

Quantum Many-Body Systems

from high-energy nuclear physics to quantum computing

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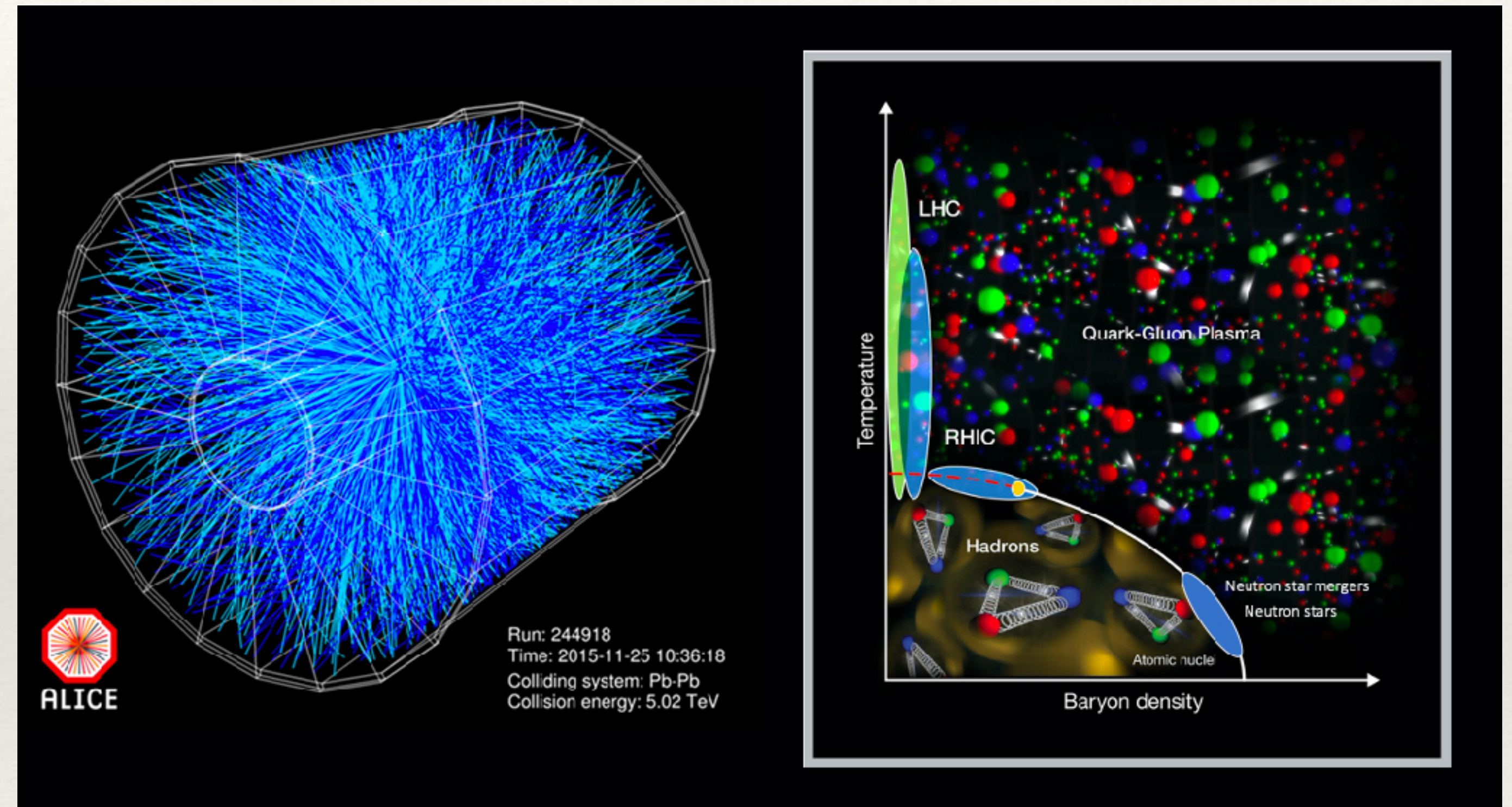
Quantum many-body systems

- ❖ Quantum many-body systems are very challenging, e.g. quark-gluon plasma, high- T_c superconductors
 - ❖ Quantum effects: wave-particle duality, uncertainty principle, spins (bosons, fermions), etc
 - ❖ Statistical effects: screenings, collective excitations, quasiparticles, etc
 - ❖ Emergent phenomena: phenomena that bear little resemblance of the underlying elementary laws
- ❖ The focus of this talk
 - ❖ High-energy nuclear physics: quark-gluon plasma
 - ❖ Quantum computing: many-qubit system

High-energy nuclear physics

Heavy-ion collisions and quark-gluon plasma

- ❖ Quark-gluon plasma (QGP)
 - ❖ Theory: weakly coupled gas
 - ❖ Experiment: strongly coupled liquid(!)
- ❖ Yang-Mills mass gap/color confinement
 - ❖ Clay Millennium Problems
- ❖ How to tackle it with thermal quantum field theory?
- ❖ What can deep learning do?



Hard-thermal-loop perturbation theory (HTLpt)

Andersen, Braaten & Strickland, 99

- ❖ Quantum Chromodynamics (QCD): massless quarks and gluons — inappropriate d.o.f. at high T

$$\mathcal{L}_{\text{QCD}} = -\frac{1}{2}\text{Tr}[G_{\mu\nu}G^{\mu\nu}] + i\bar{\psi}\gamma^\mu D_\mu\psi$$

- ❖ HTLpt: a resummation method of thermal QCD [see NS, Int. J. Mod. Phys. A, 30, 1530025 (2015) for a review]

$$\mathcal{L}_{\text{HTLpt}} = (\mathcal{L}_{\text{QCD}} + \mathcal{L}_{\text{HTL}} - \delta\mathcal{L}_{\text{HTL}}) \Big|_{g \rightarrow \sqrt{\delta}g}$$

