

# Tell-TaxContigs: Microbial Metagenome Assembly, Taxonomy, and Abundance Estimation Using UST TELL-seq Long-range Information

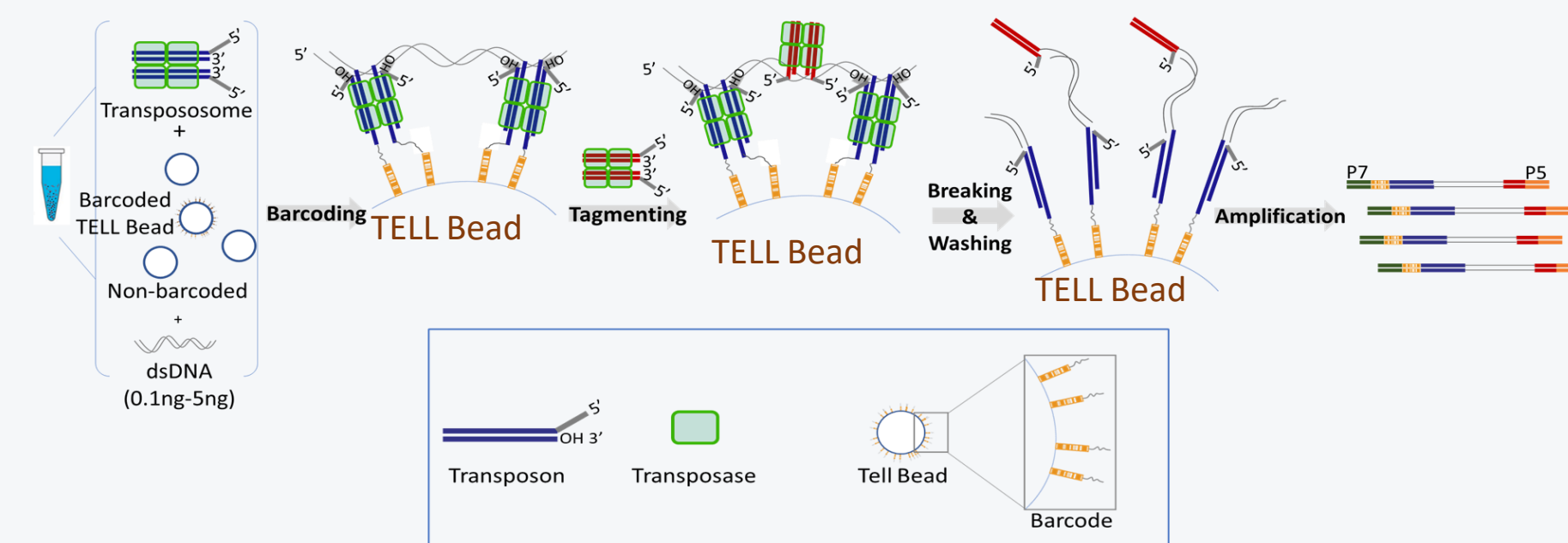
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**Abstract.** Deconvolving diversity of a metagenome is critical for understanding the role of a given microbial community in human health and disease, small molecule biosynthesis, and other complex ecosystems where more reductive analysis proves to be elusive. Metagenomic assembly is one common method for characterizing a metagenome, especially for identifying novel gene content or novel organisms. However, analyzing sequencing data from microbial mixtures with a high dynamic range of relative abundance of strains, close relatedness, and repetitive genomic content amongst members can vastly complicate the genome assembly process of individual microorganisms and strains. Furthermore, efficient assembly requires high fidelity sequencing reads to avoid ambiguities. We previously developed a method that captures long-range molecular origin information from kilobase-long genomic fragments by a process of DNA barcoding that we called transposase enzyme-linked long read sequencing (TELL-seq) developed by Universal Sequencing Technologies (UST)<sup>1</sup>. TELL-seq barcoded fragments can be sequenced with instruments that process short reads (i.e., high-fidelity sequencing). Here, we show that integration of TELL-seq data with a computational pipeline that combines *de novo* genome assembly (Tell-Link) with taxonomic classification and abundance estimation (Tell-TaxContigs) provides highly accurate metagenomic analyses. We show how the application of Tell-Link and Tell-TaxContigs on sequencing data generated from commercially available microbial mixture standards results in genome assemblies with contiguities larger than 1Mbp (N50), and highly accurate classification and relative abundance estimation for organisms at 0.18% or greater relative abundance, respectively. Therefore, Tell-Link, in combination with genome binning software (e.g. metabat2<sup>2</sup>) provides highly contiguous and high-fidelity genome assemblies of abundant organisms in a metagenomic sample. Tell-TaxContigs classifies contigs and unassembled reads with BLASTn and using a deep learning approach resolves ambiguities, rules out false positive classifications, and accurately estimates relative abundances of classified species. This approach has an average margin of error of lower than 1% in enumerating relative abundance for the microbial mixture standards tested.

