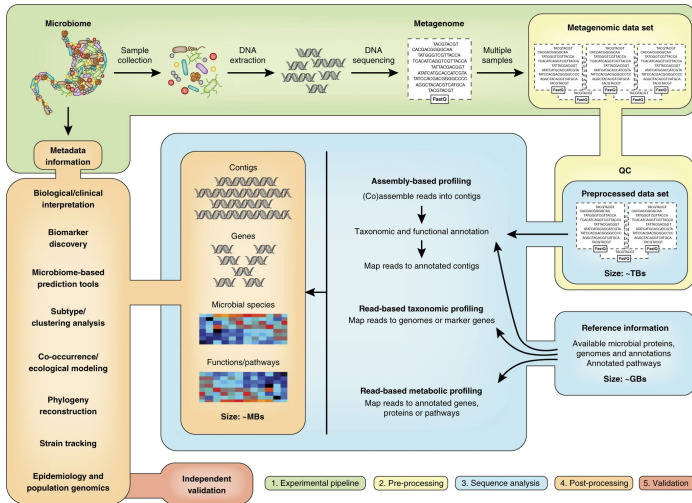


Shotgun metagenomics sequencing

Shotgun metagenomics sequencing

- ▶ Untargeted ('shotgun') sequencing of all ('meta-') microbial genomes 'genomics' present in a sample
- ▶ Could profile taxonomic composition and functional potential, and to recover whole genome sequences
- ▶ Other advanced use of metagenomics: Quantify bacterial growth rate.

Shotgun metagenomics sequencing



Shotgun metagenomics sequencing

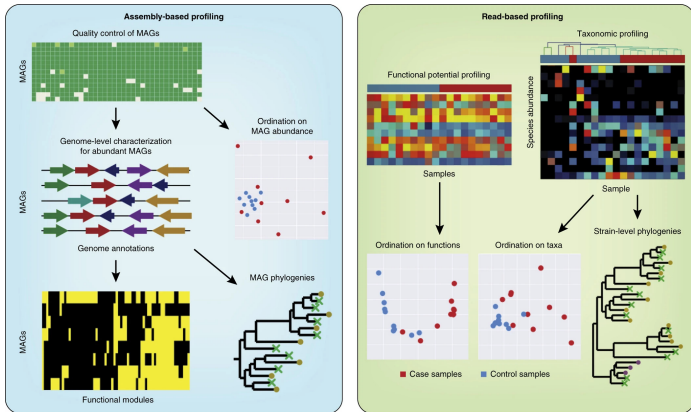


Figure: Integrative analysis leveraging information from multiple studies

Infer bacterial growth rate from
shotgun metagenomics sequencing

MICROBIOME

Growth dynamics of gut microbiota in health and disease inferred from single metagenomic samples

Tal Korem,^{1,2*} David Zeevi,^{1,2*} Jotham Suez,^{2*} Adina Weinberger,^{1,2*} Tali Avnit-Sagi,^{1,2} Maya Pompan-Lotan,^{1,2} Elad Matot,^{1,2} Ghil Jona,² Alon Harmelin,² Nadav Cohen,^{1,2} Alexandra Sirota-Madi,² Christoph A. Thaiss,² Meirav Pevsner-Fischer,² Rotem Sorek,² Rannik J. Xavier,² Eran Elinav,² Eran Segal^{1,2,†}

PTRs extracted from these samples varied across individuals, in the range of 1 to 2.4, resembling the 1 to 2.6 range of ratios measured *in vitro* (Fig. 1B). Ratios higher than 2 are indicative of multifork replication, previously documented for *E. coli* (18, 22).

To examine whether PTRs provide a quantitative measure of growth rate, we calculated the temporal growth rate of *E. coli* at different times during its growth experiment as the derivative of its abundance across time (23). PTRs were correlated with the measured growth rate, preceding it by 30 min ($R = 0.95$, $P < 10^{-4}$) (Fig. 1D), indicating that PTR predicts the change in abundance. To determine whether PTRs accurately

nature | methods

BRIEF COMMUNICATION

<https://doi.org/10.1038/s41592-018-0182-0>

Quantifying and comparing bacterial growth dynamics in multiple metagenomic samples