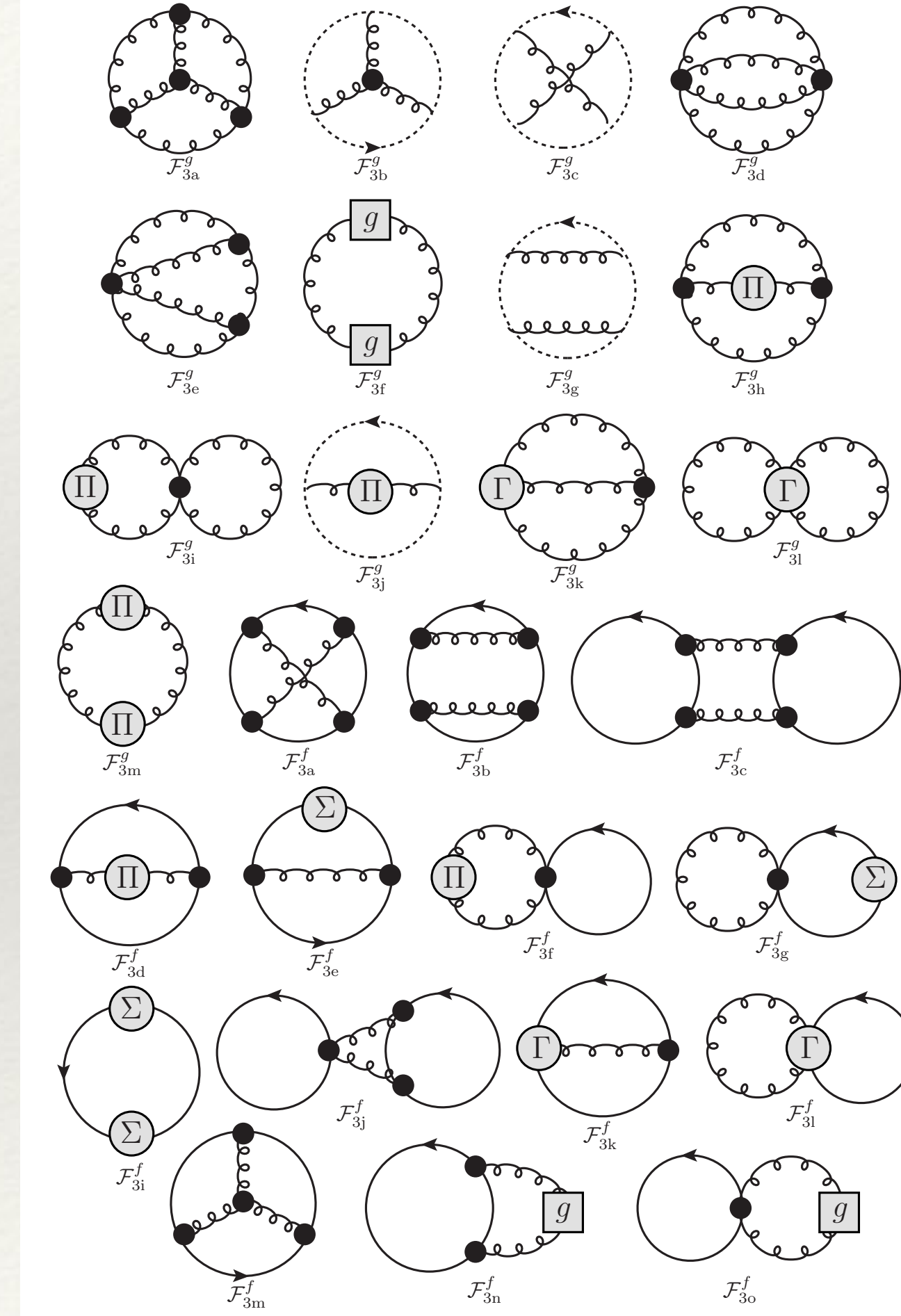
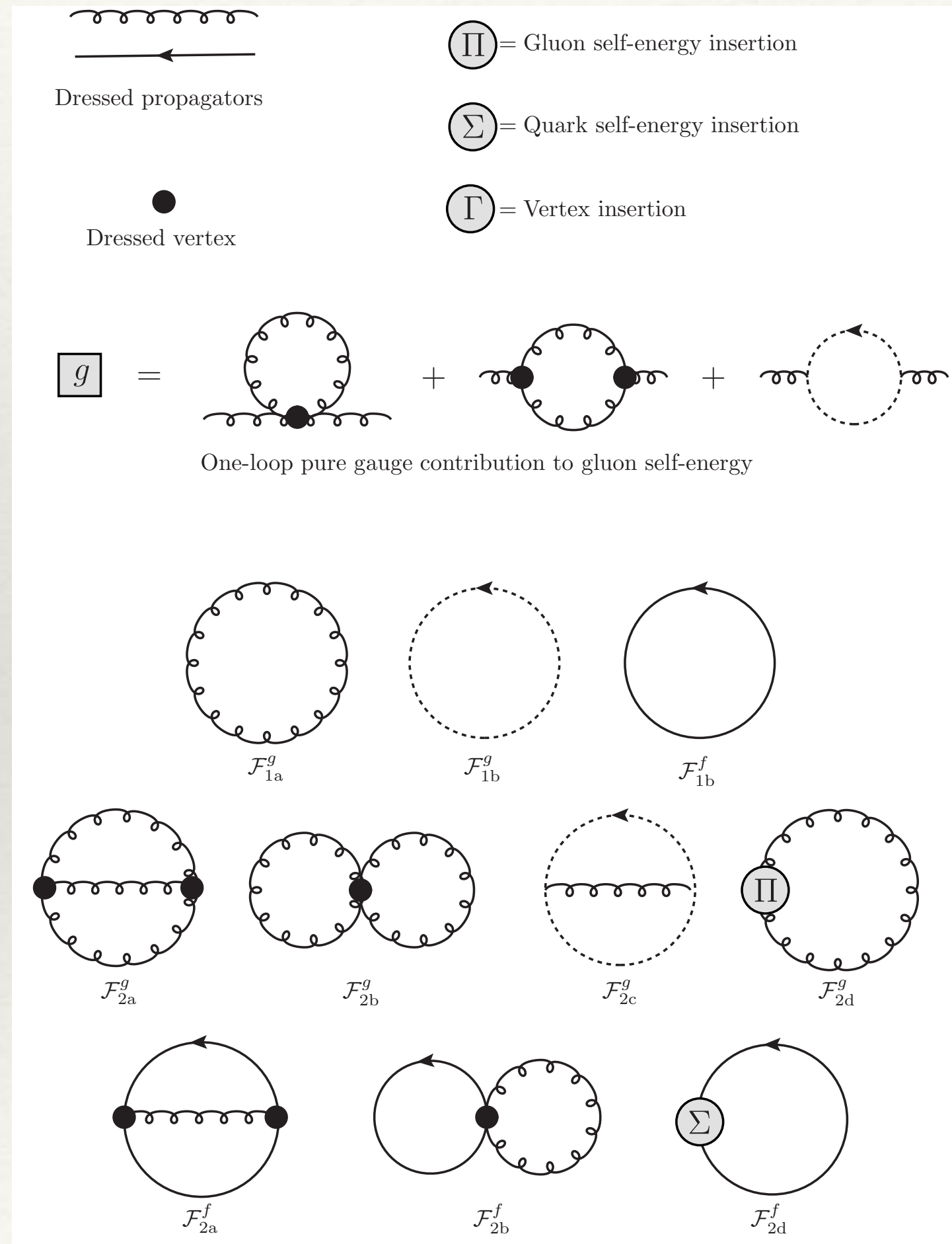
- ❖ Hard-thermal-loop effective action: contributions from the (chromo)electric scale gT [Braaten & Pisarski, 92]

$$\mathcal{L}_{\text{HTL}} = -\frac{1}{2}m_D^2\text{Tr}\left(G_{\mu\alpha}\left\langle\frac{y^\alpha y^\beta}{(y\cdot D)^2}\right\rangle_y G^\mu{}_\beta\right)$$

- ❖ Adding $\mathcal{L}_{\text{HTL}}$ **SHIFTS** expansion to an ideal gas of **massive (Debye) quasiparticles** — **appropriate** d.o.f. at high T
- ❖ EQCD from dimensional reduction — another resummation method for the electric scale gT [Braaten & Nieto, 95]

HTLpt: QCD thermodynamic potential Ω through NNLO/3 loops (46 diagrams)

Andersen, Leganger, Strickland, NS, JHEP 08, 053 (2011)



HTLpt: Ω_{NNLO} at finite T & μ for general N_c & N_f — analytic!

