

# Optimal methylation noise for best chemotactic performance of *E. coli*

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August 11, 2017

# Introduction

- ▶ Behavior of a cell is controlled by the intracellular biochemical reactions in its signaling pathway
- ▶ These reactions crucially depend on the expression levels of each protein involved
- ▶ Any fluctuations in the numbers of protein molecules have important consequences on the cell performance
- ▶ Small number of protein molecules inside a single cell: fluctuations are expected
- ▶ How the variability in protein numbers affect the cell behavior?

## *E.coli* Chemotaxis

- ▶ Chemotaxis pathway of *E. coli* bacteria: a model system to study cellular behavior
- ▶ Bacterial cells sense and respond to the changes in the nutrient concentration in its environment
- ▶ In presence of a concentration gradient, it moves up the gradient to reach a region of higher nutrient concentration
- ▶ The motion of an *E. coli* is controlled by flagellar motors
- ▶ CCW  $\rightarrow$  run: cell moves in one direction with speed  $\sim 20\mu m/s$
- ▶ CW  $\rightarrow$  tumble: random rotation with no net displacement
- ▶ By modulating the rotational bias of the motors the cell can modulate its run and tumble durations

## Chemotaxis pathway: adaptation

- ▶ Chemoreceptors in the cell membrane binds to nutrient molecules
- ▶ In a bound state the receptors suppress the phosphorylation of cytoplasmic protein CheA
- ▶ The methylation level of the receptors is controlled by two proteins CheR and CheB
- ▶ CheR raises the level and phosphorylated CheB-P causes demethylation
- ▶ In a phosphorylated state, CheA transfers the phosphate group to CheY and CheB
- ▶ When the phosphorylation of CheA is suppressed, CheB-P concentration decreases, demethylation process stops
- ▶ The receptor then reaches a highly methylated state, which in turn raises the activity of CheA
- ▶ Adaptation in the network

## Chemotaxis pathway: response

- ▶ CheY which also receives phosphate group from CheA, controls the sensing mechanism of the network
- ▶ CheY-P binds to flagellar motors and increases its CW bias, causing the cell to tumble
- ▶ In absence of phosphorylation CheY-P concentration goes down which reduces CW bias and the cell swims smoothly

## Noise in the biochemical pathway

- ▶ In absence of any noise in the signaling pathway, the switching of rotational bias of the flagellar motors is expected to be a Poisson process
- ▶ Duration of a particular run or tumble should follow an exponential distribution
- ▶ Switching events of a single cell in an isotropic medium were monitored in experiment and the residence time of the motors in the CCW bias was found to follow a power law distribution [Korobkova *et al.* Nature (2004)]
- ▶ Noise present in the signaling network of a single cell makes it possible to have large fluctuations in the CCW lifetimes and consequently, the cell can execute really long runs with significant probability

## Fluctuations in CheY-P level

- ▶ CCW and CW bias states of the motors modelled as a two-level system whose energy levels depend on the concentration of the motor protein CheY-P [Tu and Grinstein, PRL (2005)]
- ▶ As the noise present in the network causes this protein number to fluctuate, the energy levels also fluctuate with time
- ▶ Such fluctuations give rise to power law distribution for the lifetime of the CCW state
- ▶ For a single bacterial motor, when CW bias is large, which corresponds to higher CheY-P level, and hence smaller fluctuations, the CCW intervals show exponential distribution [Korobkova *et al.* PRL (2006)]
- ▶ For small CW bias, when fluctuations in CheY-P level are more significant, CCW intervals show power law distribution

## Noise and chemotactic performance

- ▶ Large fluctuations present at the single cell level do not impair the chemotactic response or robust adaptation observed at the population level
- ▶ Overall chemotactic performance improves as a result of interplay between the signaling noise and multiple flagellar motors [Sneddon *et al.* PNAS (2012)]
- ▶ Presence of multiple motors brings down the motor response time—beneficial in steep gradient of chemoattractant concentration
- ▶ Presence of noise in the signaling pathway generates longer runs and improves the chemotactic performance in shallow gradients
- ▶ Large variability in a cellular population ensures that different cells behave in different ways and each type of behavior may be suitable for one particular type of environment [Frankel *et al.* eLife (2014)]