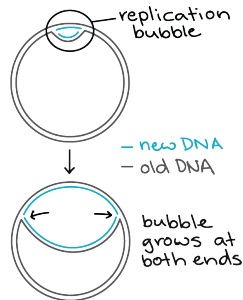
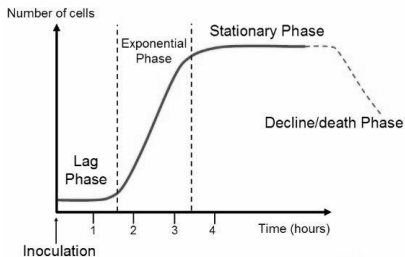
Yuan Gao and Hongzhe Li[✉]*

GENOME RESEARCH

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Accurate and robust inference of microbial growth dynamics from metagenomic sequencing reveals personalized growth

Bacteria growth

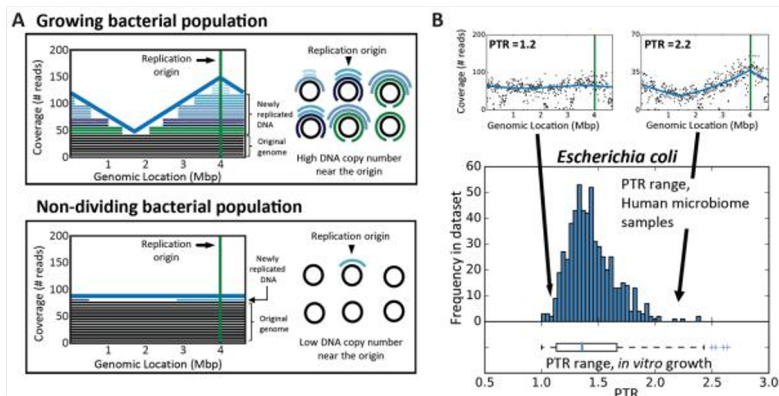


<https://orbitbiotech.com/bacterial-growth-curve-generation-time-lag-phase-log-phase-exponential-phase-decline-phase/>

FAVPNG

How do we use metagenomics sequencing to infer bacterial growth rate.

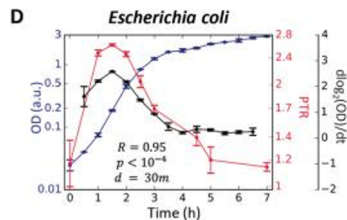
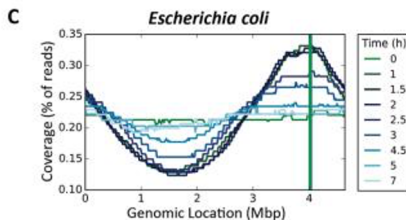
Peak-to-trough coverage ratio (PTR)



PTR accurately measures in vitro growth rates of *Escherichia coli*

Korem et al., *Science*, 2015

PTRs provide a quantitative measure of growth rate

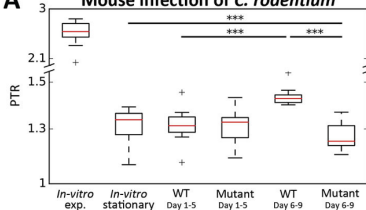


$$PTR = 2^{C/G}; 1/\log 2(PTR) \propto G/C$$

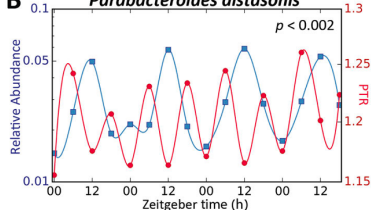
- C: replication time
- G: generation time

