, et al. 2018).

When the wild type SBW25 is placed into a static growth tube the oxygen in the

medium is rapidly consumed by growing bacteria (Figure 2A). However oxygen

levels at the surface are high and mutants that are able to colonize the air-liquid

interface have a major growth advantage and rapidly increase in frequency (Figure

2A). Several phenotypic solutions to air-liquid interface colonization, all involving

increased cell-cell adhesion, have been described and are distinguishable by their

colony morphology on agar plates (Figure 2A, 2B) (Rainey and Travisano 1998;

Ferguson, et al. 2013; Lind, et al. 2017b). The most successful of these is the Wrinkly

Spreader (WS) (Ferguson, et al. 2013; Lind, et al. 2017b) that overproduces a

cellulosic polymer that is the main structural component of the mat at the air-liquid

interface (Spiers, et al. 2002; Spiers, et al. 2003). The WS phenotype is caused by

mutational activation of c-di-GMP production by a diguanylate cyclase (DGC)

(Figure 2C) (Goymer, et al. 2006). While many different DGCs can be activated to

reach the WS phenotype, some are greatly overrepresented due to larger mutational

target sizes leading to a hierarchy of genetic routes to WS (Figure 2D) (McDonald, et

al. 2009; Lind, et al. 2015). The genotype-to-phenotype map to WS has been

characterized in detail (Goymer, et al. 2006; McDonald, et al. 2009; Lind, et al. 2018)

allowing the development of mathematical models of the three main pathways to WS

(Wsp, Aws and Mws) and the prediction of evolutionary outcomes (Figure 2E) (Lind,

et al. 2018).

Figure 2. The *Pseudomonas fluorescens* SBW25 "wrinkly spreader" model system

has several properties that could allow its extension to other species (**A**) The ancestral

strain that has smooth colony morphology on agar plate is inoculated into static

growth tubes and incubated for several days. Depletion of oxygen in the medium

leads to competition for access to the oxygen-replete surface which is colonized by

mutants with enhanced ability for cell-cell adherence and adherence to the wall of the

tube. The most successful of these mutant types is the wrinkly spreader that has a

distinctive colony morphology due to overproduction of exopolymeric substances

(EPSs) of which a cellulosic polymer is the main structural component (Spiers, et al.

2002; Spiers, et al. 2003; Ferguson, et al. 2013; Lind, et al. 2017b) (**B**) There are

many potential solutions to colonization of the the air-liquid interface and in

Pseudomonas fluorescens at least four alternative distinct phenotypes are selected for

at the surface, but they all have lower fitness than the WS type (Beaumont, et al. 2009; Ferguson, et al. 2013; Gallie, et al. 2015; Lind, et al. 2017b). The SBW25 system is of intermediate complexity in that it allows isolation of a phenotypic subset of adaptive mutants, those that are selected at the air-liquid interface and produce a difference in colony morphology, rather than all mutants that increase fitness. However there is also a large diversity of different genetic and phenotypic solutions to the dominant adaptive challenge, which makes it of greater complexity than systems based on single genes, which is often the case for experimental systems using strong selection for antibiotic resistance. (**C**) The mutational causes of WS in wild type populations are found in either of three loci, *wspABCDEFR*, *awsXRO* and *mwsR,* which all encode diguanylate cyclases that produce the second messenger c-di-GMP (McDonald, et al. 2009) that is a conserved signal for EPS production and biofilm formation in many bacterial species. In SBW25 the primary EPS used is a cellulosic polymer. (**D**) While these three pathways account for >98% of WS mutants in the wild type, 13 additional rare pathways were found when the common ones are genetically deleted (Lind, et al. 2015). The large differences in mutation rates to WS for the different pathways are explained mainly by their different capacities to translate genotypic variation into phenotypic variation, i.e. mutational target size, with the three common pathways being subject to negative regulation, which when disrupted results in overproduction of c-di-GMP and the WS phenotype (Lind, et al. 2015). The alternative phenotypic solutions are also caused by inactivating mutations occurring at high rates (Ferguson, et al. 2013; Lind, et al. 2017b). The pathways to WS of intermediate frequency are activated by mutations to promoter regions, including promoter captures and the most rare are those that require specific activating mutations in the DGC or double mutations in two negative regulators. The genetic underpinnings of several adaptive phenotypes have been elucidated providing a mechanistic understanding of the effects of mutations and why they are adaptive. (**E**) The three main pathways (Wsp, Aws, Mws) to WS are particularly well-understood allowing formulation of mathematical models of the molecular networks and prediction of the relative rates of use of the different pathways and genes (Lind, et al. 2018).

This study makes initial forecasts of phenotypic and genetic evolutionary outcomes after static experimental evolution for six *Pseudomonas* species based mainly on data

from SBW25 (McDonald, et al. 2009; McDonald, et al. 2011; Ferguson, et al. 2013;

Lind, et al. 2015, 2017b). Predictions of evolutionary outcomes are then

experimentally tested for the closely related species, *P. protegens* Pf-5 (hereafter Pf-

5) with a highly conserved genetic repertoire of DGCs but that lack the structural

component used for biofilm formation in SBW25. Results show that phenotypes,

order of pathways used and types of mutations can be predicted and that forecasts are

robust to changes in environmental conditions.

Results

Six *Pseudomonas* species (Figure 3 legend) were chosen based on phylogenetic

diversity and their complement of DGCs and EPSs for a first round of predictions.

These species encode from none to all three of the main DGCs used in SBW25 and

only three species contain genes related to cellulose biosynthesis, the main EPS used

in SBW25 (Figure 3).

Fig 3. Diversity of DGCs and biofilm-related genes for seven *Pseudomonas*

species (*P. fluorescens* SBW25, P. *protegens* Pf-5, *P. putida* KT2440, *P. syringae* pv.

tomato DC3000, *P. savastanoi* pv. phaseolicola 1448A, *P. aeruginosa* PAO1, *P.*

stutzeri ATCC 17588) **(A)** The seven species encode 251 putative DGCs, divided into

87 different homolog groups of which 8 are present in all genomes. WS mutations in

SBW25 have been found affecting 13 of these DGCs (marked with *) with an

additional nine that have been detected only in combinations with other mutations.

SBW25 and Pf-5 share 33 DGCs with 6 unique for each species. It should be noted

that not all DGCs are likely to be catalytically active. **(B)** Diversity of biofilm-related

genes including putative EPSs, LPS modification, cell chaining, adhesins and known

regulators. In SBW25 cellulose-based mats are most successful (encoded by the *wss*

operon). A secondary exopolysaccharide (PGA), encoded by *pgaABCD*, can also be

used to form a stable mat (Lind, et al. 2017b). Fuzzy spreaders (FS) forms rafts that

collapse after becoming too large and the mutational cause is inactivation of *fuzY*,

which results in a defect in lipopolysaccharide (LPS) modification (Ferguson, et al.

2013). Cell-chaining (CC) types have loss-of-function mutations in *nlpD* causing a

defect in cell division, which leads to segmented chains of cells that can form a weak

mat at the surface (Lind, et al. 2017b). Full details are available in Figure 3 - source

data. The genomes of the *P. aeruginosa* strains PA7, UCBPP-PA14 and LESB58

were also included in the analysis and results were in most cases identical to PAO1

(not shown in Figure 3) with the exception of an absence of homologues for EPS

genes *pelA-D* and the DGCs PA2771 in PA14 and PA3343 in PA7.

Ecotype predictions

Given the range of ways that cells can achieve increased adherence and surface

colonization by use of different EPSs, LPS modification and cell chaining as

demonstrated by the studies with SBW25 (Spiers, et al. 2002; Ferguson, et al. 2013;

Lind, et al. 2017b) all species are be expected to colonize the air-liquid interface if

access to oxygen is limiting for growth. This could be achieved simply by changes in

gene expression in the wild type, but for an experimental evolution study a mutational

solution is sought and environmental conditions are chosen so that the wild type strain

does not colonize the air-liquid interface. However changing environment presents a

further challenge because, as discussed above, it often leads to a different spectrum of

adaptive mutations. Thus a foundational requirement for an extended experimental

evolution system to be successful for different species is that the evolutionary

solutions are robust to differences in environmental conditions.

Phenotype predictions

Several different phenotypic solutions can be used to colonize the air-liquid interface

in SBW25 including at least two different EPSs (cellulose (WS) and PGA (PWS),

LPS modification (fuzzy spreaders FS) and cell chaining (CC) (Spiers, et al. 2002;

Ferguson, et al. 2013; Lind, et al. 2017b). However wrinkly spreaders that form

cellulose-based mats are superior and rapidly outcompete all other types (Ferguson, et

al. 2013; Lind, et al. 2017b). Based on the limited data available it is predicted that

cellulose-based biofilms are superior in other species as well and that they will be the

primary structural solution when available as for *P. syringae*, *P. putida* and *P.*

stutzeri. For the three species lacking genes for cellulose biosynthesis, other EPSs are

predicted to be used. Based on studies of *P. aeruginosa* the primary EPS required for

pellicle formation at the air-liquid interface in this species is Pel, encoded by the

pelABCDEFG operon, which is also present in the Pf-5 genome and is predicted to be

the primary phenotypic solution for these species. The genome of *P. savastanoi* lacks

genes for biosynthesis of cellulose and Pel as well as other EPSs that are known to be

able to support mat-formation, such as PGA and there is not sufficient data at this

point to make a prediction of which one is likely to be the primary phenotypic

solution.