Haque, Bandyopadhyay, Andersen, Mustafa, Strickland and NS, JHEP 05, 027 (2014)

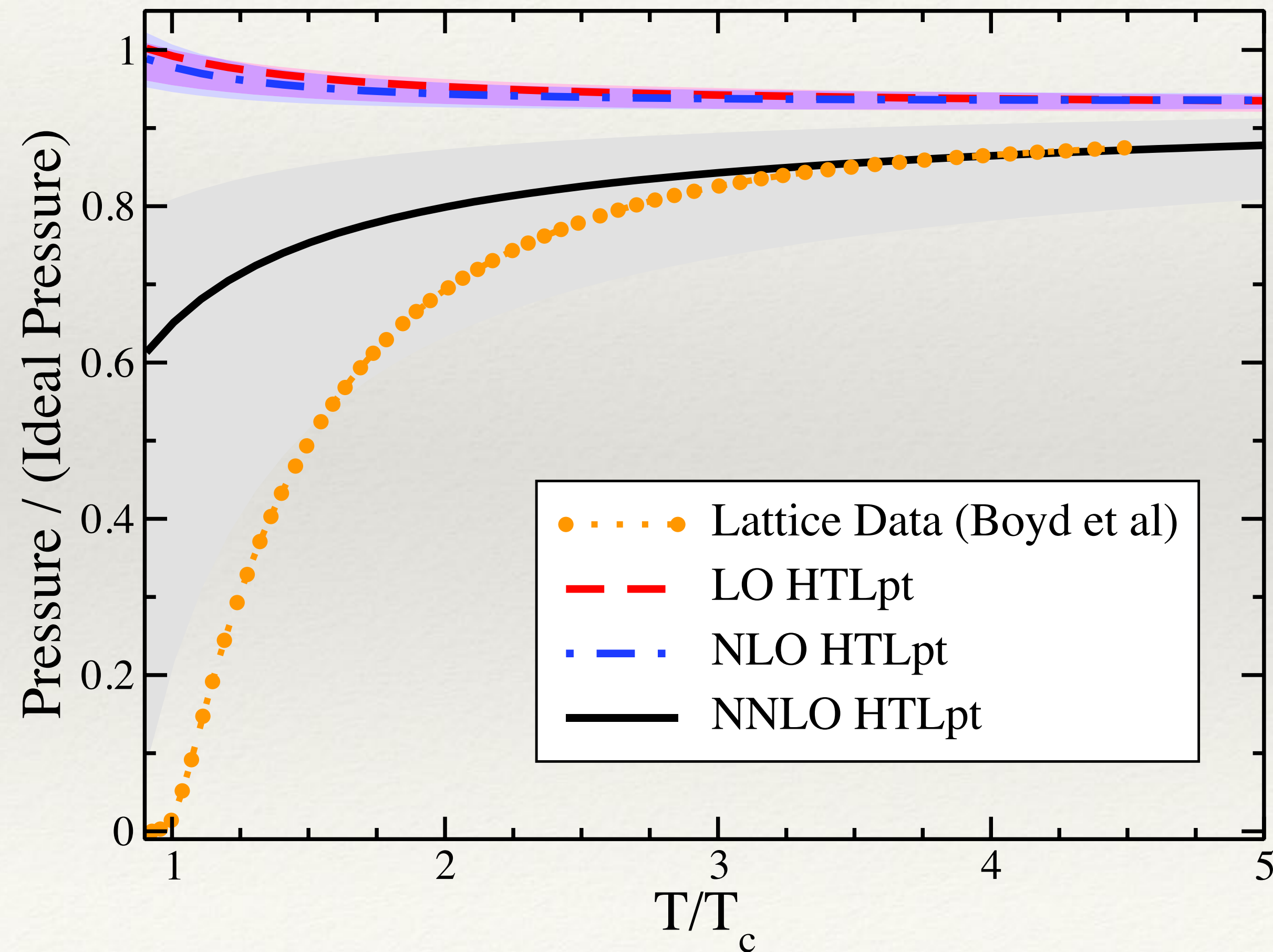
$\Omega_0 \equiv -d_A \pi^2 T^4 / 45$, $\hat{x} \equiv x / 2\pi T$, $\Omega_{\text{NNLO}}^{\text{YM}}$ taken from Andersen, Strickland, NS, PRL 104, 122003 (2010)