PTR reflects the growth dynamics of *in vivo* microbial communities

A Mouse infection of *C. rodentium*



B *Parabacteroides distasonis*



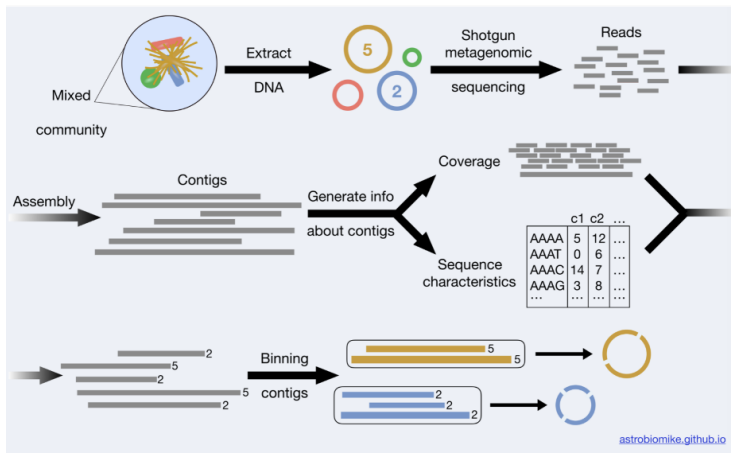
Korem et al., *Science*, 2015

Analysis pipeline

- ▶ Reads were mapped to full bacterial genomes. (requires complete reference genomes)
- ▶ Total sequencing reads were summed into non-overlapping 10Kbp bins.
- ▶ Predict replication origin: Fit $\log_2(C)$ (coverage) into a piecewise linear model. Within each sample, the highest and lowest are the origin and terminus of replication.
- ▶ $PTR = \frac{PredC_{peak}}{PredC_{trough}}$: $PredC$ predicted coverage.

When we do not have complete reference genome?

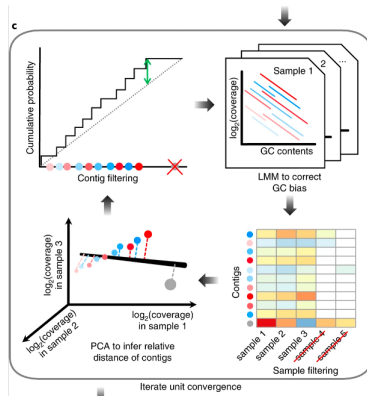
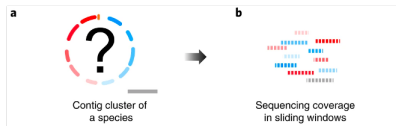
Assemble genomes from metagenomic samples! MAG (metagenomics assembly genome)



Challenges

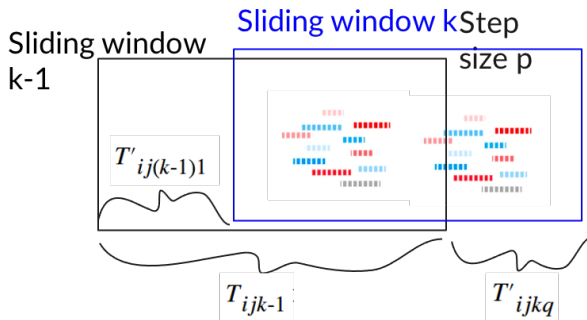
- ▶ For species with only genome assemblies, the accurate locations of the assemblies on the original genome are unknown
- ▶ The contigs assembled from metagenomic data are highly fragmented
- ▶ Binning algorithms could fail to cluster all contigs from the same species into one group (incomplete contig clusters), or mistakenly include a fraction of contigs from other species (contig contamination).

Overview of DEMIC



Gao & Li, *Nature Method*, 2018

Calculation of contig coverages for sliding windows



Average coverage of the k -th window for the j -th contig in the i -th sample:

$$Y_{ijk} = \frac{1}{l'} (T_{ijk-1} + T'_{ijkq} - T'_{ij(k-1)1})$$

Correcting for sequencing bias using LMM

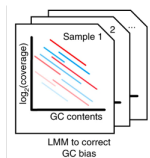
For the j -th contig of the i -th sample:

$$\log_2 Y_{ijk} = a_0 + (X_{jk} - \bar{X})a + Z_{ij} + e_{ijk}$$

Average
sequencing
coverage for the
 k -th window

GC content
for the k -th
window

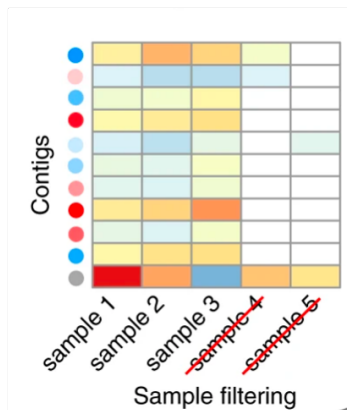
Sample- and contig-
specific random
intercept



GC-adjusted log-transformed coverage of sample i and contig j :