Overexpression of EPSs used for mat-formation at the air-liquid interface is in

SBW25 and *P. aeruginosa* linked to mutations increasing c-di-GMP production rather

than mutations in the promoters of or genes in the EPS operons themselves. This can

be explained by the role of post-translation regulation by c-di-GMP in the production

of cellulose, Pel, PGA and alginate (Lee, et al. 2007; Romling, et al. 2013; Steiner, et

al. 2013; Morgan, et al. 2014; Liang 2015; Whitney, et al. 2015). In these cases

transcriptional upregulation alone is not likely to cause overproduction because of

lack of a c-di-GMP signal. Possibly there is also an additional benefit to using

activation of the c-di-GMP network in that it reduces motility, which is not needed

when established at the air-liquid interface and which consumes large amount of

energy to sustain and thus is likely to be selected against (Koskiniemi, et al. 2012; Lee

and Marx 2012).

Prediction of types of mutations

Disabling mutations are expected to be more common than enabling mutations and

therefore the prediction is that most mutations will be in genes where loss-of-function

mutations produce an adaptive phenotype (Fig 2D) (Lind, et al. 2015). This is the case

for the large majority of mutations in SBW25 including those activating main DGCs

WspR, AwsR and MwsR, which are all under negative regulation, as well as

disruption of the genes underpinning the FS phenotype (*fuzY*, PFLU0478) and CC

phenotype (*nlpD*, PFLU1301) (McDonald, et al. 2009; Ferguson, et al. 2013; Lind, et

al. 2017b). Next in the hierarchy of mutations are promoter mutations, increasing

transcription, and promoter capture events (Lind, et al. 2015). Less common are

intragenic activating mutations that enable a particular function by for example

increase in catalytic activity or strengthening of interactions to another molecule or

another domain of the same protein (Lind, et al. 2015). Gene duplications occur at a

high rate and clearly have the ability to increase gene expression of DGCs, but they

have not yet been found to cause WS in SBW25, possibly because a two-fold increase

in gene expression is insufficient.

Prediction of pathways used

There are at least 16 different pathways to the WS phenotype in SBW25 with similar

fitness, but they are used at frequencies that vary over several orders of magnitude

based on the differing capacity to translate phenotypic variation into phenotypic

variation (Figure 2D) (Lind, et al. 2015). Mutations in three pathways, Wsp, Aws, and

Mws account for >98% of WS mutations and based on a detailed understanding of the

molecular functions of the genes involved of each pathways mathematical models

predicting at which relative rates the pathways should be used were constructed

(Figure 2E) (Lind, et al. 2018). The prediction results varies depending on the rates of

disabling and enabling mutations, but if it is assumed that disabling mutations are at

least an order of magnitude more common than enabling mutations the models predict

that Wsp will account for about 53%, Aws 28% and Mws 19% of the WS mutations

(Figure 2E) (Lind, et al. 2018).

Less common promoter mutations will also appear at rates at least a magnitude lower,

but which DGCs that will be transcriptionally activated cannot be easily predicted

except for assuming it will be homologs of the ones used in SBW25. These DGCs

must be catalytically active and also be localized to the membrane (Farr, et al. 2017).

Possibly the subset of DGCs that are primarily activated by mutations to their

promoters is mainly determined by mutation rate and a higher mutation rate might be

caused by higher transcription and also influenced by gene direction (Sankar, et al.

2016). Most promoter capture deletion events are less than five kilobases (Lind, et al.

2015) in size and the lack of an alternative promoter relatively close upstream to the

DGC is likely to rule out these DGCs. The DGCs that can be activated by intragenic

activating mutations cannot now be predicted beyond the simple prediction that these

are the same genes as in SBW25 (Lind, et al. 2015).

Prediction of mutated genes

In addition to predicting the relative rates of the three main pathways, the previously

described mathematical model can also predict which proteins are likely to be

mutated (Lind, et al. 2018). High rates of WS mutations are predicted for WspF,

WspA, WspE, AwsX and AwsR and MwsR (Figure 2E). A significantly lower rate of

enabling mutations is also predicted to occur in WspC, WspR and AwsO (Figure 2E).

Despite the simplicity of null model it closely predicted the mutational targets in

SBW25 with equal rates for WspF, WspA and WspE and rare mutations in WspC and

WspR, suggesting that it is a useful null model also for other species (Lind, et al.

2018).

Prediction of specific mutational targets and effects of mutations

The level of parallelism at the nucleotide level between species is expected to be

dependent both on the number of possible mutations to WS and the degree of

functional conservation of the proteins involved that define the genotype-to-

phenotype map. Mutational hot spots are also expected to contribute to parallelism

when they are conserved, but reduce parallelism when they are not. Based on previous

analysis of patterns of mutations in SBW25 (McDonald, et al. 2009; McDonald, et al.

2011; Lind, et al. 2018) and homology modeling of protein structure using Phyre2

(Kelley, et al. 2015) regions expected to be mutated were predicted and the likely

molecular consequences of different mutations suggested (Figure 4).

Figure 4. Predicted mutational targets and proposed molecular effects. Black

represents any inactivating mutation including frame shifts, blue represents in frame

inactivating mutations, green represents amino acid substitutions. Numbers in

brackets refers to amino acid residue numbers in SBW25 (**A**) WspA – amino acid

substitutions are expected at the tip of the stalk and in-frame deletion of methylation

sites (**B**) WspF – any inactivating mutation is predicted, amino acid substitutions are

predicted only in areas where they disrupt intermolecular interactions (**C**) WspE –

amino acid substitutions are predicted near the phosphorylation site (**D**) WspR – small

in frame deletion and amino acid substitutions in the linker is predicted to cause

constitutive activation (**E**) AwsO – amino acid substitutions disrupting AwsO

dimerization is predicted to lead to increased binding to AwsX without the presence

of an activating signal (**F**) AwsR - amino acid substitutions in the periplasmic region

or transmembrane helix that disrupt the interaction with AwsX or to the HAMP linker

is predicted (**G**) AwsX – any inactivating mutation that keep the reading frame intact

and do not interfere with expression of downstream AwsR is predicted (**H**) MwsR –

mutations are predicted in the interface between the DGC and phosphodiesterase

domains and in the most C-terminal of the PAS domains resulting in constitutive

activation.

Prediction of fitness effects of WS mutations

While conservation of relative fitness of different phenotypic variants might be

expected there is no clear reason to expect that the relative fitness of different DCG

pathways and mutations will be conserved between species. Despite this difficulty

there might be a way forward to predict the relative fitness of a large range of

mutations with limited experimental data. The distribution of fitness effects of new

mutations have been found to be bimodal for a large number of genes with different

functions with one mode close to neutrality and one corresponding to a complete loss

of a particular molecular function (Jacquier, et al. 2013; Jimenez, et al. 2013;

Firnberg, et al. 2014; Lind, et al. 2017a; Lundin, et al. 2017). Given that mutations

that allow colonization of the air-liquid interface have large phenotypic effects and

are believed to also have large effects on molecular function, often a complete

disruption of an interaction, adaptive mutations in the same region of a protein are

likely to have similar fitness effects. Thus, an approximation of the distribution of

fitness effects could be possible with relative few mutations for each gene. This is

supported by the relatively small number of WS mutants in SBW25 that have been

characterized with sensitive fitness assays and where mutations in the same gene

typically have similar fitness effects (Lind, et al. 2015; Lind, et al. 2018). If this

assumption is true the distribution of beneficial fitness effects is not continuous and

the most advantageous mutations are not predicted to be equally spread between

pathways or genes. Thus the prediction would be that mutants isolated after

experimental evolution were concentrated to certain genes even if the mutational rate

is similar so that although the prediction from the null model is equal number of

mutations for WspA, WspE and WspF such distribution is unlikely to be found. While

the mutation rates to WS for the three genes are similar in SBW25, WspA mutants are

rarely found after experimental evolution due to their lower fitness (McDonald, et al.

2009; Lind, et al. 2018). There is however no reason to expect that the relative fitness

of mutations in different genes or pathways will be conserved between species.

Inactivating mutations in *fuzY* and *nlpD* producing the alternative adaptive

phenotypes based on LPS modification or cell chaining were also found to have

similar fitness (Ferguson, et al. 2013; Farr 2015). Possibly there are other genes that

can be mutated with similar phenotypes, but that those mutants have lower fitness and

are outcompeted in SBW25. If relative fitness is not conserved between species this

could lead to high convergence on the phenotypic level but with completely different

genetic bases.

Experimental test of forecasts in *P. protegens* Pf-5

In order to test the predictions outlined above, *P. protegens* Pf-5 was used for a

parallel experimental evolution study under static conditions. Its genome encodes

homologs of all except one of the DGCs used in SBW25 including all the three

common pathways to WS (Wsp, Aws, Mws). Thus the genetic predictions in terms of

types of mutations and mutated genes are nearly identical to SBW25 and the

mathematical null models can be directly applied. However Pf-5 lacks genes for

biosynthesis of cellulose meaning that if c-di-GMP overproduction is the main

pathway used, as predicted, an alternative EPS component must be utilized. The

experimental conditions were modified to test the robustness of predictions to changes

in media composition, temperature and cell wall material compared to those used in

the original SBW25 system (Materials and Methods). After experimental evolution

for five days, dilutions were spread on agar plates and then screened for colonies with

divergent colony morphology, characteristic of many phenotypes that colonize the air-

liquid interface.