## How Does UST TELL-seq Work?

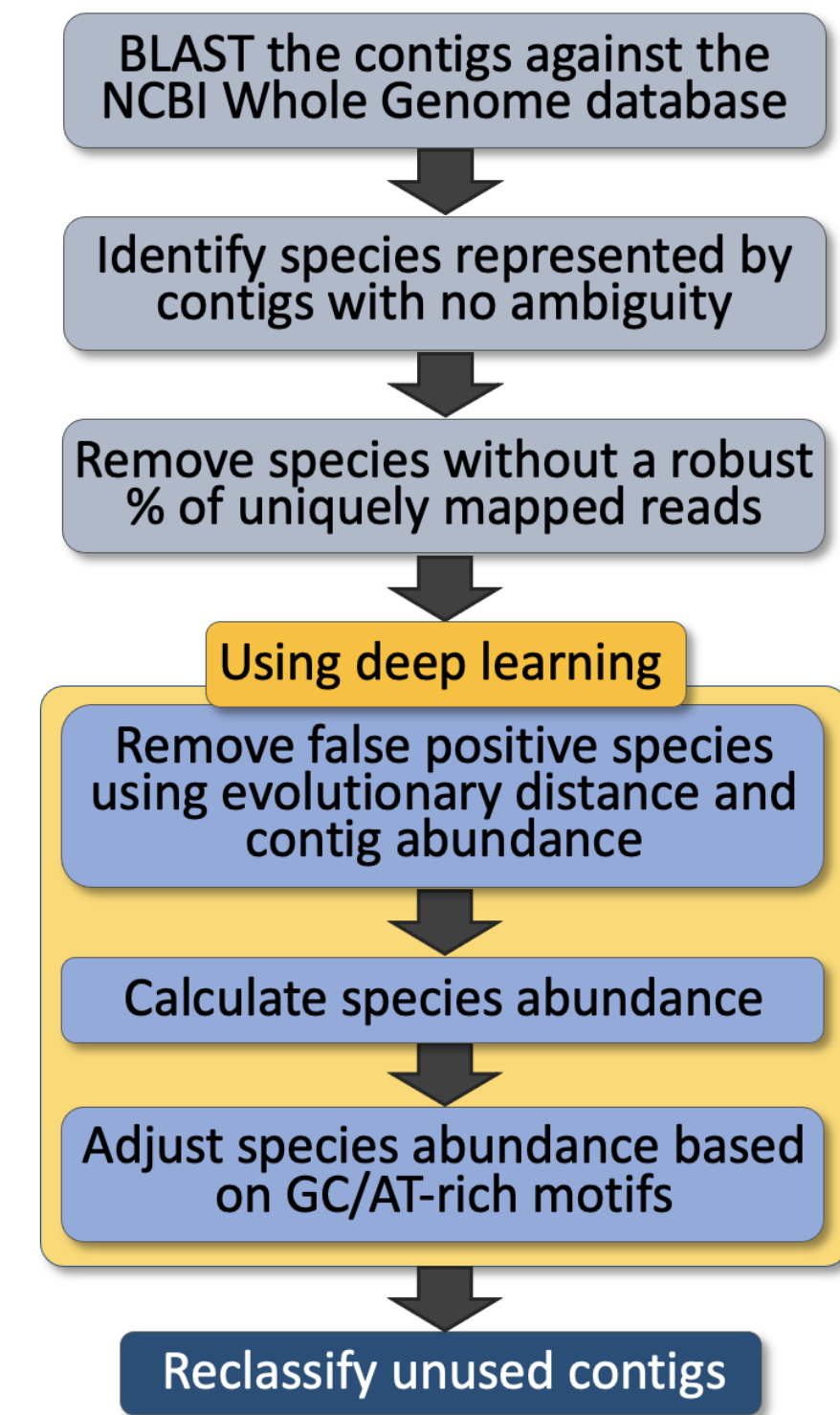
### Transposase Enzyme Linked Long-Read Sequencing