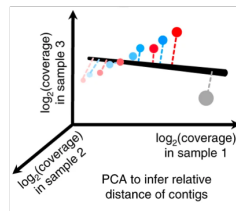
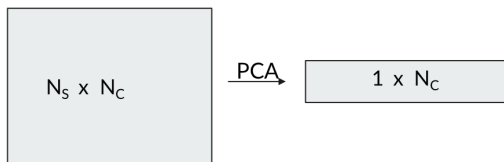
$$Y'_{ij} = \hat{a}_0 + \hat{Z}_{ij}$$

Sample filtering



- ▶ Exclude samples with low coverage of the given species (no count for more than half of the contigs within the cluster).

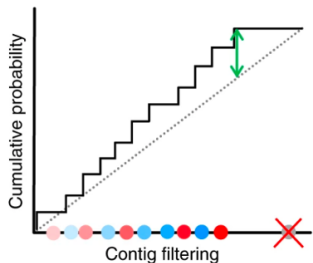
PCA to infer relative distance of contigs



N_s : number of samples; N_c : number of contigs.

The (i, j) -th entry contains GC-adjusted log-transformed coverage of sample i and contig j .

Contig filtering



- ▶ Compare the distribution of PC1 values of all contigs against the uniform distribution, $\text{Unif}(\min(U), \max(U))$.
- ▶ If there is a significant difference in distribution, remove one of the two contigs with maximum and minimum PC1 values.
 - ▶ Compare the two contigs with respect to their distance from the adjacent contig.
 - ▶ Discard the one with the larger distance

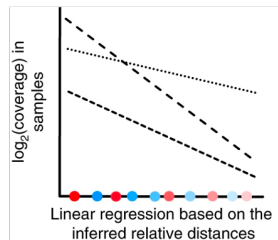
Linear regression based on inferred relative distances

For the i -th sample:

$$Y'_{ij} = b_{0i} + b_i U_j, \quad j=1,2,\dots,N_c$$

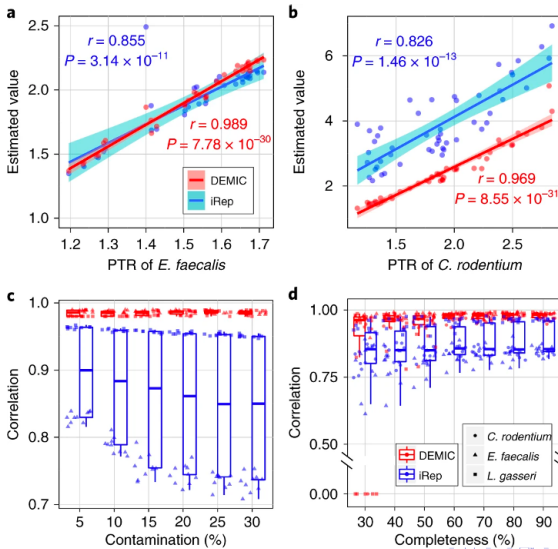
GC-adjusted
log-
transformed
coverage of
contig j

PC1 value of
contig j

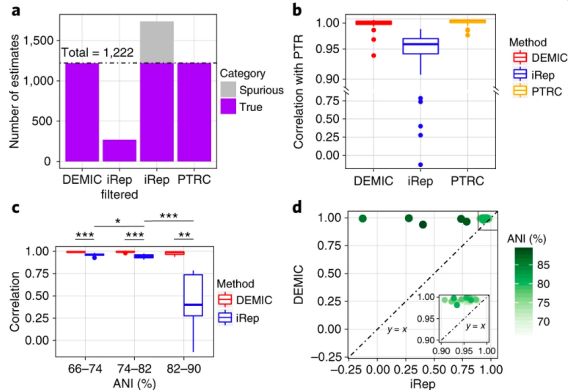


$$\text{Estimated PTR } ePTR_i = \frac{\exp(\hat{b}_{0i} + \hat{b}_i U_{(N_c)})}{\exp(\hat{b}_{0i} + \hat{b}_i U_{(N_1)})}, \quad i = 1, 2, \dots, N_s$$

Simulated results



Simulated results



Real data analysis

Metagenomic data sets of fecal samples from 26 healthy children and 86 children with Crohn's diseases.