In total 43 independent mutants were isolated and the causal mutations were

identified (Figure 5, Figure 5 – supporting data). As predicted by the null model the

majority (40/43) of mutations were associated with the Wsp, Aws, and Mws pathways

that are subject to negative regulation (Figure 1D). In addition the prediction that

promoter mutations would be the second most common type of mutation was

successful with two mutations found upstream of the *aws* operon, which were

predicted to disrupt the terminator of a high expression ribosomal RNA operon

representing a promoter capture event. Promoter mutations were also found upstream

of PFL_3078, which is the first gene of a putative EPS locus (PFL_3078-3093) that

has not previously been described and that is only present in closely related strains.

The operon encodes genes typical of exopolysaccharide biosynthetic operons making

it highly likely it encodes the main structural component used by these mutants.

Figure 5. 43 independent mutants of *Pseudomonas protegens* Pf-5 were isolated after
experimental evolution based on their divergent colony morphology and mutations
were identified in four operons. Numbers in brackets are the number of independent
mutants found. Details are available in Figure 5 – source data.

The mathematical null model (Figure 2E) successfully predicted that of the three
common pathways to WS, Wsp would be the most common one (16 mutants)
followed by Aws (14) and then Mws (10). Mutations were predominately found in the
negative regulators WspF (15 mutants) or AwsX (9), but also in interacting proteins
WspE (1) and AwsR (3). Given that the mutational target size is estimated to be
smaller for the interacting proteins (Figure 4) this is not surprising. No mutations
were found in WspA despite a predicted high rate.

Mutations were predominantly found in predicted regions (Figure 4) for WspF,
WspE, AwsX, AwsR and MwsR, but in most cases they were not identical to those in
SBW25 (Figure 5 – supporting data). A mutational hot spot was apparent in WspF
with 12 out of 15 mutations being identical V271G missense mutations. The
previously described mutational hot spots in SBW25 in the *awsX* and *mwsR* genes
(Lind, et al. 2018) appeared absent, demonstrating how mutation rate differences can
skew evolutionary outcomes even for closely related species with similar genetic
architecture.

In total 60 wells were inoculated and subjected to experimental evolution for five

days after which air-liquid interface colonization was observed for the majority of the

wells. Mutants with clearly visible changes in colony morphology were isolated from

43 wells. Typically a single type of divergent colonies was observed and one colony

for each well was selected for further characterization at random based on a pre-

determined position on the agar plate. Representative mutations were reconstructed

using an allelic exchange protocol to determine that the mutations are the sole cause

of the air-liquid interface colonization and colony phenotypes and to exclude the

influence of secondary mutations (Figure 6A) before further characterization.

Figure 6. Phenotypic characterization of reconstructed mutants in WspF (V271G,

Q297K), WspE (S661L), AwsX (deletion V133-P143), AwsR (A194V), MwsR

(M925I, E1081K) and upstream *awsXRO* (*recC* -129 g->a) and PFL_3078 (-98 c->t)

(**A**) Motility, colony morphology and air-liquid interface colonization (**B**) Motility

assay. As expected if the c-di-GMP network is activated motility was reduced for

most mutants with the exception of PFL_3078 that is not expected to have increased

c-di-GMP levels.

The lack of cellulose biosynthetic genes also shows that these ecotypes can evolve by

different phenotypes than in SBW25. Increased Congo Red binding ability suggests

than alternative EPS components are used. Two clearly different phenotypes were

observed with one very similar to the original WS types in SBW25 with a clear

motility defect and mutations in the Wsp, Aws and Mws pathways (Figure 6A, 6B).

The other type was less wrinkly, had similar motility as the wild type and promoter

mutations upstream of the PFL_3078-3093 operon (Figure 6A, 6B).

Two types of fitness assays were performed, similarly as previously described (Lind,

et al. 2015) to measure differences in fitness between the different DGC mutants and

the alternative phenotypic solution with the mutation upstream of PFL_3078. The first

assay measures "invasion fitness" where the mutant is allowed to invade a wild type

population from an initial frequency of 1%. This confirms that the mutations are

adaptive and that mutants can colonize the air-liquid interface. The invasion assays

showed that all reconstructed mutants could rapidly invade an ancestral wild type

population (Figure 7A, Figure 7 – source data). Although there were significant

differences between selection coefficients of the mutants (one-way ANOVA $p <$

0.0001), no mutant was significantly different from the most common mutant (WspF

V271G, two-tailed t-test $P > 0.01$).

The second fitness assay measures "competition fitness" and here each mutant is

instead mixed 1:1 with the most common WS type (WspF V271G) at the start of the

competition. The competition assay showed that the ancestral wild type was rapidly

outcompeted by the mutants also at a 1:1 initial ratio (Figure 7B). There was

significant variation in fitness between the WS mutants (one-way ANOVA $P <$

0.0001) and the AwsX had significantly lower selection coefficient (two-tailed t-test P

< 0.009) compared to the reference WspF V271G and one of the MwsR mutants

(E1081K) had significantly higher selection coefficient (two-tailed t-test $P < 0.003$).

The alternative phenotypic solution used by the PFL_3078 promoter mutant resulted

in the lowest fitness ($s = -0.1$, two-tailed t-test $p < 0.005$) meaning that it is expected

to be rapidly outcompeted by the WS mutants (Figure 7B).

Figure 7. Fitness of reconstructed *P. protegens* Pf-5 WS mutants was measured in pairwise competitions. **(A)** Invasion fitness was measured relative a dominant ancestral wild type strain with a 1:100 initial ratio to show that mutations were adaptive and that they can increase from rare to colonize the air-liquid interface. Six independent competitions were performed for each pair. **(B)** Competition fitness was measured relative the most common WS mutant (WspF V271G) in a 1:1 initial ratio to compare the fitness of different WS mutants and the alternative phenotypic solution. Four independent competitions were performed for each pair.

SBW25 WS mutants use cellulose as the main structural component, but even though there is high parallelism at the genetic level for Pf-5 WS mutants this cannot be the case at the phenotypic level as its genome does not encode genes for cellulose biosynthesis. Given that production of Pel exopolysaccharide has been shown to be induced by mutations in *wspF* in *P. aeruginosa* (Hickman, et al. 2005) and that Pel in this species is required for pellicle formation under static growth (Friedman and

Kolter 2004) this was predicted to be the main structural component used by Pf-5. To

test this prediction the *pel* operon (PFL_2972-PFL_2978) was deleted from Pf-5 and

combined with previously characterized WS mutations and fitness was measured.

Both invasion fitness (Figure 8A) and competition fitness (Figure 8B) was

significantly lower (two-tailed t-tests $p < 0.01$) compared to isogenic strains with an

intact *pel* operon (Figure 7A, 7B) except invasion fitness for the AwsX mutant (two-

tailed t-tests $p < 0.08$, one outlier). This suggests that Pel polysaccharide serves as an

important structural component for colonizing the air-liquid interface and that its

production is activated by mutations leading to increased c-di-GMP levels. Although

deletion of *pel* in WS mutants resulted in less wrinkly colony morphology it did not

result in a smooth ancestral type. Neither did deletion of *pel* abolish the ability to

colonize the air liquid interface (Figure 8C) or the ability to invade wild type

populations (Figure 8A). This suggests that production of an additional EPS

component is induced by increased c-di-GMP levels caused by mutations in Wsp,

Aws and Mws, at least in the absence of *pel*. As expected if the motility defect

observed for WS mutants are primarily caused by high c-di-GMP levels rather than

high production of Pel, the motility was also reduced for Wsp, Aws and Mws mutants

with the *pel* operon deleted (Figure 8D).

Fig 8. Contribution of *pel* to WS phenotype and fitness (**A**). Deletion of *pel* in WS

mutants reduces invasion fitness. Fitness of reconstructed *P. protegens* Pf-5 WS

mutants without the *pel* operon was measured in pairwise competitions. Invasion

fitness was measured relative a dominant ancestral wild type strain with a 1:100 initial

ratio. Six independent competitions were performed for each pair. (**B**) Deletion of *pel*

in WS mutants reduces competition fitness. Competition fitness was measured

relative the most common WS mutant (WspF V271G) in a 1:1 initial ratio. Four

independent competitions were performed for each pair. (**C**) Deletion of *pel* in WS

mutants did not result in ancestral smooth colony morphology or loss of ability to

colonize the air-liquid interface suggesting a secondary EPS component is produced.

(**D**). Deletion of *pel* did not restore motility showing that Pel overproduction is not the

cause of the motility defect in WS mutants.

Discussion

The extension of the *P. fluorescens* SBW25 experimental evolution system to related species shows promise for true testing of evolutionary forecasting models. While there is a diversity of DGCs and EPSs between species leading to differences in forecasts, the conserved role of c-di-GMP and limited number of phenotypes allow the use of previous data to improve predictions and makes the experimental system robust to changes in environmental conditions. The test of initial forecasts for *P. protegens* Pf-5 presented here provides support for the ability to predict some aspects of both genetic and phenotypic evolution while recognizing that the probability of specific mutations cannot in most cases be predicted.