$$\begin{aligned}
 \frac{\Omega_{\text{NNLO}}}{\Omega_0} = & \frac{7}{4} \frac{d_F}{d_A} \frac{1}{N_f} \sum_f \left(1 + \frac{120}{7} \hat{\mu}_f^2 + \frac{240}{7} \hat{\mu}_f^4 \right) - \frac{s_F \alpha_s}{\pi} \frac{1}{N_f} \sum_f \left[\frac{5}{8} (1 + 12 \hat{\mu}_f^2) (5 + 12 \hat{\mu}_f^2) - \frac{15}{2} (1 + 12 \hat{\mu}_f^2) \hat{m}_D \right. \\
 & - \frac{15}{2} \left(2 \ln \frac{\hat{\Lambda}}{2} - 1 - \mathfrak{N}(z_f) \right) \hat{m}_D^3 + 90 \hat{m}_q^2 \hat{m}_D \left. \right] + \frac{s_{2F}}{N_f} \left(\frac{\alpha_s}{\pi} \right)^2 \sum_f \left[\frac{15}{64} \left\{ 35 - 32 (1 - 12 \hat{\mu}_f^2) \frac{\zeta'(-1)}{\zeta(-1)} + 472 \hat{\mu}_f^2 \right. \right. \\
 & + 1328 \hat{\mu}_f^4 + 64 \left(-36 i \hat{\mu}_f \mathfrak{N}(2, z_f) + 6(1 + 8 \hat{\mu}_f^2) \mathfrak{N}(1, z_f) + 3 i \hat{\mu}_f (1 + 4 \hat{\mu}_f^2) \mathfrak{N}(0, z_f) \right) \left. \right\} - \frac{45}{2} \hat{m}_D (1 + 12 \hat{\mu}_f^2) \left. \right] \\
 & + \left(\frac{s_F \alpha_s}{\pi} \right)^2 \frac{1}{N_f} \sum_f \frac{5}{16} \left[96 (1 + 12 \hat{\mu}_f^2) \frac{\hat{m}_q^2}{\hat{m}_D} + \frac{4}{3} (1 + 12 \hat{\mu}_f^2) (5 + 12 \hat{\mu}_f^2) \ln \frac{\hat{\Lambda}}{2} + \frac{1}{3} + 4 \gamma_E + 8(7 + 12 \gamma_E) \hat{\mu}_f^2 \right. \\
 & + 112 \hat{\mu}_f^4 - \frac{64}{15} \frac{\zeta'(-3)}{\zeta(-3)} - \frac{32}{3} (1 + 12 \hat{\mu}_f^2) \frac{\zeta'(-1)}{\zeta(-1)} - 96 \left\{ 8 \mathfrak{N}(3, z_f) + 12 i \hat{\mu}_f \mathfrak{N}(2, z_f) - 2(1 + 2 \hat{\mu}_f^2) \mathfrak{N}(1, z_f) - i \hat{\mu}_f \mathfrak{N}(0, z_f) \right\} \left. \right] \\
 & + \left(\frac{s_F \alpha_s}{\pi} \right)^2 \frac{1}{N_f^2} \sum_{f,g} \left[\frac{5}{4 \hat{m}_D} (1 + 12 \hat{\mu}_f^2) (1 + 12 \hat{\mu}_g^2) + 90 \left\{ 2 (1 + \gamma_E) \hat{\mu}_f^2 \hat{\mu}_g^2 - \left\{ \mathfrak{N}(3, z_f + z_g) + \mathfrak{N}(3, z_f + z_g^*) \right. \right. \right. \\
 & + 4 i \hat{\mu}_f \left[\mathfrak{N}(2, z_f + z_g) + \mathfrak{N}(2, z_f + z_g^*) \right] - 4 \hat{\mu}_g^2 \mathfrak{N}(1, z_f) - (\hat{\mu}_f + \hat{\mu}_g)^2 \mathfrak{N}(1, z_f + z_g) - (\hat{\mu}_f - \hat{\mu}_g)^2 \mathfrak{N}(1, z_f + z_g^*) \\
 & \left. \left. - 4 i \hat{\mu}_f \hat{\mu}_g^2 \mathfrak{N}(0, z_f) \right\} \right\} - \frac{15}{2} (1 + 12 \hat{\mu}_f^2) (2 - 1 - \mathfrak{N}(z_g)) \hat{m}_D \left. \right] + \left(\frac{c_A \alpha_s}{3\pi} \right) \left(\frac{s_F \alpha_s}{\pi N_f} \right) \sum_f \left[\frac{15}{2 \hat{m}_D} (1 + 12 \hat{\mu}_f^2) \right. \\
 & - \frac{235}{16} \left\{ \left(1 + \frac{792}{47} \hat{\mu}_f^2 + \frac{1584}{47} \hat{\mu}_f^4 \right) \ln \frac{\hat{\Lambda}}{2} - \frac{144}{47} (1 + 12 \hat{\mu}_f^2) \ln \hat{m}_D + \frac{319}{940} \left(1 + \frac{2040}{319} \hat{\mu}_f^2 + \frac{38640}{319} \hat{\mu}_f^4 \right) - \frac{24 \gamma_E}{47} (1 + 12 \hat{\mu}_f^2) \right. \\
 & \left. \left. - \frac{44}{47} \left(1 + \frac{156}{11} \hat{\mu}_f^2 \right) \frac{\zeta'(-1)}{\zeta(-1)} - \frac{268}{235} \frac{\zeta'(-3)}{\zeta(-3)} - \frac{72}{47} \left[4 i \hat{\mu}_f \mathfrak{N}(0, z_f) + (5 - 92 \hat{\mu}_f^2) \mathfrak{N}(1, z_f) + 144 i \hat{\mu}_f \mathfrak{N}(2, z_f) + 52 \mathfrak{N}(3, z_f) \right] \right\} \right. \\
 & \left. + 90 \frac{\hat{m}_q^2}{\hat{m}_D} + \frac{315}{4} \left\{ \left(1 + \frac{132}{7} \hat{\mu}_f^2 \right) + \frac{11}{7} (1 + 12 \hat{\mu}_f^2) \gamma_E + \frac{9}{14} \left(1 + \frac{132}{9} \hat{\mu}_f^2 \right) + \frac{2}{7} \mathfrak{N}(z_f) \right\} \hat{m}_D \right] + \frac{\Omega_{\text{NNLO}}^{\text{YM}}}{\Omega_0}
 \end{aligned}$$

HTLpt: NNLO free energy / pressure for pure Yang-Mills at finite T

YM: Andersen, Strickland & NS, PRL 104, 122003 (2010); JHEP 08, 113 (2010)