# Methylation noise

- ▶ How noise affects the chemotactic efficiency of a single cell?
- ▶ For an efficient chemotactic performance, the cell should be able to find the nutrient-rich regions quickly and localize there
- ▶ The most important source of noise is the methylation demethylation reactions in the network
- ▶ Time-scales of these reactions are order of magnitude larger than all other time-scales present in the pathway
- ▶ Slow methylation fluctuations cannot be integrated out by the downstream processes in the network

## Different aspects of chemotactic performance

- ▶ **Chemotactic drift velocity**: average velocity with which the cell climbs up the chemical concentration gradient
- ▶ Larger values of drift velocity indicates a better performance
- ▶ **Localization**: Nutrient concentration, averaged over the steady state distribution of the cell position
- ▶ Localization takes a high value, when in the long time limit most of the cells are present in the regions which contains maximum nutrient
- ▶ High values of localization and drift velocity ensures a good chemotactic performance in the long time limit
- ▶ **Non-monotonic variation of localization and drift velocity as a function of methylation noise strength**
- ▶ **An optimal noise strength at which each of these quantities becomes maximum**

## Explanation from detailed CheY-P level statistics

- ▶ Chemotactic response is the result of differential behavior of the cell up and down the nutrient concentration gradient
- ▶ For large signaling noise, the tumble frequency is almost totally controlled by the stochastic fluctuations of the methylation level, and not by the local nutrient concentration
- ▶ Above difference becomes small  $\Rightarrow$  poor performance
- ▶ Why the performance gets worse for very low noise level?
- ▶ When CheY-P level falls below a certain threshold value, the cell shows a tendency to migrate towards regions of low nutrient concentration
- ▶ This affects its overall chemotactic performance adversely
- ▶ This threshold value decreases with noise strength, and hence the performance improves with noise
- ▶ Within our model, we give a clear explanation behind this threshold behavior

- ▶ In a shallow ligand gradient, the chemotactic drift velocity shows a peak at a specific noise strength, while the localization remains constant at low noise level [Flores *et al.* PRL (2012), He *et al.* Biophys J (2016)]
- ▶ Existence of a threshold behavior has never been studied before
- ▶ Threshold of CheY-P level below which the behavior of the cell is detrimental to its chemotactic performance, plays a crucial role when noise is low

## Model description

- ▶  $a(t)$ : activity of the receptor complex  
 $m(t)$ : methylation level  
 $y_P(t)$ : CheY-P level
- ▶ probability that a given receptor complex is in an active state

$$a = \frac{1}{1 + \exp(N\epsilon(m, [L]))}$$

$\epsilon(m, [c])$  is the free energy difference between the active and inactive states:

$$\epsilon(m, [L]) = f_m + f_{[L]} = \alpha(m_0 - m) - \log \left( \frac{1 + c(x)/K_A}{1 + c(x)/K_I} \right)$$

- ▶  $N = 6$ ,  $K_A = 3mM$ ,  $K_I = 18.2\mu M$ ,  $\alpha = 1.7$ ,  $m_0 = 1$   
[Jiang *et al.* PLoS Comp Biol (2010)]

# Stochastic methylation dynamics

$$\frac{dm}{dt} = k_R(1 - a) - k_B a + \eta(t)$$

- ▶  $\langle \eta(t)\eta(t') \rangle = \lambda(k_R(1 - \bar{a}) + k_B\bar{a})\delta(t - t')$ , and  $\bar{a} = 1/2$
- ▶ Dimensionless parameter  $\lambda$  controls noise strength
- ▶  $k_R = k_B = 0.015s^{-1} \Rightarrow$  methylation fluctuation is a slow process  $\Rightarrow \eta(t)$  cannot be integrated out
- ▶ Fluctuations in methylation level cause fluctuations in activity which in turn affects the phosphorylation of CheY proteins

$$\frac{dy_P}{dt} = k_Y a(1 - y_P) - k_Z y_P$$

$y_P$ : fraction of phosphorylated CheY proteins,  $k_Y = 1.7s^{-1}$   
and  $k_Z = 2s^{-1}$