That experimental populations of *Pseudomonas* will colonize the air-liquid interface when incubated under static condition is a prerequisite of extending the model. Given that a range of phenotypic solutions is predicted to be available for all species the evolution of such mutants for *P. protegens* is not surprising. The specific environmental conditions used for experimental evolution often have a major impact on evolutionary outcomes and is also likely to influence relative fitness and possibly mutational biases also in the WS system. However despite major changes in growth medium, temperature and material and physical dimensions of the growth vessel, predictions on both the genetic and phenotypic levels proved successful demonstrating robustness to environmental change and the establishment of a dominant selective pressure, i.e. access to oxygen solved by air-liquid interface colonization.

Phenotypic predictions of the structural basis supporting air-liquid colonization is challenging given the limited previous experimental data. For SBW25 cellulose-based solutions are superior in fitness, but for Pf-5 this solution in not available. The prediction that overproduction of structural exopolysaccharides, rather than fuzzy, cell-chaining or mucoid types, would be the primary solution was successful. One of the two phenotypes found here used the Pel EPS, which could be predicted based on its role in *P. aeruginosa*. However it appears to use a secondary EPS as well, that remains to be identified, given that mutants lacking Pel but with activated DGCs still colonize the air-liquid interface and have a distinct colony morphology. The second

phenotype used another EPS, encoded by PFL_3078-3093, which had not previously
be described and given that several EPS loci are usually encoded in *Pseudomonas*
genomes its use could not be predicted. However, repeating experimental evolution
using other *Pseudomonas* species is likely to provide more information about which
EPSs can be used to colonize the air-liquid interface and their relative fitness to allow
improved phenotypic predictions. Deletion of the *pel* operon, the unidentified
secondary EPS and PFL_3078-3093 and subsequent experimental evolution could
reveal less fit phenotypic solutions that are expected to exist including fuzzy types
caused by defects in LPS modification, cell-chaining types with defects in cell
division, adhesive proteins or mucoid types using alginate or levan, two EPSs with
lower structural stability.
620
The general prediction of types of mutations, as described in the hierarchy in Figure
2D, was also successful although the relatively few mutants identified here did not
allow for detection of rare activating mutations or double inactivating mutations. The
existence of such types could be confirmed by deletion of the main pathways (Wsp,
Aws, and Mws) followed by experimental evolution as previously described for
SBW25 (Lind, et al. 2015). The majority of mutations were loss-of-function
mutations in negative regulators or interacting proteins followed by less common
promoter mutations and promoter captures. In contrast to SBW25, where all promoter
mutations resulted in up-regulation of DGCs, the mutation upstream of PFL3078-
3093 demonstrates the possibility of direct transcriptional activation of EPS
components that are not under post-translational control of c-di-GMP. Two identical
mutations were found over 9 kb upstream of the *aws* operon, in between a ribosomal
RNA operon and the *recCBD* operon, which encodes key genes for recombination.
The molecular effects of these mutations have not been further investigated, but the
resulting WS phenotype is dependent of the presence of the *aws* operon, deletion of
which reversed the phenotype. This is consistent with an up-regulation of c-di-GMP
by AwsR presumably caused by increased transcription. The mutation is located in
the predicted terminator of the ribosomal RNA operon and increased transcriptional
read-through could put the *aws* operon under control of a very strong *rrn* promoter
that is most highly transcribed during exponential growth. This could explain the
relatively mild colony morphology phenotype as well as high motility of this WS
mutant (Figure 6).

The mathematical null model (Lind, et al. 2018) successfully predicted that Wsp

would be the most commonly used pathway followed by Aws, and Mws the most

rare. However, the number of mutants isolated here is rather small and the high

frequency of Wsp mutants seems mainly to be caused by a mutational hot spot in

wspF. Still the prediction that the three pathways together would contribute the large

majority of adaptive mutations (40 out of 43) is not trivial given that in SBW25 at

least 13 additional pathways are available to the high fitness WS phenotype (Lind, et

al. 2015). It is also worth noting that direct use of mutation rate data from SBW25

(Lind, et al. 2018) would result in poorer predictions than the mathematical null

model due to a strong mutational hot spot in *aws*X in that species.

For the multi-protein pathways Wsp and Aws, the null model predicted (Figure 2E)

that mutations would primarily be found in WspA, WspE, WspF, AwsX and AwsR.

Mutations were detected in all these except in WspA and the majority was found in

the negative regulators WspF and AwsX. WspA mutations were not found in the

original study in SBW25 either (McDonald, et al. 2009), but this was shown to be due

to lower fitness relative WspF and WspE mutants rather than a lower mutation rate to

WS (Lind, et al. 2018). Possibly this explains the absence of WspA mutants here as

well, but it is not clear if this fitness difference would be conserved in other species or

if sometimes WspA mutants are more fit. Thus the null model prediction of equal

rates for WspA, WspF and WspE is not changed for future experimental tests.

The molecular effects of the mutations found here are unknown, but knowledge from

SBW25 and *P. aeruginosa* and their positions in protein structure allowed some

predictions to be made. Inactivating mutations in the negative regulator WspF were

predicted to be either indels or missense mutations in four specific regions. Mutations

were found in two of the predicted regions, one in the vicinity to the methylesterase

active site where mutations are predicted to cause large disruptions in protein

structure and the other one directly disrupting the phosphorylation active site in the

signal receiver domain. No mutations were found in the surface exposed regions

hypothesized to be involved in interactions with WspA and WspE, which could be

due to differences in function between SBW25 and Pf-5 or simply that they appear at

lower frequency and would be detected if additional mutations were isolated. The sole

mutation in WspE is, as predicted, located in the direct vicinity of the phosphorylation
active site. Mutations in AwsX were amino acid substitutions throughout the gene as
well as in frame deletions inactivating the gene as predicted. Mutations in AwsR and
MwsR were also found in predicted regions, but no mutations were found in the
periplasmic region of AwsR, which is the most commonly targeted region in SBW25.
Known mutational hot spots in *awsX, awsR* and *mwsR* in SBW25 (Lind, et al. 2018)
were not conserved in Pf-5 resulting in divergent spectra of mutations, while mutated
regions and predicted functional effects remain conserved between the two species.
685
The diversity of phenotypic solutions observed after experimental evolution is
dependent on fitness differences between the phenotypes, but also on the rate of
which phenotypes are introduced by mutations, which is dependent on the genetic
architecture underlying the trait as well as mutational biases. The Pf-5 strain has at
least three DGC pathways (Wsp, Aws and Mws) that are subject to negative
regulation leading to prediction of a high rate of WS mutants, which are then expected
to outcompete other phenotypic solutions. If instead only one of these pathways were
present, a larger diversity of phenotypes would be expected to be observed with
relative fitness becoming less important as the first mutant that gains a foothold at the
air-liquid interface will have a large advantage and priority effects, i.e. being first,
will increasingly determine which adaptive mutants are observed.
697
Given that the mutational target upstream of PFL_3078-3093 is likely be relatively
small and that these mutants are rapidly outcompeted by all WS types tested, their
relatively high frequency (3/43) is unexpected. Possibly this is due to a higher
mutation rate at these sites (Sankar, et al. 2016) or that population structure limits
direct competition between these different phenotypes and reduces the importance of
relative fitness. In SBW25 low fitness phenotypes that colonize the air-liquid
interface based on LPS modification or cell-chaining are observed prior to the rise of
WS to high frequencies (Lind, et al. 2017b) due to the presence of mutational hot
spots in these genes which make these mutants appear early during the growth phase
despite their relatively small mutational targets (Ferguson, et al. 2013; Farr 2015;
Lind, et al. 2017b).
709

In successfully predicting evolutionary outcomes in *P. protegens*, this work lays the foundation for future tests of evolutionary forecasting in related *Pseudomonas* species by clearly stating predictions on several different levels from phenotype down to which specific regions of proteins are likely to be mutated. Given what is already known about the effects of (for now) unpredictable mutational biases and differences in fitness between different WS types many of the forecasts will inevitably fail. However hopefully they will fail in interesting ways thereby revealing erroneous assumptions. The ability to remove common genetic and phenotypic pathways provides a unique opportunity to also find those pathways that evolution does not commonly use. This is necessary to determine why forecasts fail and update the predictive models for the next cycle of prediction, experimental evolution and mutant characterization to define the information necessary to predict short-term evolutionary processes.