### TELL-seq Library Workflow is Simple & Scalable



## Tell-TaxContigs Overview



### Metagenomic assembly summary

Microbial Mix	Even	Staggered	Staggered-deep
Sequencing depth	59M	98M	700M
True positives (out of 20)	16	7	8
Unclassified	1	0	3
Quality draft	15	5	7
Avg. contiguity	1.52	2.73	1.64

### Notes on metagenomic assembly:

Contiguity is defined as assembly length divided by N50 score.  
Quality draft denotes medium quality draft assembly according to Bowers et al. (2017)<sup>3</sup> minimum information of MAGs.

## 1. Metagenomic Assembly & Genome Binning

### Detailed metagenomic assembly and genome binning results

Sample (color-coded)	Even / Staggered / Staggered-deep																	
Species Name	Contigs			Length (Mbp)			N50 (Mbp)			Length/N50			Completeness		Contamination		Quality draft	
Staphylococcus epidermidis	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Streptococcus mutans	NA	6	NA	NA	1.87	NA	NA	1.33	NA	NA	1.40	NA	NA	92.77	NA	NA	0.00	NA
Porphyromonas gingivalis	10	15	10	2.12	2.07	2.13	1.23	2.02	1.17	1.72	1.02	1.82	98.82	98.79	99.29	0.00	0.00	0.00
Escherichia coli	3	11	4	4.51	4.41	4.52	4.50	4.38	4.50	1.00	1.01	1.00	99.87	98.32	99.97	0.04	0.04	0.04
Rhodobacter sphaeroides	23	11	5	4.37	4.39	4.41	3.03	3.11	3.10	1.44	1.41	1.42	94.11	97.36	99.24	0.15	0.15	0.15
Staphylococcus aureus	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Streptococcus agalactiae	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Bacillus cereus	7	173	NA	5.33	1.19	NA	2.72	2.07	NA	1.96	5.75	NA	97.61	95.57	NA	0.33	193.23	NA
Clostridium beijerinckii	42	110	NA	5.92	5.68	NA	2.64	1.37	NA	2.24	4.16	NA	99.19	96.77	NA	2.42	2.42	NA
Pseudomonas aeruginosa	6	262	3	6.29	5.53	6.29	6.27	1.26	6.28	1.00	4.39	1.00	98.40	86.96	99.68	0.45	0.92	0.45
Lactobacillus gasseri	2	NA	NA	1.80	NA	NA	1.79	NA	NA	1.00	NA	NA	98.36	NA	NA	0.00	NA	NA
Helicobacter pylori	1	NA	14	1.59	NA	1.59	1.59	NA	1.55	1.00	NA	1.03	99.09	NA	98.20	0.00	NA	0.00
Acinetobacter baumannii	15	NA	5	7.93	NA	3.83	2.05	NA	3.81	3.86	NA	1.00	100.00	NA	99.45	175.51	NA	0.27
Neisseria meningitidis	8	NA	48	2.02	NA	1.84	1.99	NA	1.69	1.01	NA	1.09	98.23	NA	92.14	0.19	NA	0.19
Cutibacterium acnes	1	NA	75	2.48	NA	2.25	2.48	NA	2.01	1.00	NA	1.12	98.90	NA	79.07	0.00	NA	0.00
Enterococcus faecalis	3	NA	NA	2.68	NA	NA	2.67	NA	NA	1.00	NA	NA	98.89	NA	NA	0.00	NA	NA
Bacteroides vulgatus	7	NA	NA	4.82	NA	NA	3.12	NA	NA	1.55	NA	NA	98.45	NA	NA	0.19	NA	NA
Deinococcus radiodurans	5	NA	NA	3.03	NA	NA	2.61	NA	NA	1.16	NA	NA	99.04	NA	NA	0.21	NA	NA
Actinomyces odontolyticus	2	NA	NA	2.36	NA	NA	2.36	NA	NA	1.00	NA	NA	97.17	NA	NA	0.47	NA	NA
Bifidobacterium adolescentis	1	NA	NA	1.99	NA	NA	1.99	NA	NA	1.00	NA	NA	97.15	NA	NA	0.00	NA	NA