- ▶ In the phosphorylated state, CheY-P proteins bind to the flagellar motors and cause the cell to tumble
- ▶ Tumbling rate  $\omega(y_P)$  is a sigmoidal function of  $y_P$

$$\omega(y_P) \sim y_P^{10}$$

- ▶ Instantaneous tumbling considered: a finite tumbling duration does not affect our conclusions
- ▶ Motion of the cell in one dimension
- ▶ Recent experiments study bacterial chemotaxis in narrow micro-fluidic channel whose width is comparable to the average run length of the cell [Li *et al.* PRL (2017), Blinz *et al.* Microelectron. Eng. (2010)]

## Simulation details

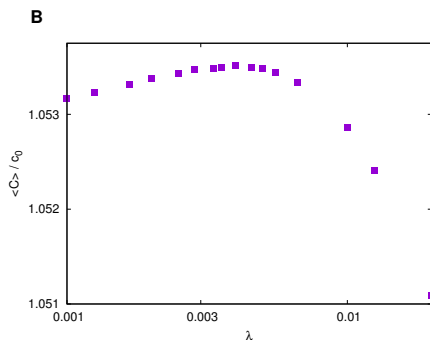
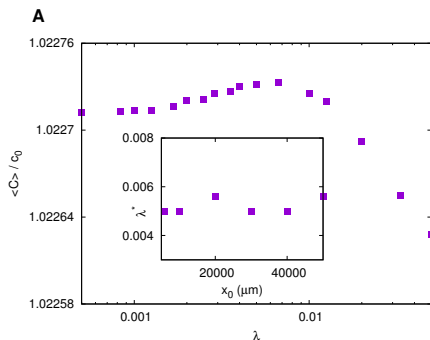
- ▶ 1d box of length  $L$ , with reflecting walls at two ends
- ▶ In a time-step  $dt$ , the cell moves a distance  $vdt$
- ▶  $a(t)$ ,  $m(t)$  and  $y_P(t)$  are updated
- ▶ the tumbling probability  $\omega(y_P)dt$  is calculated
- ▶ If a tumble does take place, the sign of  $v$  is reversed with probability  $q$
- ▶  $L = 1000\mu m$ ,  $v = 10\mu m/s$ ,  $dt = 0.01s$
- ▶ Finite size effects negligible



## Steady state distribution of cell position

- ▶  $\mathcal{P}_\lambda(x)$  : steady state probability to find the cell at position  $x$ , for a given noise strength  $\lambda$
- ▶  $\mathcal{P}_\lambda(x)$  should be large whenever  $c(x)$  is large and  $\mathcal{P}_\lambda(x)$  should be small where nutrient is sparse
- ▶ Average nutrient concentration experienced by the cell population in steady state :  $\langle C \rangle = \int_0^L dx c(x) \mathcal{P}_\lambda(x)$
- ▶ Integrand has a large value only when both  $c(x)$  and  $\mathcal{P}_\lambda(x)$  are large, indicating strong localization in favorable region
- ▶  $\langle C \rangle$  shows a non-monotonic variation with noise strength  $\lambda$
- ▶ There is an optimum level of the signaling noise when the chemotactic performance, as measured by  $\langle C \rangle$ , is at its best

# Localization vs noise



- ▶ For linear concentration profile best chemotaxis is observed for  $\lambda = \lambda^* \simeq 0.007$
- ▶  $\lambda^*$  does not depend strongly on the concentration gradient
- ▶ For a Gaussian  $c(x)$  also a similar  $\lambda^*$  obtained