Materials and methods

Strains and media

Pseudomonas protegens Pf-5 (previously known as *P. fluorescens* Pf-5) and derivatives thereof were used for all experimental evolution and phenotypic characterization. *E. coli* DH5α was used for cloning PCR fragments for genetic engineering (Paulsen et al). *P. protegens* Pf-5 was grown in tryptic soy broth (Tryptone 17g, Soytone 3g, Glucose 2.5g, NaCl 5g, K_2HPO_4 2.5g per liter) supplemented with 10 mM MgSO4 and 0.2% glycerol (TSBGM) for experimental evolution and fitness assays. Lysogeny broth (LB) was used during genetic engineering and LB without NaCl and supplemented with 8% sucrose was used for counter-selection of *sacB* marker. Solid media were 1.5% agar added to LB or TSB supplemented with 10 mM MgSO4, 0.2% glycerol and 10 mg/l Congo red. Motility assays were conducted in 0.3% agar TSB supplemented with 10 mM MgSO4, 0.2% glycerol. Kanamycin was used at 50 mg/l for *E. coli* or 80 mg/l for *P. protegens* and gentamicin at 10 mg/l for *E. coli* or 15 mg/L for *P. protegens*. Selection plates for cloning contained 5-Bromo-4-Chloro-3-Indolyl β-D-Galactopyranoside (X-gal) at 40 mg/l. 100 mg/L nitrofurantoin was used to inhibit growth of *E. coli* donor cells after conjugation. All strains were stored at -80°C in LB with 10% DMSO.

Experimental evolution

30 central wells of a deep well plate (polypropylene, 1.1 mL, round walls, Axygen

Corning Life Sciences) were inoculated with approximately 10^3 cells each and

incubated at 36°C for 5 days without shaking on two different occasions. Suitable

dilutions were plated on TSBGM plates with Congo red after 5 days and incubated at

36°C for 48 h. Plates were screened for colonies with a visible difference in colony

morphology and one divergent colony per well were randomly selected based only on

its position on the agar plate. In total 43 independent mutants were streaked for single

cells twice before overnight growth in LB and freezing.

Genome sequencing

Seven mutant strains that did not contain mutations in the *wspF* and *awsX* genes were

analyzed by genome resequencing. The strains had mutations in *awsR*, *mwsR*, *wspE,*

upstream PFL_3078 (2 strains) and in the intergenic region between *rrfB* and *recC*

upstream of the awsXRO operon. Genomic DNA was isolated with Genomic DNA

Purification Kit (Thermo Fisher). Sequencing libraries were prepared from 1μg DNA

using the TruSeq PCRfree DNA sample preparation kit (cat# FC- 121-3001/3002,

Illumina Inc.) targeting an insert size of 350bp. The library preparation was

performed according to the manufacturers' instructions (guide#15036187).

Sequencing was performed with MiSeq (Illumina Inc.) paired-end 300bp read length

and v3 sequencing chemistry.

Sequencing was performed by the SNP&SEQ Technology Platform in Uppsala. The

facility is part of the National Genomics Infrastructure (NGI) Sweden and Science for

Life Laboratory. The SNP&SEQ Platform is also supported by the Swedish Research

Council and the Knut and Alice Wallenberg Foundation. Sequencing data were

analyzed with using Geneious v. 10.2.3 with reads assembled against the *P. protegens*

Pf-5 genome sequence (CP000076.1).

Sanger sequencing

774 Sanger sequencing were performed by GATC biotech and used to sequence candidate

775 genes to find adaptive mutations and to confirm reconstructed mutations

776 (oligonucleotide primer sequences are available in Table S3).

Reconstruction of mutations

Nine mutations representing all candidate genes found using Sanger or Illumina
sequencing were reconstructed in the wild type ancestral *P. protegens* Pf-5 to show
that they are the cause of the adaptive phenotype and to be able to assay their fitness
effects without the risk of secondary mutations that might have occurred during
experimental evolution. A two-step allelic replacement protocol was using to transfer
the mutation into the ancestor. First a 1-2 kb fragment surrounding the putative
adaptive mutations were amplified using PCR (Phusion High- Fidelity DNA
polymerase, Thermo Scientific) and ligated into the multiple cloning site of the
mobilizable pK18mobsac suicide plasmid (FJ437239) using standard molecular
techniques. The ligation mix was then transformed into competent *E. coli* DH5α using
heat shock. After confirmation of correct insert size by PCR the plasmid was
transferred to *P. protegens* Pf-5 by conjugation with the donor strain and an *E. coli*
strain carrying the conjugation helper plasmid pRK2013. Cultures were grown
overnight of the recipient *P. protegens* Pf-5 (20 ml per conjugation at 30°C in LB)
and 2 ml each of the donor and helper *E. coli* strains per conjugation at 37°C in LB
with kanamycin. The culture of *P. protegens* Pf-5 was heat shocked for 10 minutes at
42°C prior to centrifugation at 4000 rpm for 10 minutes and resuspension in a small
volume of LB. Donor and helper cells were collected by centrifugation 4000 rpm for
10 minutes, resuspended in LB, and mixed with the concentrated recipient cells. After
another round of centrifugation the conjugation mix was resuspended in 50 μl LB and
spread onto several spots on a LA plate followed by incubation overnight at 30°C.
Each spot of the conjugation mix was scraped of the plate and resuspended in 200 μl
LB each and plated on LA plates with kanamycin, to select for transfer of the plasmid,
and nitrofurantoin that prevents growth of the *E. coli* donor and helper cells. The
pK18mobsac plasmid has a pBR322 type origin and cannot replicate in *P. protegens*
Pf-5 and only cells where the plasmid has integrated into the chromosome by
homologous recombination, with the homology provided by the cloned fragment, can
grow in the presence of kanamycin. After streaking for single cells on LA plates with
kanamycin, the *P. protegens* Pf-5 strains with integrated plasmids were grown
overnight in LB at 30°C without antibiotics to allow for double crossover
homologous recombination resulting in loss of the integrated plasmid. The plasmid

also contains the *sacB* marker conferring sucrose sensitivity, which allows for

counter-selection by plating on LA plates with sucrose. Sucrose resistant colonies

were checked for loss of the kanamycin marker and DNA sequencing of the cloned

region to find strains with the reconstructed mutation and no other mutations.

Deletion of the *pelABCDEFG* (PFL_2972-PFL_2978) region was accomplished using

the same two-step allelic exchange protocol using SOE-PCR to generated at fragment

surrounding the *pel* operon as previously described (Ferguson, et al. 2013; Farr 2015;

Lind, et al. 2017b). All oligonucleotide primer sequences are available in Table S3.

Fitness assays

Two types of competition fitness assays were performed similarly to previously

described (Ferguson, et al. 2013; Farr 2015; Lind, et al. 2015). The first assay

measures invasion fitness, where a mutant is mixed 1:100 with the wild type ancestor,

simulating early stages of air-liquid interface colonization where a rare mutant

establishes and grows at the surface with no competition from other mutants. The

second assay instead measures competition fitness in a 1:1 competition against a

reference mutant strain, which here was chosen to be the WspF V271G mutant

because it was the most commonly found in the experimental evolution study and thus

is highly successful either because of a high rate of emergence, i.e. a mutational hot

spot, or higher fitness than most other WS mutants. In addition, the WspF V271G

mutant has a temperature sensitive colony morphology phenotype in that it is highly

wrinkly at 30°C, but only have a very mild phenotype when grown at room

temperature, thus allowing it to be distinguishable from both the smooth ancestor and

all other wrinkly mutants isolated here.

Fluorescent reference strains of the wild type ancestor and the WspF V271G mutants

were created using a miniTn7 transposon (miniTn7(Gm) PA1/04/03 Gfp.AAV-a)

(Lambertsen, et al. 2004) that allows integration at a defined locus (attTn7) in the

chromosome. This allows the colonies to be distinguished not only by morphology

but also by fluorescence under blue/UV light and gentamicin resistance, which

provides a way to ascertain that secondary adaptive mutants that might occur during

the competition experiment do not bias the results (for example the ancestor could

evolve WS types or a WS mutant can evolve to cheat on the other type by inactivation

of EPS production or reduced c-di-GMP signalling). Introduction of the transposon

into *P. protegens* Pf-5 was performed by tri-parental conjugation from *E. coli* with

helper plasmids pRK2013 (conjugation helper) and pUX-BF13 containing the

transposase genes) using the same conjugation protocol described above.

The invasion assay was performed by mixing shaken overnight cultures of the

competitor 1:100 with the GFP-labeled reference ancestor followed by 1000-fold

dilution and static incubation at 36°C for 48 h in TBSGM medium in deep well plates

(1 ml per well, using only the central 60 wells). For the competition assay, the GFP-

labeled reference strain WspF 271G was mixed 1:1 with the competitor and diluted 6-

fold and grown for 4 h (shaken at 30°C), before plating to determine initial ratios, to

ensure the cells were in a similar physiological state at the start of the competition.

The competition cultures were then diluted 1000-fold in TBSGM medium and grown

in deep well plates (1 ml per well, using only the central 60 wells) static for 24 h at

36°C. Selection coefficients (s) were calculated as previously described (Dykhuizen

1990), where s = 0 is equal fitness, positive is increased fitness and negative is

decreased fitness relative to the reference strain. Briefly s is calculated as the change

in logarithmic ratio over time according to $s = [\ln(R(t)/R(0))]/[t]$, where R is the ratio

of mutant to reference and t is the number of generations of the entire population

during the experiment (estimated from viable counts). The cost of the fluorescent

marker were calculated from control competitions where the GFP-labeled reference

strains (wild type and WspF V271G) were competed against isogenic strains without

the marker and included in each plate under identical conditions during the fitness

assays and used to adjust the selection coefficients to compensate for the cost.

Motility assays

Swimming motility assays were performed in TBSGM plates with 0.3% agar (BD)

and the diameter was measured after 24 h of growth at room temperature. Each strain

was assayed in duplicates on two different plates.