## 2. Taxonomic Classification & Abundance Estimation

### Detailed classification results

Bacterial strains	True Abundance		Tell-TaxContigs Estimated Abundance		
	Even	Staggered	Even	Staggered	Staggered-deep
Staphylococcus epidermidis	5%	18.00%	5.13%	17.28%	15.63%
Streptococcus mutans	5%	18.00%	6.23%	14.45%	14.68%
Porphyromonas gingivalis	5%	18.00%	5.44%	16.26%	15.82%
Escherichia coli	5%	18.00%	4.47%	20.67%	21.79%
Rhodobacter sphaeroides	5%	18.00%	4.91%	18.18%	16.12%
Staphylococcus aureus	5%	1.80%	5.62%	1.57%	1.49%
Streptococcus agalactiae	5%	1.80%	5.68%	1.56%	2.14%
Bacillus cereus	5%	1.80%	4.33%	2.17%	2.18%
Clostridium beijerinckii	5%	1.80%	5.67%	1.57%	1.33%
Pseudomonas aeruginosa	5%	1.80%	5.49%	1.61%	0.91%
Lactobacillus gasseri	5%	0.18%	4.15%	0.23%	0.11%
Helicobacter pylori	5%	0.18%	4.08%	0.24%	0.10%
Acinetobacter baumannii	5%	0.18%	4.46%	0.21%	0.15%
Neisseria meningitidis	5%	0.18%	3.09%	0.65%	0.10%
Cutibacterium acnes	5%	0.18%	3.67%	0.31%	0.10%
Enterococcus faecalis	5%	0.02%	6.08%	NA	0.02%
Bacteroides vulgatus	5%	0.02%	6.79%	NA	0.05%
Deinococcus radiodurans	5%	0.02%	4.55%	NA	0.04%
Actinomyces odontolyticus	5%	0.02%	3.68%	NA	NA
Bifidobacterium adolescentis	5%	0.02%	5.87%	NA	NA

### Classification summary

Microbial Mix	Even	Staggered	Staggered-deep
Total sequencing reads	59M	98M	700M
True positives (out of 20)	20	15	18
False positives	0	0	2
Avg. RA error	0.83%	0.72%	0.91%
Max. RA error	1.91%	3.55%	3.79%

## References

- Chen, Z et al. 2020. *Genome Research* 30: 898-909.
- Dongwan et al. 2019. *PeerJ*, 7:e7359
- Bowers et al. 2017. *Nat. Biotechnol.* 35:725-731
- Wick et al. 2017. *PLoS Comput Biol* 13(6): e1005595.