Supplemental Information

Evolutionary Remodeling of Bacterial Motility

Checkpoint Control

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Figure S1. Experimental Evolution of *E. coli* Spreading and Its Characterization. Related to Figure 1. (A) Examples of time course of evolution of spreading in TBSA for lines L3, L4 and L7. (B) Competition assay in spreading between evolved strains and reference parental strain RP437. Evolved strains (labeled with CFP; pVS129) were co-inoculated 1:1 with RP437 (labeled with YFP; pVS132). Samples were collected at indicated positions along the colony spreading in TBSA (left panel) and fractions of both strains at each position were quantified using flow cytometry (right panels). Data in (A) and (B) are represented as mean ± SEM of three replicates. (C) Correlation between spreading in TBSA and biased movement (chemotactic bias) in minimal medium SA with pre-established gradient (0 - 100 µM) of MeAsp. Chemotactic bias was quantified as a ratio of spreading up the gradient to the spreading normal to the gradient (a/b). Scale bar is 1 mm. (D) Correlation between spreading in TBSA and chemotactic drift velocity in MeAsp gradient established in a liquid-filled PDMS chamber. Gradient formation in the microfluidic device was monitored using fluorescein. The channel dimensions are 2 x 1 mm. Correlation values are 0.65 (C) and 0.78 (D). Line in (D) indicates an exponential fit.
Figure S2. Regression Analysis of Correlation between Spreading, Motility and Flagellation in Evolved Strains. Related to Figure 2. (A) Correlation between spreading and velocity, tumbling rate and bias for motility analysis performed in liquid in absence of chemical gradients. Chemotactic bias in liquid was defined as chemotactic drift (Fig. S1D) normalized by cell swimming velocity. Asterisks indicate significant differences from PR437 according to two-tailed t-test ($P < 0.05$). Regression analysis was performed using linear regression model as described in the Supplemental Experimental Procedures. (B) Correlation between velocity and flagellar length, flagellar number and motor speed.
Rotation rate of individual flagellar motors in evolved cells was determined by tracking 1.1 µm beads attached to flagella. Electron microscopy was used to quantify flagellar length and number in evolved strains. (C) Correlation between tumbling rate and flagellar length, flagellar number and motor speed. (D) Western Blot analysis of flagellin levels inside and outside the cell for RP437 and evolved strains.
Figure S3. Changes in Gene Expression during Evolution. Related to Figure 2. (A-C) Regression analysis of correlation between flagellar length (A), flagellar number (B) or motor speed (C) and $P_{fliC}$, $P_{fliE}$ or $P_{flhD}$ promoter activities. Regression analysis was performed using linear regression model as described in the Supplemental Experimental Procedures. (D) Titration of FliC (left) or MotA and MotB (right) either in RP437 or in corresponding knockout strains. (E) Example of time evolution of spreading efficiency for L1 strain grown in liquid TB (left) and spreading efficiency after 20-30 rounds of selection for individual lines of evolution (right). Asterisks indicate significance for decrease (green)
or increase (red) in spreading compared to RP437 according to t-test ($P < 0.05$). Data in (D and E) are represented as mean ± SEM of three replicates. (F) Genes that changed their transcription independent of selection for motility according to the RNAseq analysis.
Figure S4. Analysis of Effects of Individual Mutations. Related to Figure 3. (A) $P_{fliC}$ promoter activity for RP437 strain carrying individual adaptive mutations, normalized to RP437. Significant increase (black asterisks, $P < 0.05$; red asterisks, $P < 0.01$) in expression compared to the parental strain RP437 is indicated. (B) Regression analysis of correlation between spreading in TBSA and velocity, tumbling and chemotactic bias. Regression analysis was performed using linear regression model as described in the Supplemental Experimental Procedures. (C) Effects of R82C mutation in $sspA$ on flagellar length and number. Asterisks indicate significant increase in number and length of flagella compared to RP437 according to $t$-test ($P < 0.05$). Data in (A and C) are represented as mean ± SEM of three replicates.
Figure S5. FlgM Secretion Controls on Spreading in TBSA. Related to Figure 4. (A,B) Spreading of evolved strains (A) or of RP437 carrying individual mutations (B), either in presence of FlgM (black bars) or in flgM knockout strain (grey bars). In each case, spreading was normalized to the respective reference strain, RP437 or RP437ΔflgM. Asterisks indicate significant difference between relative spreading in FlgM+ and ΔflgM backgrounds (black: $P < 0.05$; red: $P < 0.01$) according to two-tailed $t$-test. (C) Correlation between FliC secretion efficiency and $P_{\beta C}$ promoter activity in strains containing individual mutation.
## Table S1. Mutations Found in Evolved Strains by Genome Resequencing. Related to Figure 3.