Bioinformatics analysis of DGCs and EPS genes

Homologs for all DGCs in *P. fluorescens* SBW25 were found using the *Pseudomonas*

Ortholog Database at Pseudomonas.com (Winsor, et al. 2016). Blast-p searches for

GGDEF domains were performed to find remaining DGCs in the six *Pseudomonas*

species and their homologs again found using the *Pseudomonas* Ortholog Database

(Whiteside, et al. 2013) and manually inspected. Annotations (Pseudomonas.com. DB

version 17.2) were also searched for diguanylate cyclase and GGDEF. Not all DCCs

found are likely to have diguanylate cyclase activity, but given the difficulties of

predicting which of the partly degenerate active sites are likely to be inactive

combined with the possibilities of mutational activation during experimental

evolution, none were excluded.

There is no simple way to find all genes that can function as structural or regulatory

genes to allow colonization of the air-liquid interface. Thus the selection in Figure 3B

and Figure 3 – source data should not be considered complete. Putative EPS genes

were found using blastp searches with sequences from known exopolysaccharide

biosynthesis proteins including cellulose, PGA, Pel, Psl, Pea, Peb, alginate and levan.

Homologs were then found using the *Pseudomonas* Ortholog Database (Whiteside, et

al. 2013) at Pseudomonas.com (Winsor, et al. 2016). Annotations (Pseudomonas.com.

DB version 17.2) were also searched for glycosyltransferase, glycosyl transferase,

flippase, polysaccharide, lipopolysaccharide, polymerase, biofilm, adhesion and

adhesion. Based on previous work in SBW25 and literature searches a few additional

genes were added.

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Competing interests

902 The author declares no competing interests.

References

Bantinaki E, Kassen R, Knight CG, Robinson Z, Spiers AJ, Rainey PB. 2007.

Adaptive divergence in experimental populations of Pseudomonas fluorescens. III.

Mutational origins of wrinkly spreader diversity. Genetics 176:441-453.

http://dx.doi.org/10.1534/genetics.106.069906

Barrick JE, Lenski RE. 2013. Genome dynamics during experimental evolution.
Nature reviews. Genetics 14:827-839. http://dx.doi.org/10.1038/nrg3564

Barrick JE, Yu DS, Yoon SH, Jeong H, Oh TK, Schneider D, Lenski RE, Kim JF.
2009. Genome evolution and adaptation in a long-term experiment with
Escherichia coli. Nature 461:1243-1247. http://dx.doi.org/10.1038/nature08480

Beaumont HJ, Gallie J, Kost C, Ferguson GC, Rainey PB. 2009. Experimental
evolution of bet hedging. Nature 462:90-93. http://dx.doi.org/10.1038/nature08504

Blank D, Wolf L, Ackermann M, Silander OK. 2014. The predictability of molecular
evolution during functional innovation. Proceedings of the National Academy of
Sciences of the United States of America 111:3044-3049.
http://dx.doi.org/10.1073/pnas.1318797111

Brandis G, Pietsch F, Alemayehu R, Hughes D. 2015. Comprehensive phenotypic
characterization of rifampicin resistance mutations in Salmonella provides insight
into the evolution of resistance in Mycobacterium tuberculosis. J Antimicrob
Chemother 70:680-685. http://dx.doi.org/10.1093/jac/dku434

Conrad TM, Joyce AR, Applebee MK, Barrett CL, Xie B, Gao Y, Palsson BO. 2009.
Whole-genome resequencing of Escherichia coli K-12 MG1655 undergoing short-
term laboratory evolution in lactate minimal media reveals flexible selection of
adaptive mutations. Genome biology 10:R118. http://dx.doi.org/10.1186/gb-2009-
10-10-r118

Deatherage DE, Kepner JL, Bennett AF, Lenski RE, Barrick JE. 2017. Specificity of
genome evolution in experimental populations of Escherichia coli evolved at
different temperatures. Proceedings of the National Academy of Sciences of the
United States of America 114:E1904-E1912.
http://dx.doi.org/10.1073/pnas.1616132114

Dykhuizen DE. 1990. Experimental studies of natural selection in bacteria. Annu.
Rev. Ecol. Syst. 21:373-398.

Farr AD. 2015. Formalist features determining the tempo and mode of evolution in
Pseudomonas fluorescens SBW25. [[Auckland, New Zealand]: Massey University.

Farr AD, Remigi P, Rainey PB. 2017. Adaptive evolution by spontaneous domain
fusion and protein relocalization. Nat Ecol Evol 1:1562-1568.
http://dx.doi.org/10.1038/s41559-017-0283-7

Ferguson GC, Bertels F, Rainey PB. 2013. Adaptive Divergence in Experimental
Populations of Pseudomonas fluorescens. V. Insight into the Niche Specialist
"Fuzzy Spreader" Compels Revision of the Model Pseudomonas Radiation.
Genetics. http://dx.doi.org/10.1534/genetics.113.154948

Firnberg E, Labonte JW, Gray JJ, Ostermeier M. 2014. A comprehensive, high-
resolution map of a gene's fitness landscape. Molecular biology and evolution
31:1581-1592. http://dx.doi.org/10.1093/molbev/msu081

Fraebel DT, Mickalide H, Schnitkey D, Merritt J, Kuhlman TE, Kuehn S. 2017.
Environment determines evolutionary trajectory in a constrained phenotypic space.
Elife 6. http://dx.doi.org/10.7554/eLife.24669

Friedman L, Kolter R. 2004. Genes involved in matrix formation in Pseudomonas
aeruginosa PA14 biofilms. Mol Microbiol 51:675-690.

Gallie J, Libby E, Bertels F, Remigi P, Jendresen CB, Ferguson GC, Desprat N,
Buffing MF, Sauer U, Beaumont HJ, et al. 2015. Bistability in a metabolic network
underpins the de novo evolution of colony switching in Pseudomonas fluorescens.
PLoS biology 13:e1002109. http://dx.doi.org/10.1371/journal.pbio.1002109

Goymer P, Kahn SG, Malone JG, Gehrig SM, Spiers AJ, Rainey PB. 2006. Adaptive
divergence in experimental populations of Pseudomonas fluorescens. II. Role of
the GGDEF regulator WspR in evolution and development of the wrinkly spreader
phenotype. Genetics 173:515-526. http://dx.doi.org/10.1534/genetics.106.055863

Herron MD, Doebeli M. 2013. Parallel evolutionary dynamics of adaptive
diversification in Escherichia coli. PLoS biology 11:e1001490.
http://dx.doi.org/10.1371/journal.pbio.1001490

Hickman JW, Tifrea DF, Harwood CS. 2005. A chemosensory system that regulates
biofilm formation through modulation of cyclic diguanylate levels. Proceedings of
the National Academy of Sciences of the United States of America 102:14422-
14427. http://dx.doi.org/10.1073/pnas.0507170102

Jacquier H, Birgy A, Le Nagard H, Mechulam Y, Schmitt E, Glodt J, Bercot B, Petit

E, Poulain J, Barnaud G, et al. 2013. Capturing the mutational landscape of the

beta-lactamase TEM-1. Proceedings of the National Academy of Sciences of the

United States of America 110:13067-13072.

http://dx.doi.org/10.1073/pnas.1215206110

Jahn LJ, Munck C, Ellabaan MMH, Sommer MOA. 2017. Adaptive Laboratory

Evolution of Antibiotic Resistance Using Different Selection Regimes Lead to

Similar Phenotypes and Genotypes. Front Microbiol 8:816.

http://dx.doi.org/10.3389/fmicb.2017.00816

Jerison ER, Desai MM. 2015. Genomic investigations of evolutionary dynamics and

epistasis in microbial evolution experiments. Curr Opin Genet Dev 35:33-39.

http://dx.doi.org/10.1016/j.gde.2015.08.008

Jimenez JI, Xulvi-Brunet R, Campbell GW, Turk-MacLeod R, Chen IA. 2013.

Comprehensive experimental fitness landscape and evolutionary network for small

RNA. Proceedings of the National Academy of Sciences of the United States of

America 110:14984-14989. http://dx.doi.org/10.1073/pnas.1307604110

Kelley LA, Mezulis S, Yates CM, Wass MN, Sternberg MJ. 2015. The Phyre2 web

portal for protein modeling, prediction and analysis. Nat Protoc 10:845-858.

http://dx.doi.org/10.1038/nprot.2015.053

Kishimoto T, Iijima L, Tatsumi M, Ono N, Oyake A, Hashimoto T, Matsuo M,

Okubo M, Suzuki S, Mori K, et al. 2010. Transition from positive to neutral in

mutation fixation along with continuing rising fitness in thermal adaptive

evolution. PLoS genetics 6:e1001164.

http://dx.doi.org/10.1371/journal.pgen.1001164

Knoppel A, Knopp M, Albrecht LM, Lundin E, Lustig U, Nasvall J, Andersson DI.

2018. Genetic Adaptation to Growth Under Laboratory Conditions in Escherichia

coli and Salmonella enterica. Front Microbiol 9:756.

http://dx.doi.org/10.3389/fmicb.2018.00756

Koskiniemi S, Sun S, Berg OG, Andersson DI. 2012. Selection-driven gene loss in

bacteria. PLoS genetics 8:e1002787.

http://dx.doi.org/10.1371/journal.pgen.1002787

Kram KE, Geiger C, Ismail WM, Lee H, Tang H, Foster PL, Finkel SE. 2017.