## Chemotactic drift velocity in steady state

- ▶ Runs in the direction of increasing concentration of the chemo-attractant are extended and those in the opposite direction are shortened
- ▶ An overall drift motion up the concentration gradient
- ▶ Large drift velocity indicates good chemotactic performance
- ▶ Chemotactic drift velocity of the cell in presence of a linear concentration profile of the nutrient
- ▶ Consider an arbitrary position  $x$  where the cell tumbles and a new run begins
- ▶  $N_R(x)$  and  $N_L(x)$  are total number of rightward and leftward runs starting at  $x$ , within an observation time window  $t_{obs}$
- ▶  $d_R(x)$  and  $d_L(x)$  are total durations of these rightward and leftward runs

Average run duration (in either direction) starting at  $x$

$$\tau(x) = \frac{d_R(x) + d_L(x)}{N_R(x) + N_L(x)}$$

Probability that a run starts from the position  $x$

$$Q_{tum}(x) = \mathcal{N}^{-1}[N_R(x) + N_L(x)]$$

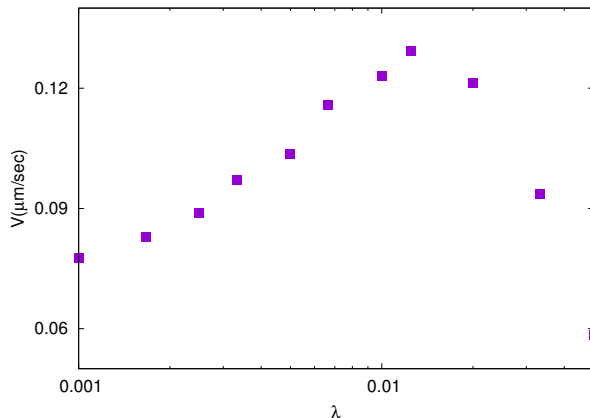
Average displacement in a run

$$\Delta = \int dx Q_{tum}(x) v \frac{d_R(x) - d_L(x)}{N_R(x) + N_L(x)}$$

Chemotactic drift velocity =  $\frac{\text{average displacement in a run}}{\text{average run duration}}$

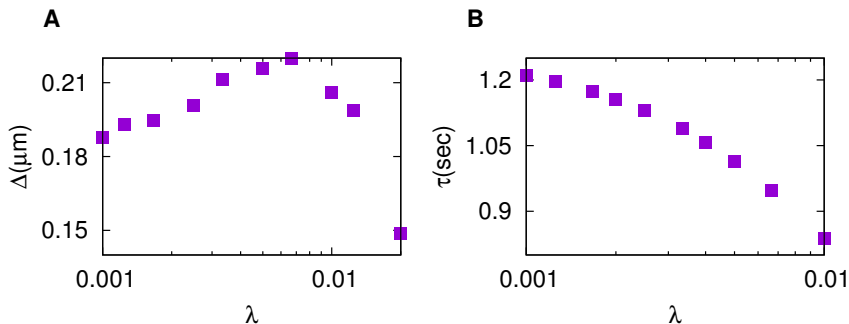
$$V = \frac{\Delta}{\tau} = \frac{v \int dx [d_R(x) - d_L(x)]}{\int dx' [d_R(x') + d_L(x')]}$$

## Drift velocity vs noise



- ▶ Position of the peak does not match exactly with that for localization  $\langle C \rangle$

## $\Delta$ and $\tau$ for different noise strengths



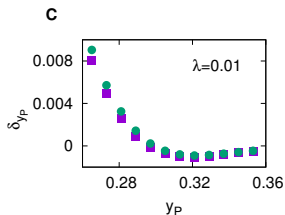
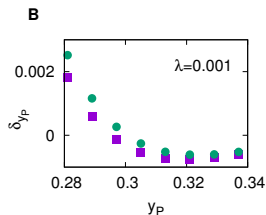
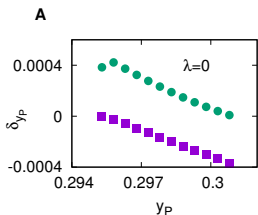
- ▶  $\Delta$  shows a peak at a  $\lambda$ , which matches with  $\lambda^*$  for  $\langle C \rangle$
- ▶  $\tau$  decreases monotonically with noise, as in a homogeneous nutrient environment