| Function | Genes | Annotation | L1 | L2 | L3 | L4 | L5 | L6 | L7 |
|----------|-------|------------|----|----|----|----|----|----|----|
| Motility | flhI   | ATPase of flagellar export | E258D | M178I | G228R | V38E | E221Q |
|          | flhM   | flagellar motor switch protein | M67I |
|          | flhG   | flagellar motor switch protein | R123H |
|          | tar    | chemoreceptor | L310F |
|          | cheZ   | phosphatase for CheY-P | F127L |
|          | flgM   | anti-sigma factor for FliA | T22I |
| Regulatory factor | sevT   | serine tRNA | C>T<sup>a</sup> |
|          | sspA   | stringent starvation protein A | R82C |
|          | yciH   | translation initiation factor | Q36Stop |
| Cell envelope biogenesis | ldtB   | L, D-transpeptidase | IS1 |
|          | mdoH   | membrane glycosyltransferase | N757Y |
|          | idi    | isopentenyl diphosphate | L72P |
| Metabolism | hisA   | L-histidine biosynthesis | N73I |
|          | argH   | L-arginine biosynthesis | Q10Stop |
|          | glpK   | glycerol kinase | I73S |
|          | ydfG   | 3-hydroxy acid dehydrogenase | IS1<sup>b</sup> |
| Transporter | cusC   | copper / silver efflux system | IS1 |
|          | lacY   | lactose permease | Del |
|          | yeoO   | unknown | IS1 |
| Unknown | yceO   | possible regulator of biofilm | IS1 |
|          |        | IS1 |

<sup>a</sup>Mutation from guanine to adenine at position 31 in sevT RNA.

<sup>b</sup>Insertion of transposable IS1 sequence.
Table S2. Mutations in Motility Genes Found in Strains Evolved in Minimal Medium. Related to Figure 7.

| Genes | Annotation | Asp₁ | Asp₂ | Ser₁ | Ser₂ | Glc₁ | Glc₂ | MA₁ | MA₂ |
|-------|------------|------|------|------|------|------|------|-----|-----|
| fliH  | regulatory component of FliI | Q217H |      |      |      |      |      |     |     |
| fliI  | ATPase of flagellar export apparatus | R231L | V301A | T377M | V70L |      | S106C |     |     |
| fliM  | flagellar motor switch protein |      |      |      |      |      |      |     | T192N |

*All lines were evolved in Minimal A medium SA with 0.2% glucose as carbon source. Lines Asp₁, Asp₂, Ser, and Ser₂ were additionally supplemented with 1 mM aspartate or serine, respectively, as chemoattractant. For lines Glc₁ and Glc₂ glucose was both carbon source and chemoattractant. Lines MA₁ and MA₂ were evolved in a pre-established gradient of MeAsp (see Fig. S1C).
SUPPLEMENTAL EXPERIMENTAL PROCEDURES

Strains and Plasmid Construction
pKOV was used to integrate single SNP mutation into the genome of wild type strain (Link et al., 1997). Gene deletions were constructed by PI transduction from KEIO collection (Baba et al., 2006). Kanamycin resistance cassette was removed using pCP20 (Cherepanov and Wackernagel, 1995). GFP reporters for flhD and fliC promoters were constructed based on pUA66 (Zaslaver et al., 2006), yielding pAM104 and pAM109, respectively. Promoter reporters for other genes were from a pUA66-based promoter library (Zaslaver et al., 2006). pTrc99a vector inducible by isopropyl β-D-1-thiogalactopyranoside (IPTG) was used to express FliA (pBN2), FlgM (pBN3), FlgM fused C-terminally to HA tag (pBN4) or FliC (pBN5). pBAD18-Kan vector inducible by arabinose was used to express MotAB (pBN6). pKAF131 carrying sticky fliC allele under native fliC promoter (Yuan et al., 2010) was a gift of Karen A. Fahrner and Howard C. Berg. For bead assay, fliC knockout strains carrying pKAF131 were used.

Immunoblot Analysis of FlgM or FliC Secretion
Cell cultures were adjusted to the same density (OD$_{600}$ = 0.5-0.6) and centrifuged at 4,000 rpm for 8 min. Proteins in cell free supernatant were precipitated by 25 % trichloroacetic acid (TCA) and incubated on ice for 5 min. Precipitated proteins were collected by centrifugation (10,000 rpm) at 4 °C and washed twice with ice-cold acetone. After drying, 1X sample buffer containing 8 M urea was used to dissolve pellet for SDS-PAGE. Samples were separated on 15 % PAGE and transferred to PVDF membrane using wet blotting. Primary anti-HA antibody (Sigma) at a 1:10,000 dilution and secondary anti-mouse IgG antibody labeled with peroxidase (Sigma) at a 1:5,000 dilution were used for detection.