HTLpt with the electric scale gT incorporated works down to $T \sim 3T_c$



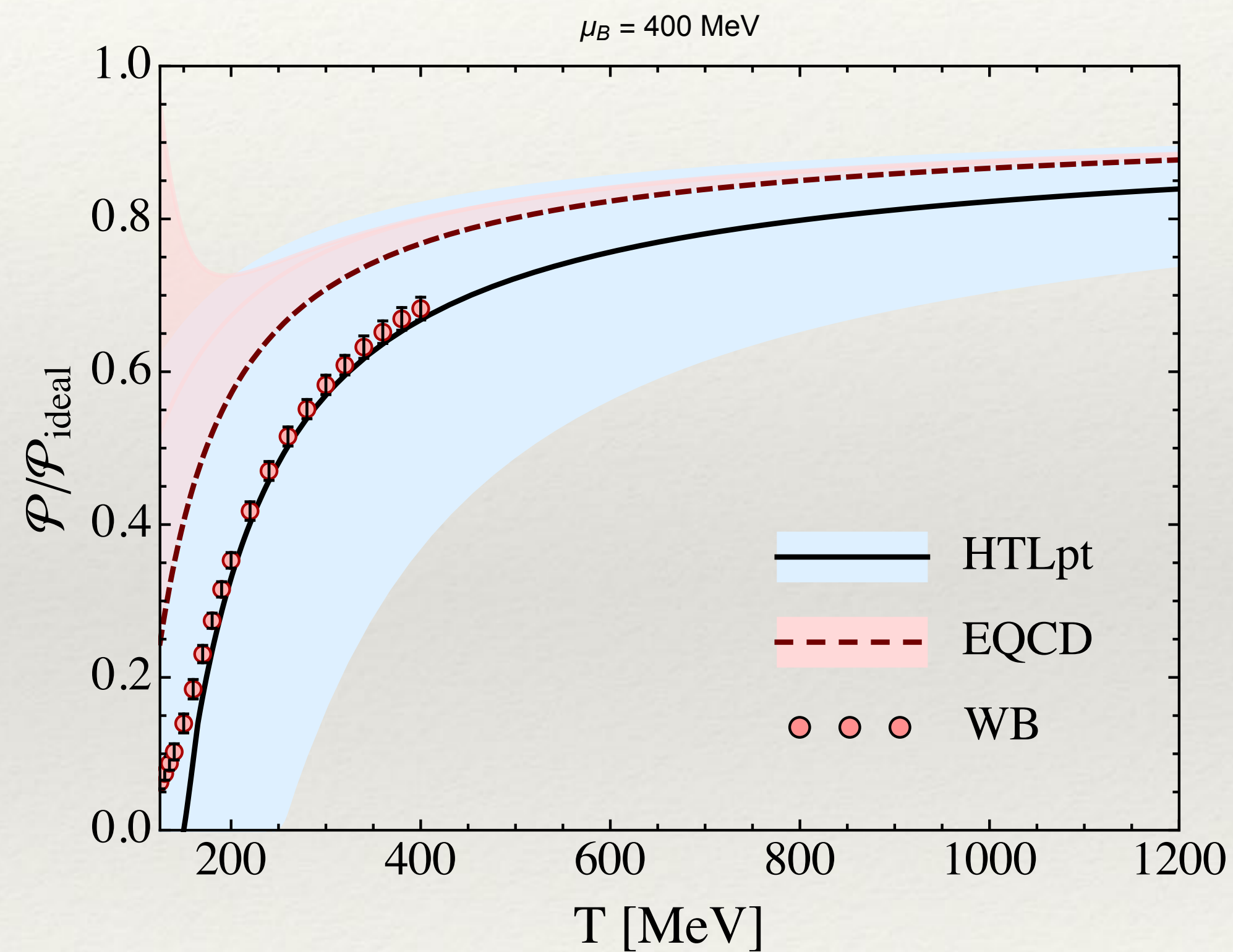
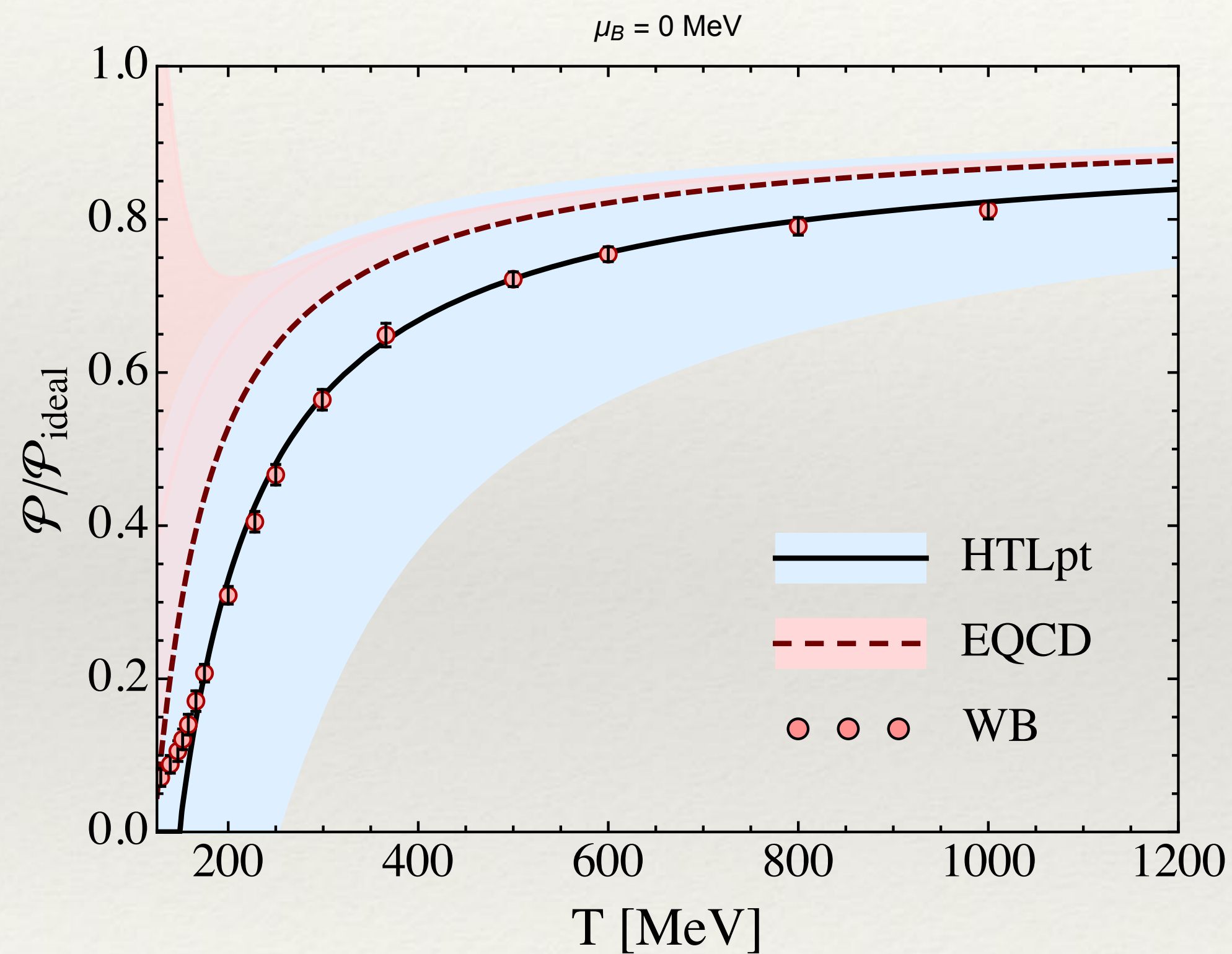
Universality class
of
Yang-Mills / $SU(N)$
phase transition:
 $Z(N)$

$Z(2) \rightarrow$ Ising model
Quantum annealing(!)

HTLpt: NNLO free energy / pressure for QCD at finite T & μ

HTLpt: Andersen, Leganger, Strickland & NS, PLB 696, 468-472 (2011); JHEP 08, 053 (2011) Haque, Bandyopadhyay, Andersen, Mustafa, Strickland & NS, JHEP 05, 027 (2014)

EQCD: Mogliacci, Andersen, Strickland, NS & Vuorinen, JHEP 12, 055 (2013)



Collective excitations of hot plasmas — challenges!

- ❖ Collective excitations of quarks and gluons are crucial in studying the QGP
 - ❖ Debye screening $\rightarrow$ massive quasiparticles $m_D \sim gT$ (electric scale)
 - ❖ **Confining (chromo)magnetic scale g^2T** not included due to the Linde problem
 - ❖ QED and QCD (collective excitations/modes) are **INDISTINGUISHABLE(!)** in conventional thermal-field methods $\rightarrow$ weakly coupled gas
 - ❖ **HOW TO DISTINGUISH?** (key in generating long-range correlations!)