## Peak for $V$ shifts to a higher noise value

- ▶ At the peak position  $\lambda_o$  one must have  $\tau\Delta' - \Delta\tau' = 0$
- ▶ Since  $\tau' < 0$  for all  $\lambda$  values,  $\Delta' < 0$  at  $\lambda = \lambda_o$
- ▶ Peak of  $\Delta$  occurs at a smaller  $\lambda$  value and at  $\lambda_o$  it is decreasing with  $\lambda$
- ▶ How the non-monotonic variation of  $\Delta$  with noise can be explained from a detailed analysis of the biochemical pathway?
- ▶ Detailed study of CheY-P level statistics needed

# Explanation of optimal noise strength

- ▶ For very low methylation noise, the only source of fluctuations in activity, methylation or CheY-P level is the stochastic change in the cell position
- ▶ As the cell moves rightward, the ligand concentration increases, and the free energy  $f_{[L]}$  increases, causing the activity to decrease
- ▶ In a leftward run, activity increases
- ▶ CheY-P level goes down (up) in a rightward (leftward) run
- ▶ Measure the average change in CheY-P level in between two tumbles, when the intervening run is directed rightward (leftward).

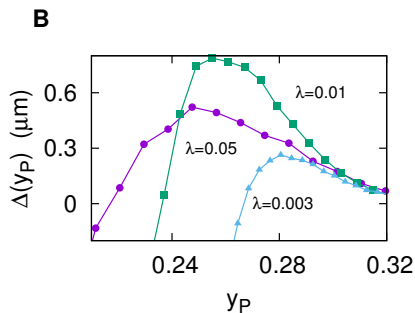
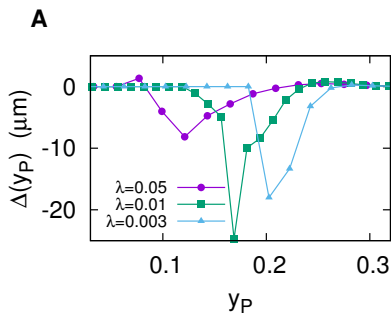




## Average displacement in a run that starts with $y_P$

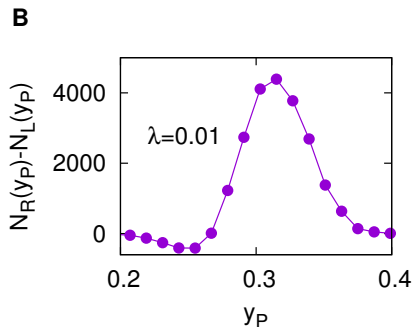
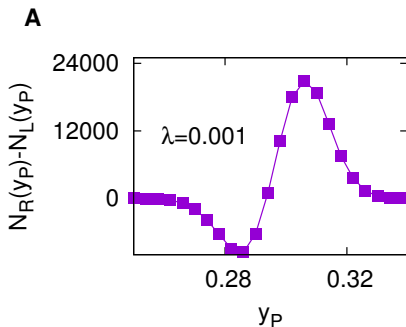
$$\Delta(y_P) = \frac{d_R(y_P) - d_L(y_P)}{N_R(y_P) + N_L(y_P)}$$