Adaptation of Escherichia coli to Long-Term Serial Passage in Complex Medium:

Evidence of Parallel Evolution. mSystems 2.

http://dx.doi.org/10.1128/mSystems.00192-16

LaCroix RA, Sandberg TE, O'Brien EJ, Utrilla J, Ebrahim A, Guzman GI, Szubin R,

Palsson BO, Feist AM. 2015. Use of adaptive laboratory evolution to discover key

mutations enabling rapid growth of Escherichia coli K-12 MG1655 on glucose

minimal medium. Appl Environ Microbiol 81:17-30.

http://dx.doi.org/10.1128/AEM.02246-14

Lambertsen L, Sternberg C, Molin S. 2004. Mini-Tn7 transposons for site-specific

tagging of bacteria with fluorescent proteins. Environmental microbiology 6:726-

732. http://dx.doi.org/10.1111/j.1462-2920.2004.00605.x

Lee MC, Marx CJ. 2012. Repeated, selection-driven genome reduction of accessory

genes in experimental populations. PLoS genetics 8:e1002651.

http://dx.doi.org/10.1371/journal.pgen.1002651

Lee VT, Matewish JM, Kessler JL, Hyodo M, Hayakawa Y, Lory S. 2007. A cyclic-

di-GMP receptor required for bacterial exopolysaccharide production. Mol

Microbiol 65:1474-1484. http://dx.doi.org/10.1111/j.1365-2958.2007.05879.x

Liang ZX. 2015. The expanding roles of c-di-GMP in the biosynthesis of

exopolysaccharides and secondary metabolites. Nat Prod Rep 32:663-683.

http://dx.doi.org/10.1039/c4np00086b

Lind PA, Arvidsson L, Berg OG, Andersson DI. 2017a. Variation in Mutational

Robustness between Different Proteins and the Predictability of Fitness Effects.

Molecular biology and evolution 34:408-418.

http://dx.doi.org/10.1093/molbev/msw239

Lind PA, Farr AD, Rainey PB. 2017b. Evolutionary convergence in experimental

Pseudomonas populations. ISME J 11:589-600.

http://dx.doi.org/10.1038/ismej.2016.157

Lind PA, Farr AD, Rainey PB. 2015. Experimental evolution reveals hidden diversity

in evolutionary pathways. Elife 4. http://dx.doi.org/10.7554/eLife.07074

Lind PA, Libby E, Herzog J, Rainey PB. 2018. Disentangling the effects of genetic
architecture, mutational bias and selection on evolutionary forecasting. bioRxiv.
http://dx.doi.org/10.1101/335711

Long A, Liti G, Luptak A, Tenaillon O. 2015. Elucidating the molecular architecture
of adaptation via evolve and resequence experiments. Nature reviews. Genetics
16:567-582. http://dx.doi.org/10.1038/nrg3937

Lukacisinova M, Bollenbach T. 2017. Toward a quantitative understanding of
antibiotic resistance evolution. Curr Opin Biotechnol 46:90-97.
http://dx.doi.org/10.1016/j.copbio.2017.02.013

Lundin E, Tang PC, Guy L, Nasvall J, Andersson DI. 2017. Experimental
determination and prediction of the fitness effects of random point mutations in the
biosynthetic enzyme HisA. Molecular biology and evolution.
http://dx.doi.org/10.1093/molbev/msx325

McDonald MJ, Cooper TF, Beaumont HJE, Rainey PB. 2011. The distribution of
fitness effects of new beneficial mutations in Pseudomonas fluorescens. Biology
Letters 7:98-100. http://dx.doi.org/Doi 10.1098/Rsbl.2010.0547

McDonald MJ, Gehrig SM, Meintjes PL, Zhang XX, Rainey PB. 2009. Adaptive
divergence in experimental populations of Pseudomonas fluorescens. IV. Genetic
constraints guide evolutionary trajectories in a parallel adaptive radiation. Genetics
183:1041-1053. http://dx.doi.org/10.1534/genetics.109.107110

McElroy KE, Hui JG, Woo JK, Luk AW, Webb JS, Kjelleberg S, Rice SA, Thomas
T. 2014. Strain-specific parallel evolution drives short-term diversification during
Pseudomonas aeruginosa biofilm formation. Proceedings of the National Academy
of Sciences of the United States of America 111:E1419-1427.
http://dx.doi.org/10.1073/pnas.1314340111

Morgan JL, McNamara JT, Zimmer J. 2014. Mechanism of activation of bacterial
cellulose synthase by cyclic di-GMP. Nat Struct Mol Biol 21:489-496.
http://dx.doi.org/10.1038/nsmb.2803

O'Neill AJ, Huovinen T, Fishwick CW, Chopra I. 2006. Molecular genetic and
structural modeling studies of Staphylococcus aureus RNA polymerase and the

fitness of rifampin resistance genotypes in relation to clinical prevalence.

Antimicrob Agents Chemother 50:298-309.

http://dx.doi.org/10.1128/AAC.50.1.298-309.2006

Orgogozo V. 2015. Replaying the tape of life in the twenty-first century. Interface

Focus 5:20150057. http://dx.doi.org/10.1098/rsfs.2015.0057

Rainey PB, Travisano M. 1998. Adaptive radiation in a heterogeneous environment.

Nature 394:69-72. http://dx.doi.org/10.1038/27900

Romling U, Galperin MY, Gomelsky M. 2013. Cyclic di-GMP: the first 25 years of a

universal bacterial second messenger. Microbiology and molecular biology

reviews : MMBR 77:1-52. http://dx.doi.org/10.1128/MMBR.00043-12

Sandberg TE, Pedersen M, LaCroix RA, Ebrahim A, Bonde M, Herrgard MJ, Palsson

BO, Sommer M, Feist AM. 2014. Evolution of Escherichia coli to 42 degrees C

and subsequent genetic engineering reveals adaptive mechanisms and novel

mutations. Molecular biology and evolution 31:2647-2662.

http://dx.doi.org/10.1093/molbev/msu209

Sankar TS, Wastuwidyaningtyas BD, Dong Y, Lewis SA, Wang JD. 2016. The nature

of mutations induced by replication-transcription collisions. Nature.

http://dx.doi.org/10.1038/nature18316

Schenk MF, Szendro IG, Krug J, de Visser JA. 2012. Quantifying the adaptive

potential of an antibiotic resistance enzyme. PLoS genetics 8:e1002783.

http://dx.doi.org/10.1371/journal.pgen.1002783

Silby MW, Cerdeno-Tarraga AM, Vernikos GS, Giddens SR, Jackson RW, Preston

GM, Zhang XX, Moon CD, Gehrig SM, Godfrey SA, et al. 2009. Genomic and

genetic analyses of diversity and plant interactions of Pseudomonas fluorescens.

Genome biology 10:R51. http://dx.doi.org/10.1186/gb-2009-10-5-r51

Sleight SC, Orlic C, Schneider D, Lenski RE. 2008. Genetic basis of evolutionary

adaptation by Escherichia coli to stressful cycles of freezing, thawing and growth.

Genetics 180:431-443. http://dx.doi.org/10.1534/genetics.108.091330

Sommer MOA, Munck C, Toft-Kehler RV, Andersson DI. 2017. Prediction of
antibiotic resistance: time for a new preclinical paradigm? Nature reviews.
Microbiology 15:689-696. http://dx.doi.org/10.1038/nrmicro.2017.75

Spiers AJ, Bohannon J, Gehrig SM, Rainey PB. 2003. Biofilm formation at the air-
liquid interface by the Pseudomonas fluorescens SBW25 wrinkly spreader requires
an acetylated form of cellulose. Molecular Microbiology 50:15-27.

Spiers AJ, Kahn SG, Bohannon J, Travisano M, Rainey PB. 2002. Adaptive
divergence in experimental populations of Pseudomonas fluorescens. I. Genetic
and phenotypic bases of wrinkly spreader fitness. Genetics 161:33-46.

Spiers AJ, Rainey PB. 2005. The Pseudomonas fluorescens SBW25 wrinkly spreader
biofilm requires attachment factor, cellulose fibre and LPS interactions to maintain
strength and integrity. Microbiology 151:2829-2839.
http://dx.doi.org/10.1099/mic.0.27984-0

Steiner S, Lori C, Boehm A, Jenal U. 2013. Allosteric activation of exopolysaccharide
synthesis through cyclic di-GMP-stimulated protein-protein interaction. EMBO J
32:354-368. http://dx.doi.org/10.1038/emboj.2012.315

Tenaillon O, Rodriguez-Verdugo A, Gaut RL, McDonald P, Bennett AF, Long AD,
Gaut BS. 2012. The molecular diversity of adaptive convergence. Science
335:457-461. http://dx.doi.org/10.1126/science.1212986

Whiteside MD, Winsor GL, Laird MR, Brinkman FS. 2013. OrtholugeDB: a bacterial
and archaeal orthology resource for improved comparative genomic analysis.
Nucleic Acids Res 41:D366-376. http://dx.doi.org/10.1093/nar/gks1241

Whitney JC, Whitfield GB, Marmont LS, Yip P, Neculai AM, Lobsanov YD,
Robinson H, Ohman DE, Howell PL. 2015. Dimeric c-di-GMP is required for
post-translational regulation of alginate production in Pseudomonas aeruginosa. J
Biol Chem 290:12451-12462. http://dx.doi.org/10.1074/jbc.M115.645051

Wichman HA, Badgett MR, Scott LA, Boulianne CM, Bull JJ. 1999. Different
trajectories of parallel evolution during viral adaptation. Science 285:422-424.