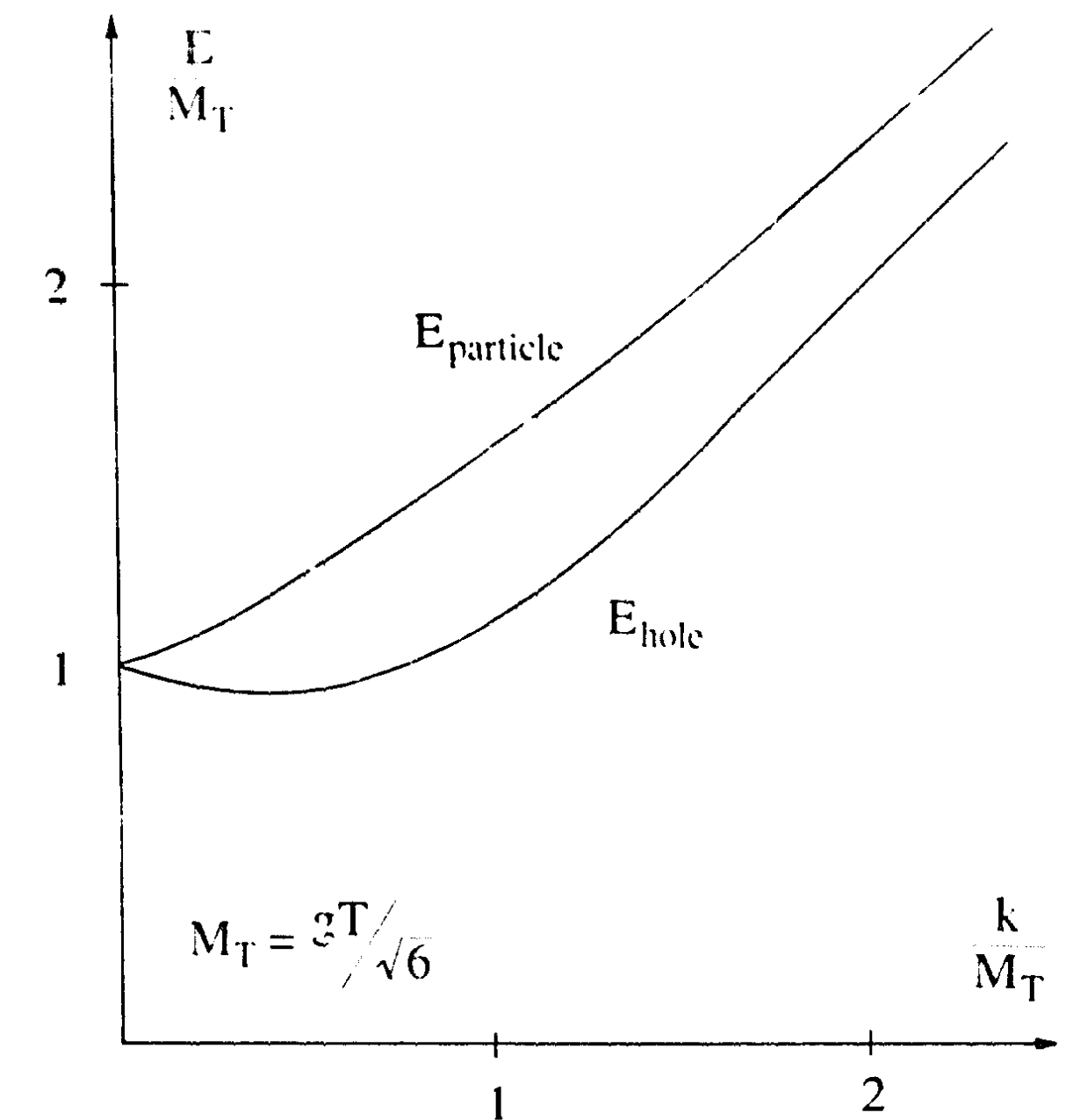
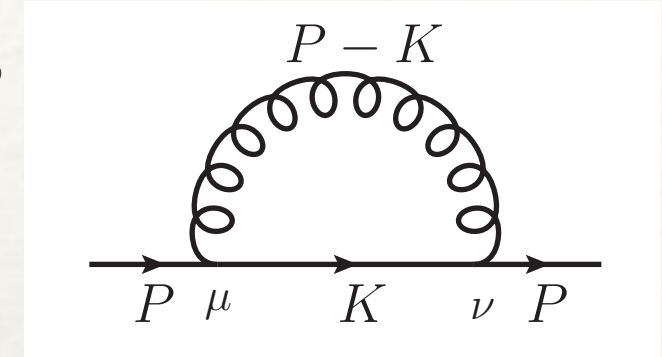


Fig. 1. Dispersion relation for particles and holes.

Gribov-Zwanziger (GZ) quantization: quark thermal self-energy

- ❖ Crucial measure of collective behaviors: screening masses, dispersion relations, spectral functions

$$\Sigma(P) = (ig)^2 C_F \oint_{\{K\}} \gamma^\mu S(K) \gamma^\nu D^{\mu\nu}(P-K)$$



- ❖ At high T ($g \ll 1$), **Gribov-Zwanziger action** [Gribov 78; Zwanziger 89] provides a **renormalizable** framework to study confinement effects at the magnetic scale $g^2 T$ — improvement over Faddeev-Popov action in the **IR**

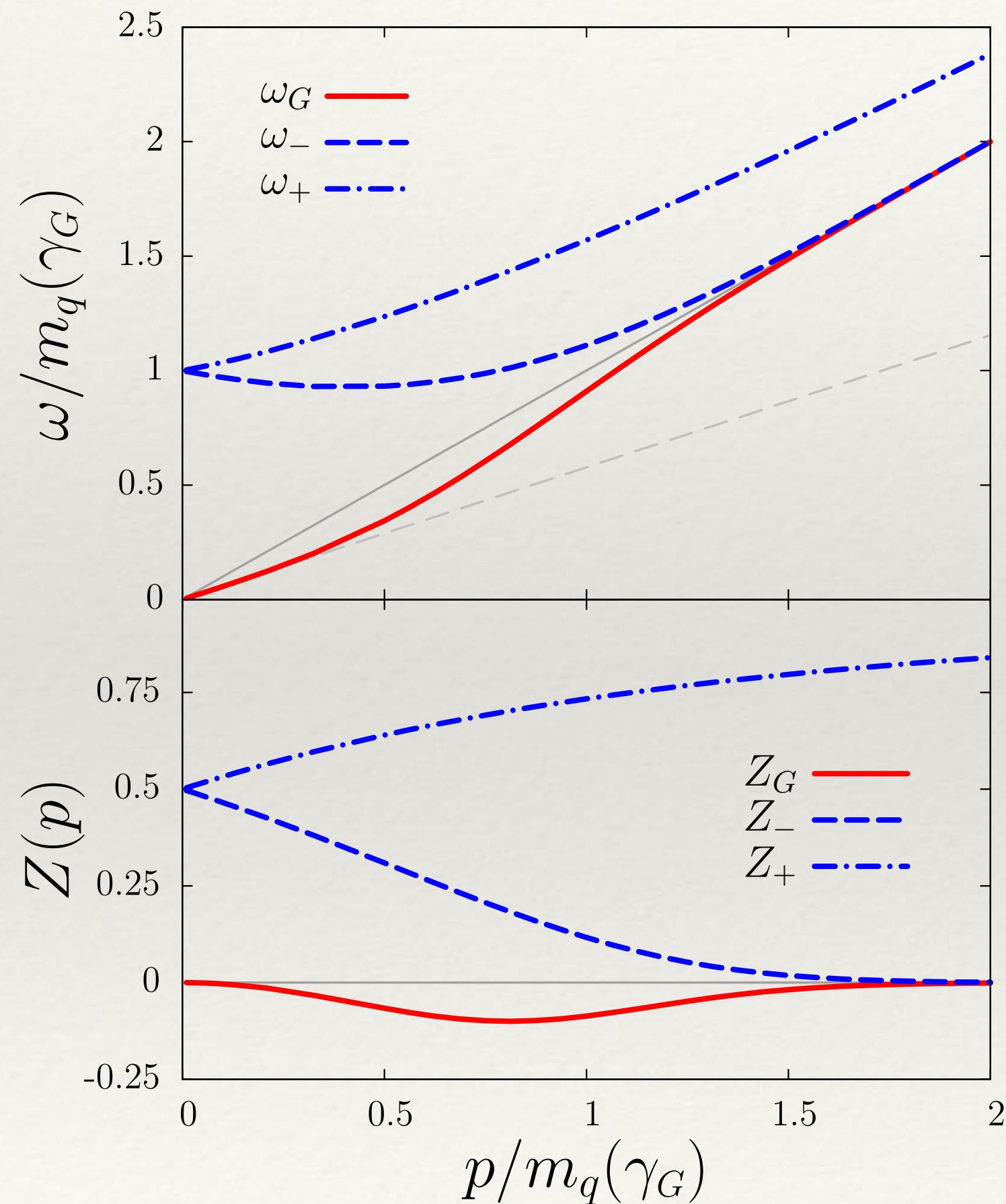
$$\diamond \text{ Faddeev-Popov: } D_{\text{FP}}^{\mu\nu}(P) = \frac{1}{P^2} \left(\delta_{\mu\nu} - \frac{P_\mu P_\nu}{P^2} \right) \rightarrow \text{IR divergent(!)}$$