A negative peak at small  $y_P$ , followed by a positive peak at large  $y_P$



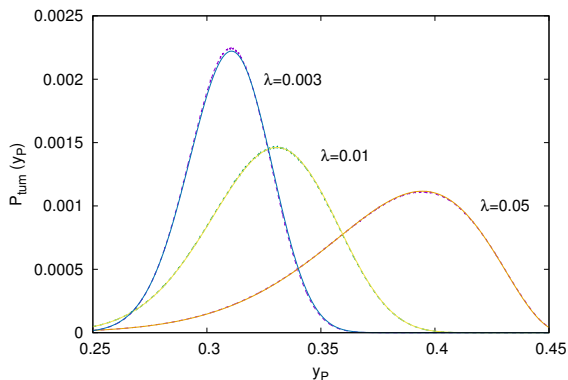
- ▶ For low noise, a rightward run starting with a given  $y_P$  must be preceded by a leftward run which terminates at the same  $y_P$
- ▶ This leftward run must have originated from a lower  $y_P$  value since for low noise,  $y_P$  value can only increase during a leftward run
- ▶ This event becomes particularly unlikely when  $y_P$  values are already small, near the left-tail of the distribution  $P_{tum}(y_P)$
- ▶ Therefore, for small  $y_P$  values,  $N_R(y_P) < N_L(y_P)$  and as a result,  $d_R(y_P) < d_L(y_P)$ , which makes  $\Delta(y_P)$  negative
- ▶ As  $y_P$  increases and comes out of the tail region,  $N_R(y_P)$  gradually increases and overtakes  $N_L(y_P)$ , and  $\Delta(y_P)$  becomes positive
- ▶ However, as  $y_P$  becomes very large, run durations become rather small and while  $N_R(y)$  remains above  $N_L(y)$ , their individual values start decreasing for large  $y_P$
- ▶  $\Delta(y_P)$  decreases again for large  $y_P$

$N_R(y_P) < N_L(y_P)$  for small  $y_P$  values



- ▶ For very small and large  $y_P$  both  $N_R(y_P)$  and  $N_L(y_P)$  vanish
- ▶ For intermediate  $y_P$  values,  $N_R(y_P) - N_L(y_P)$  shows a positive and negative peak

## CheY-P distribution for different noise



- ▶ While averaging  $\Delta(y_P)$  over the distribution  $P_{tum}(y_P)$ , small  $y_P$  values give negative contribution and reduces  $\Delta$
- ▶ Negative  $\Delta(y_P)$  values are near the left tail of  $P_{tum}(y_P)$  and hence occur with low probability
- ▶ Thus overall drift velocity still remains positive

## Threshold shifts with noise

- ▶ As noise increases, the distribution  $P_{tum}(y_P)$  becomes wider and the left tail becomes much longer than the right tail
- ▶ For very small  $y_P$  both rightward and leftward runs raise the  $y_P$  level, but leftward runs do so by a larger magnitude
- ▶ As a result, a leftward run that precedes a rightward run and that terminates at a small  $y_P$  must have to start from an even smaller  $y_P$ , which has a low probability associated with it
- ▶  $N_R(y_P) < N_L(y_P)$ , but at a much smaller  $y_P$  value, when  $\lambda$  is relatively large
- ▶ Zero-crossing of  $\Delta(y_P)$  and its positive peak are both shifted towards smaller  $y_P$

- ▶ Starting from a large value, as  $y_P$  is decreased,  $\Delta(y_P)$  keeps increasing and this trend continues till a much smaller  $y_P$  value, after which it finally starts declining again
- ▶ Averaging over such a curve yields a higher value of  $\Delta$  than what was observed for small noise
- ▶ Hence  $\Delta$  increases as noise increases
- ▶ But when noise becomes too large, the cell cannot distinguish between rightward and leftward runs
- ▶ The change in activity or  $y_P$  in a run is completely controlled by methylation level fluctuations now, and ligand concentration plays an insignificant role
- ▶ This again reduces  $\Delta$

## Noise induced sensitivity [Flores *et al.* PRL (2012)]

- ▶ In a shallow gradient, chemotactic drift velocity shows a peak with noise, while localization remains flat at low noise and decreases to zero as noise increases
- ▶ (a) Internal state of the signaling pathway is described just in terms of activity and both methylation level and CheY-P level are expressed as a function of activity
- ▶ (b) Sigmoidal nature of dependence of tumbling rate on activity was approximated by making the tumbling rate zero as the activity level falls below some value
- ▶ Drift motion results from the difference in the amount of time a right-mover and a left-mover spends in the small activity state
- ▶ With increasing noise these small activity states are reached more often and hence drift velocity also increases
- ▶ For large noise, the difference between right- and left-mover again decreases, causing the drift velocity to go down