Winsor GL, Griffiths EJ, Lo R, Dhillon BK, Shay JA, Brinkman FS. 2016. Enhanced
annotations and features for comparing thousands of Pseudomonas genomes in the

Pseudomonas genome database. Nucleic Acids Res 44:D646-653.

http://dx.doi.org/10.1093/nar/gkv1227

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Figure 5 - source data. Mutations found after experimental evolution

Genome position CP000076.1	Type	Change	Gene position	Gene locus	Gene symbol	Effect	Mutation found in SBW25
859290	Transition	g->a	-129	PFL_0740	recC	rrfB terminator/recCBD promoter upstream awsXRO	No
859290	Transition	g->a	-129	PFL_0740	recC	rrfB terminator/recCBD promoter upstream awsXRO	No
868744	Transversion	c->a	101	PFL_0743	awsX, yfiR	S34Y	No
868932	Transversion	t->g	289	PFL_0743	awsX, yfiR	Y97D	No
869088	Transversion	t->g	445	PFL_0743	awsX, yfiR	C149G	No
869281	Transversion	t->g	638	PFL_0743	awsX, yfiR	V213G	No
869281	Transversion	t->g	638	PFL_0743	awsX, yfiR	V213G	No
869281	Transversion	t->g	638	PFL_0743	awsX, yfiR	V213G	No
868832-868840	Deletion	Deletion 9 bp	189-197	PFL_0743	awsX, yfiR	Deletion F63-L65	No
868832-868840	Deletion	Deletion 9 bp	189-197	PFL_0743	awsX, yfiR	Deletion F63-L65	No
869040-869072	Deletion	Deletion 33 bp	397-429	PFL_0743	awsX, yfiR	Deletion V133-P143	No
869913	Transition	c->t	581	PFL_0744	awsR, yfiN	A194V	Yes
869994	Transition	a->g	662	PFL_0744	awsR, yfiN	D221G	D->A
870002	Transition	c->t	670	PFL_0744	awsR, yfiN	L226F	No
1301093	Transition	c->t	1982	PFL_1133	wspE	S661L	Yes
1302246	Transition	g->a	808	PFL_1134	wpsF	G270R	Yes
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302250	Transversion	t->g	812	PFL_1134	wspF	V271G	No
1302327	Transition	c->a	889	PFL_1134	wspF	Q297K	Q->R
1302327	Transition	c->a	889	PFL_1134	wspF	Q297K	Q->R
1301579-1301593	Duplication	Duplication 15 bp	141-155	PFL_1134	wspF	Duplication L51-I55	Del L51-I55
3548383	Transition	c->t	-98	PFL_3078		PFL_3078-PFL_3093 promoter	
3548383	Transition	c->t	-98	PFL_3078		PFL_3078-PFL_3093 promoter	
3548385	Transition	c->t	-96	PFL_3078		PFL_3078-PFL_3093 promoter	
6114402	Transversion	a->t	2237	PFL_5345	mwsR, morA	Q746L	No
6114696	Transition	a->g	2531	PFL_5345	mwsR, morA	H844R	No
6114940	Transition	g->a	2775	PFL_5345	mwsR, morA	M925I	Yes
6115236	Transversion	t->a	3071	PFL_5345	mwsR, morA	L1024Q	No
6115406	Transition	g->a	3241	PFL_5345	mwsR, morA	E1081K	Yes
6115406	Transition	g->a	3241	PFL_5345	mwsR, morA	E1081K	Yes
6115406	Transition	g->a	3241	PFL_5345	mwsR, morA	E1081K	Yes
6115406	Transition	g->a	3241	PFL_5345	mwsR, morA	E1081K	Yes
6115415	Transition	g->a	3250	PFL_5345	mwsR, morA	G1084R	G->S
6115107-6115130	Duplication	Duplication 24 bp	2942-2965	PFL_5345	mwsR, morA	P981-S988	No

Figure 7 - source data. Fitness assays

Invasion - selection coefficients

Wild type	0,003	0,002	-0,005	0,012	0,013	0,010	-0,039	0,021	-0,017
WspF V271G	0,363	0,330	0,303	0,318	0,319	0,340			
WspF Q297K	0,362	0,225	0,347	0,331	0,269	0,282			
WspE S661L	0,334	0,349	0,377	0,362	0,335	0,391			
AwsX del V133-P143	0,300	0,219	0,358	0,223	0,271	0,261			
AwsR A194V	0,308	0,351	0,280	0,262	0,292	0,262			
Promoter AwsXRO	0,412	0,526	0,476	0,348	0,336	0,376			
MwsR M925I	0,392	0,394	0,432	0,346	0,341	0,373			
MwsR E1081K	0,322	0,420	0,357	0,326	0,321	0,283			
PFL_3078 Promoter	0,313	0,351	0,320	0,321	0,334	0,326			

Competition - selection coefficients

Wild type	-0,230	-0,207	-0,196	-0,175				
WspF V271G	0,033	-0,021	-0,011	-0,029	-0,014	0,027	-0,015	0,010
WspF Q297K	0,015	0,017	-0,009	-0,010				
WspE S661L	-0,008	0,001	0,002	0,050				
AwsX ΔV133-P143	-0,030	-0,082	-0,088	-0,058				
AwsR A194V	-0,113	-0,079	0,008	-0,020				
Promoter AwsXRO	0,041	0,009	0,034	0,039				
MwsR M925I	-0,006	-0,007	-0,011	0,003				
MwsR E1081K	0,085	0,072	0,036	0,050				
PFL_3078 Promoter	-0,066	-0,146	-0,088	-0,102				

Figure 8 - source data. Fitness assays for del pel mutants

Invasion - selection coefficients

Δ*pel* WspF V271G	0,212	0,253	0,227	0,199	0,197	0,201
Δ*pel* WspF Q297K	0,210	0,162	0,173	0,176	0,212	0,223
Δ*pel* AwsX del V133-P143	0,311	0,222	0,184	0,202	0,203	0,178
Δ*pel* mwsR E1081K	0,225	0,209	0,248	0,204	0,185	0,205

Competition - selection coefficients

Δ*pel* WspF V271G	-0,171	-0,177	-0,206	-0,174
Δ*pel* WspF Q297K	-0,041	-0,114	-0,104	-0,069
Δ*pel* AwsX del V133-P143	-0,171	-0,104	-0,200	-0,151
Δ*pel* MwsR E1081K	-0,042	-0,025	-0,112	-0,072

Table S1. Oligonucleotide primers used for PCR, sequencing and genetic reconstructions

Cloning

pEX18Gm_MCS_F	tgttgtgtggaattgtgag
pEX18Gm_MCS_R	ctgcaaggcgattaagttg

***wsp* operon**

PFL_1133_EcoRI_F	agttgctggcggagaaaac
WspE_1550F	ttcccgctggcccatatcga
WspE_2119F	atcaccgacatcgacatgc
WspE_2213F	gtcctacaaggaccgtga
WspF_154R	caccggcatgatcaggtccat
WspR_70R	gcttcgccgatcatggcctg
WspR_398F	tggtggcgcgcattcgcta
WspR_553R	ccactccagctccaggta
WspR_620R	gtaggtcttgaagtagtcg

***aws* operon**

RecC_F_HindIII	tagaagcttctgataccgcccaagagttc
RecC_R_KpnI	catggtaccccagtcggctcgatatacct
recD_F_EcoRI	acggaattcccactacctgaatgtactg
PFL_0743-SalI_F	tgtgtgctgtcgaccattc
awsR_182R	agctgatggagcgggcgatca
awsR_677F	cgacttcaacgccctgct
awsO_DR	ttacatccgcgaggtgac
PFL_0746_F_HindIII	tgcaagcttcttcttcatcaccccgcaa

mwsR

mwsR_HindIII_F	gttaagcttagacccagctgttcctgttc
mwsR_2144F	gccgcgacatcagccagca
PFL_5345_SeqF	gaaaaggacctgcgcatg
PFL_5345_SeqR	gaactgcttgaggtagttc
mwsR_3599R	gaaggtgcggtcgatcttca
mwsR_SalI_R	aggccgtcgacgaaggtgc

PFL_3078-3093

PFL_3077_KpnI_F	catggtaccgcgaaagtcccggttgaag
PFL_3078_HindIII_R	tagaagcttgttggccatctcgttcatg

pelABCDEFG

PFL_2971_F_EcoRI	ctggaattcagcgagtactacctggactt
PFL_2972_DelCh	caaggccaatgcggtaaaca
PFL_2972_UR_SOE	attgggccctctatgtcgacatcaactcactcttcagtagatcaatct
PFL_2978_DF_SOE	gatgtcgacatagagggcccaatttgaggagcatcggcaagc
PFL_2979_R_HindIII	agtgcccaagagcagaagc