Regression Analysis
In order to assess relative contributions of individual independent variables on the dependent, we constructed a multilinear regression model, where the dependent variable (y) is assumed to be a linear additive function of the independent variables ($x_1,x_2,x_3$)

$$ y = a_0 + a_1x_1 + a_2x_2 + a_3x_3 $$

Linear least square method was used to calculate the values of coefficients. The value of $R^2$ gives an estimate of the proportion of variance of the dependent variable predictable from the full multilinear regression model. Statistical significance of estimating each parameter of the model is measured by calculating the p-value of the t-statistics. The p-value of the parameter is the measure of the corresponding independent variable's relative contribution in predicting the dependent variable. The significance of the full model is calculated by estimating the p-value of the F-statistics of the full model. Since in the full model it is assumed that the independent variables are uncorrelated, we further constructed linear models where one of the independent parameters ($x_1$) is removed from the model so that the dependent variable is

$$ y = a_0 + a_2x_2 + a_3x_3 $$

and calculated $R^2$. The gain in $R^2$ by adding the variable is a measure of effect of the variable $x_1$ in explaining the variance of the dependent variable:

$$ \Delta R^2(x_1) = R^2(\text{full model}) - R^2(\text{model after removing } x_1) $$

The values of $\Delta R^2$ corroborate the significances calculated from the p-values of the full model. The calculations were performed using the lm function of the software R.
Spearman Rank Correlation (\( \rho \))

The association between each pair of variables was calculated using Spearman rank correlation (\( \rho \)). The values of each variable were ranked and the correlation between the ranks was calculated. Since rank correlation does not assume any functional relationship between the variables, it is appropriate when the relationship is unknown. The corresponding significance of the correlation value was estimated by Spearman rho test which computes a probability of the correlation value being zero. The calculations were performed by the `cor.test` function of the software R.

Evolutionary Algorithm

Evolutionary simulations for topologies shown in Figure 7B were performed in MATLAB using standard genetic algorithm framework (Mitchell, 1998). Each simulation starts with a homogeneous population of 500 individual networks with the same topology and identical parameter values. In each generation, one of the model parameters is mutated randomly from a uniform distribution from 0 to 1 in each of the individual networks selected by the mutation rate 0.8 (i.e., in 80% of the networks). The I/O behavior of the network is modeled as a linear relationship. Thus, for the linear network with one input (\( x \)), one intermediate (\( y \)) and one output (\( z \)),

\[
\begin{align*}
y &= y_T w_1 x \\
z &= z_T w_2 y
\end{align*}
\]

Here \( w_1 \) and \( w_2 \) denote the strength of the links between the input and the intermediate and between the intermediate and the output, respectively; \( y_T \) and \( z_T \) represent maximal attainable values of \( y \) and \( z \), respectively, for one incoming link.

For the linear network with three inputs and three intermediates but one output,

\[
\begin{align*}
y_1 &= y_T w_1 x_1; y_2 = y_T w_2 x_2; y_3 = y_T w_3 x_3; \\
z &= z_T (w_4 y_1 + w_5 y_2 + w_6 y_3)
\end{align*}
\]

For the bow-tie network,

\[
\begin{align*}
y &= y_T (w_1 x_1 + w_2 x_2 + w_3 x_3) \\
z &= z_T w_4 y
\end{align*}
\]

The fitness is defined to be proportional to the exponential of the output,

\[ F = e^{(0.1+2)} \]

and it is calculated for each individual network in the population. Notably, in our model the maximal attainable fitness of the network is determined by the number of inputs that can acquire mutations, so that it is lower for a linear network with one input but the same for the linear and bow-tie networks with three inputs.

During in-silico evolution, the networks for the next generation are selected among the parental networks with replacement following proportional reproduction with the exponential scaling (Lampert and Tlusty, 2009; Burda et al., 2010; Friedlander et al., 2015). The fitness of the whole population is calculated as average fitness over the 500 networks. For each case the simulation was repeated for 100 different initial parameter sets so that evolution starts from a different point in the genotype space in order to avoid dependence on the initial parameter values. Median value of the fitness over these 100 different evolutionary runs was taken as a measure of the time trace of evolution.

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