## Detrimental response below a threshold

- ▶ In comparison, we find high activity level gives negligible contribution to noise
- ▶ When activity decreases, contribution increases
- ▶ When activity becomes lower than a certain threshold, contribution becomes negative
- ▶ The threshold value decreases as noise increases
- ▶ A crucial factor in explaining the noise induced enhancement of chemotactic drift velocity



## Effect of signaling noise with time-varying nutrient concentration

- ▶ In many physical situations the chemical environment experienced by the cell changes with time
- ▶ Diffusion or degradation of chemo-attractant
- ▶ Short time behavior of the cell more important here

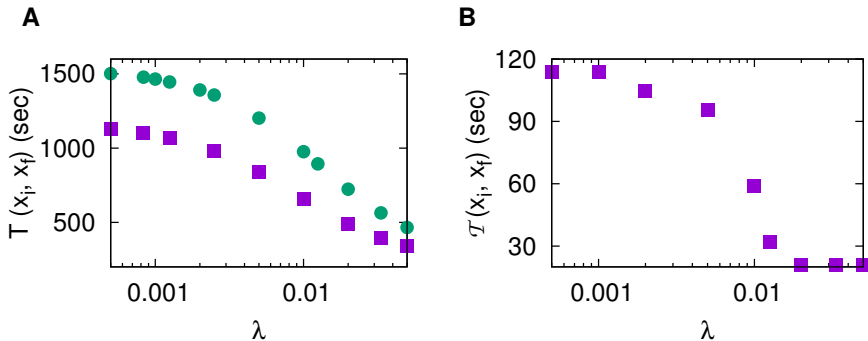
$$c(x, t) = c_0 e^{-t/\tau_d} \left[ 1 + \frac{\exp\left(-\frac{(x - \bar{x})^2}{\sigma_0^2 + 4Dt}\right)}{\sqrt{2\pi(\sigma_0^2 + 4Dt)}} \right]$$

- ▶ In a harsh chemical environment, the cell needs to find the favorable spot quickly and its trajectory should encounter large number of nutrient molecules
- ▶ First passage time of the cell measured at a region close to the peak of the Gaussian where the nutrient concentration is highest
- ▶ Uptake  $\mathcal{U} = \int_0^{t_{obs}} dt \int_0^L dx c(x, t) \mathcal{P}_\lambda(x, t)$
- ▶ Mean amount of nutrient encountered by the cell along its trajectory upto a large enough observation time

## Decaying nutrient profile

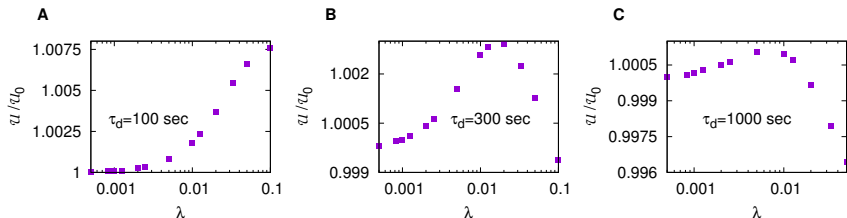
- ▶ Nutrient diffusivity  $\mathcal{D}$  very small
- ▶ Cell experiences a Gaussian concentration profile with almost fixed width  $\sigma_0$ , and an exponential decay of the overall concentration level
- ▶ Mean and typical first passage time decreases with  $\lambda$ , as in a homogeneous medium without degradation
- ▶ However, degradation increases the first passage time for a given  $\lambda$
- ▶ When nutrient degrades, even when the cell is moving in a homogeneous medium, it experiences a decreasing concentration along its trajectory, which makes it tumble more
- ▶ Average run durations shorter and hence the mean first passage time longer

# Mean and typical first passage time vs $\lambda$



- ▶ The circles show the data for  $c(x, t) = c_0 e^{-t/\tau_d}$  and squares are for degrading Gaussian profile with  $\mathcal{D} = 0$
- ▶  $\tau_d = 500\text{sec}$ ,  $\sigma_0 = 100\mu\text{m}$
- ▶ For small  $\lambda$ , typical first passage time is more useful