$$\diamond \text{ Gribov-Zwanziger: } D_{\text{GZ}}^{\mu\nu}(P) = \frac{P^2}{P^4 + \gamma_G^4} \left(\delta_{\mu\nu} - \frac{P_\mu P_\nu}{P^2} \right) \rightarrow \text{IR regular}$$

$$\text{with } \gamma_G = \frac{d}{d+1} \frac{N_c}{4\sqrt{2}\pi} g^2 T \text{ [Zwanziger, PRD 76, 125014 (2007); Fukushima, NS, PRD 88, 076008 (2013)]}$$

GZ: thermal quark dispersion relations & residues of poles

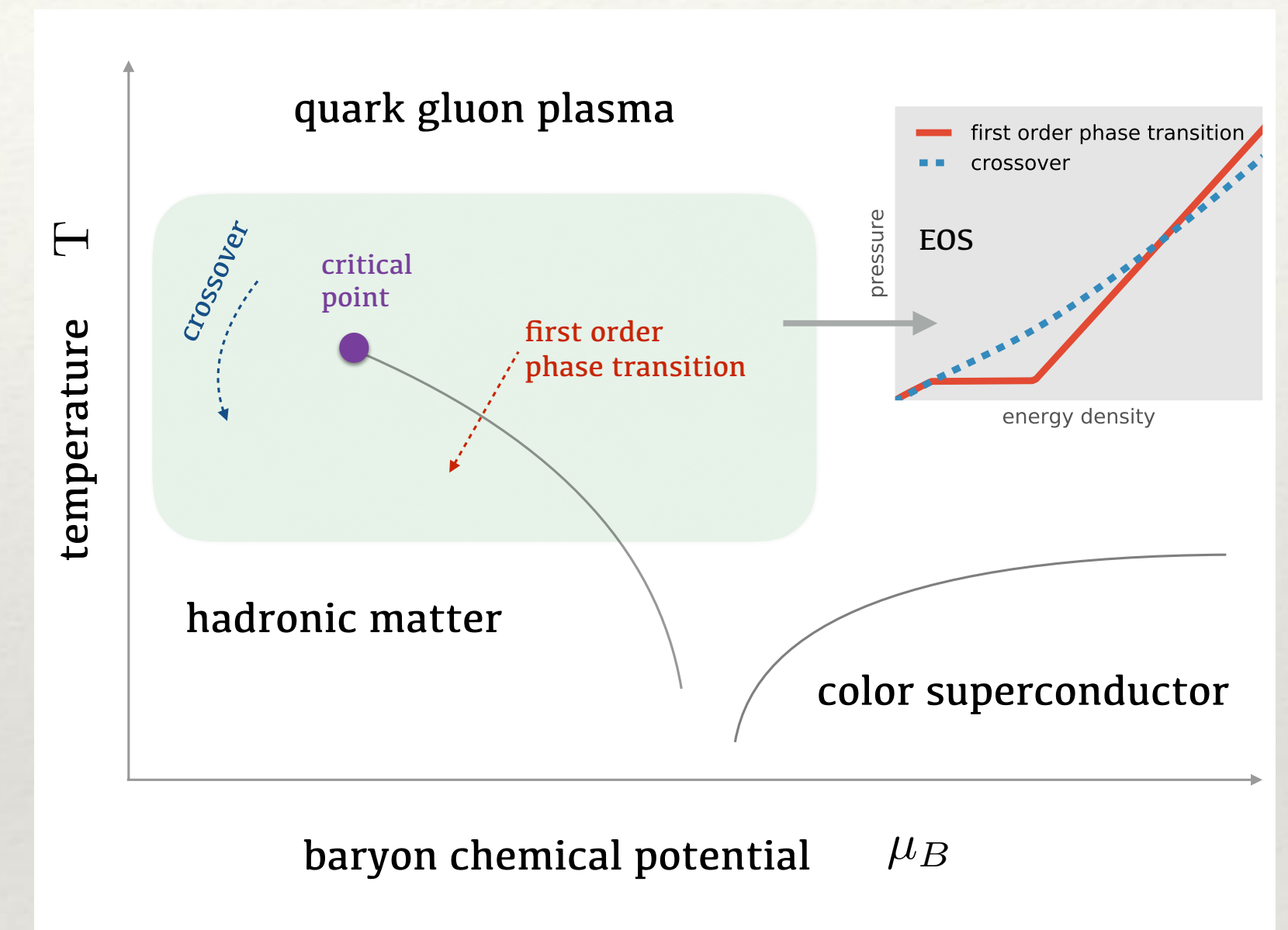
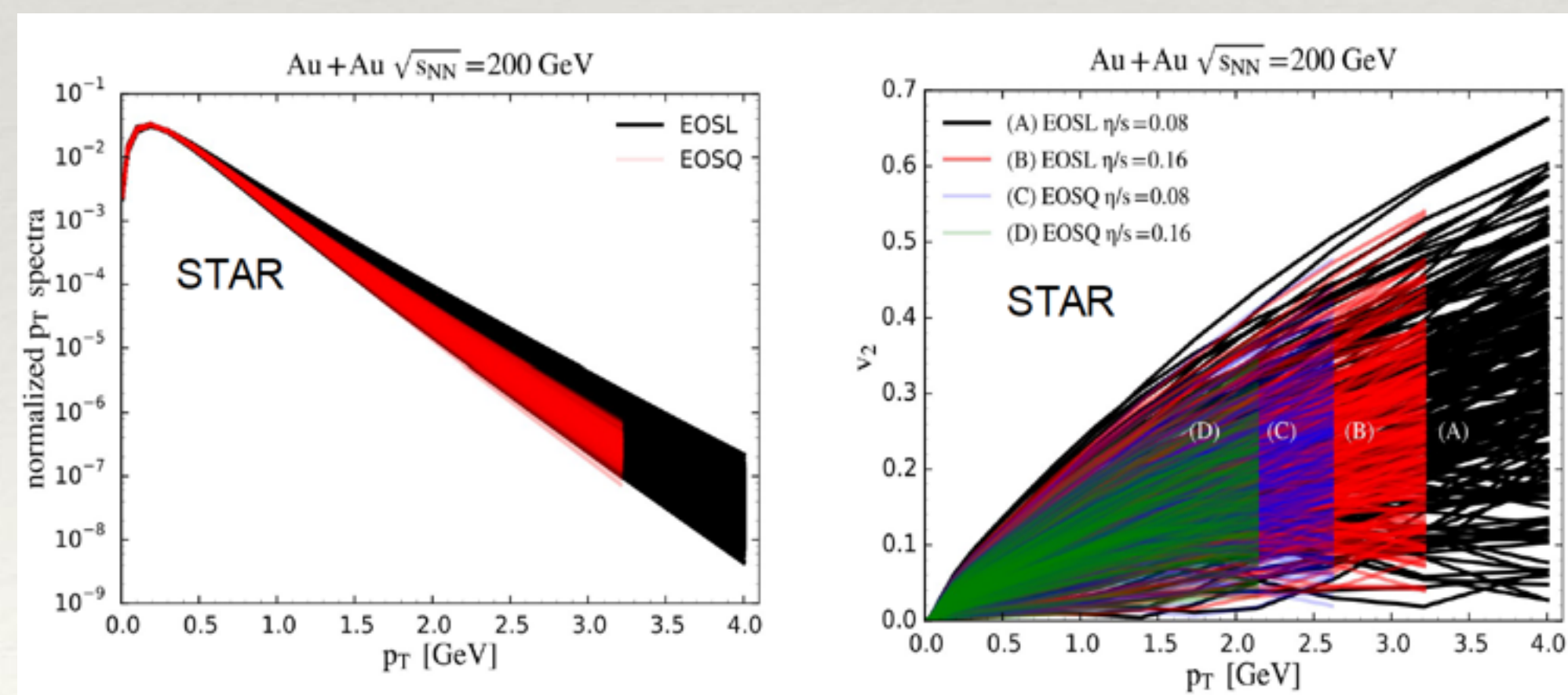
NS & Tywoniuk, PRL 114, 161601 (2015)