## Uptake variation with noise depends on $\tau_d$

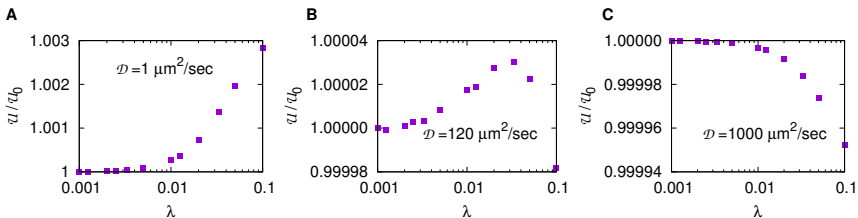


- ▶  $t_{obs} = 1000s$
- ▶ For small  $\tau_d$ , nutrient decays rapidly, only those trajectories with very long runs contribute
- ▶ As  $\lambda$  increases, such trajectories become more probable
- ▶ For large  $\tau_d$ , nutrient degrades slowly, time-independent limit recovered
- ▶ Uptake peak approaches localization peak  $\lambda^*$

# Nutrient profile with decay and diffusion

- ▶ Same qualitative behavior of first passage time
- ▶ Uptake behavior depends on the interplay between degradation and diffusion time-scales
- ▶ For large  $\mathcal{D}$  concentration gradient in the medium disappears fast and the uptake does not depend on cell trajectory anymore, except for very small times
- ▶ The more time the cell is able to spend close to the peak of the Gaussian profile before the profile flattens or nutrient degrades, larger will be its uptake
- ▶ For large  $\lambda$ , cells execute long runs which decreases the residence time near the peak: uptake is low

# Uptake vs noise with degradation and diffusion



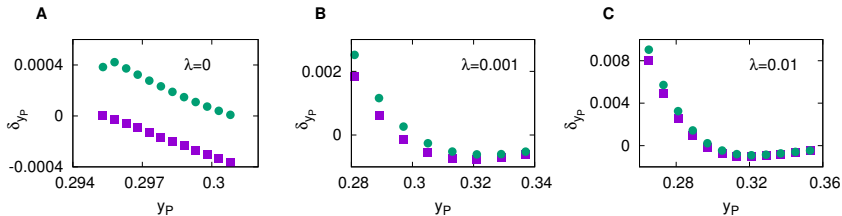
- ▶  $\tau_d = 100\text{sec}$ ,  $\sigma = 10\mu\text{m}$ ,  $t_{obs} = 200\text{sec}$
- ▶ For small  $\lambda$ , FPT is larger
- ▶ Cell spends most of its short time trajectory trying to climb up the concentration gradient, reaching the peak of the Gaussian
- ▶ Uptake is larger
- ▶ A peak for intermediate  $D$  value

# Conclusions

- ▶ Effect of methylation noise on the chemotactic performance of a single *E. coli* cell
- ▶ An optimum noise strength for the best performance
- ▶ Explanation from CheY-P level fluctuations for cell motion up and down the gradient
- ▶ Detrimental behavior below noise-dependent threshold
- ▶ Adaptation of flagellar motors or spatial organization of chemo-receptors neglected
- ▶ Fluctuations can originate from clustering of chemo-receptors [Colin *et al.* (2017)]



## Change in CheY-P level during a run



- ▶ as  $\lambda$  increases, the change in activity is also controlled by the methylation level fluctuations and the feedback it produces on the reaction network
- ▶ When activity becomes too low (high), the methylation level increases, which in turn causes the activity to increase (decrease)
- ▶ For large  $\lambda$ , when the activity varies over a wider range, the feedback effect is more prominent and can easily override the change in activity due to change in cell position

- ▶ When  $\lambda$  becomes high, activity can also increase during a rightward run, especially when its value at the start of the run is sufficiently small
- ▶ Similarly, in a leftward run activity may decrease when its value is high enough
- ▶ In terms of CheY-P level, this means that during a rightward (leftward) run the CheY-P level  $y_P$  can increase (decrease) when  $y_P$  has small (large) values