- ❖ Quasiparticles recovered — time-like; novel massless mode(!) — space-like
- ❖ Sound wave at small p : $\omega \sim v_s p$ with $v_s = 1/\sqrt{3}$
 - Hydrodynamic-like behaviors → long-range correlations incorporated
- ❖ Crucial for heavy-ion phenomenology at RHIC and LHC energies
 - Viscosities: $\frac{\zeta}{\eta} \propto \left(\frac{1}{3} - c_s^2\right)$, strongly coupled liquid(!)
[Florkowski, Ryblewski, NS & Tywoniuk, Acta Phys. Polon. B 47, 1833 (2016); PRC 94, 044904 (2016)]
- ❖ 1st evidence of positivity violation for thermal quarks/collective mode
 - Genuine feature of non-Abelian theories, distinctive from QED(!)
 - Confinement ($\gamma_G \sim g^2 T$) indispensable for thermal QCD even at $T \rightarrow \infty$ (!)
- ❖ Thermal spectral sum rule: $\int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \rho(\omega, p) = 1$, math(!) [Du, NS, Tywoniuk, preliminary]

Extracting physical info from heavy-ion measurements — challenge!

- ❖ Heavy-ion measurements $\rho(p_T, \Phi)$ — highly complex
- ❖ $\rho(p_T, \Phi) \rightarrow$ bulk properties, e.g. equation of state (EoS)
 - ❖ No direct link
 - ❖ Conventional observables — model dependent

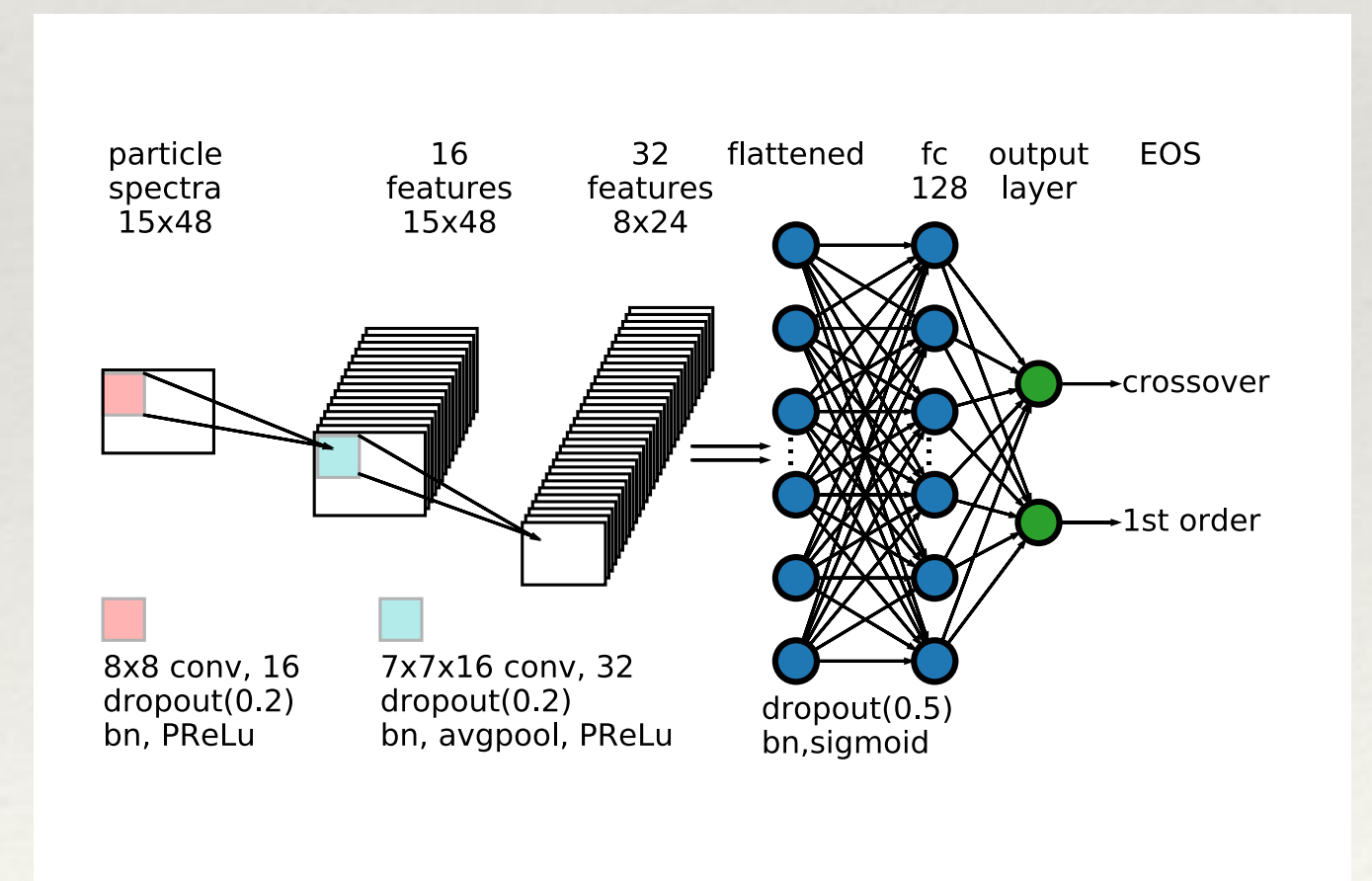
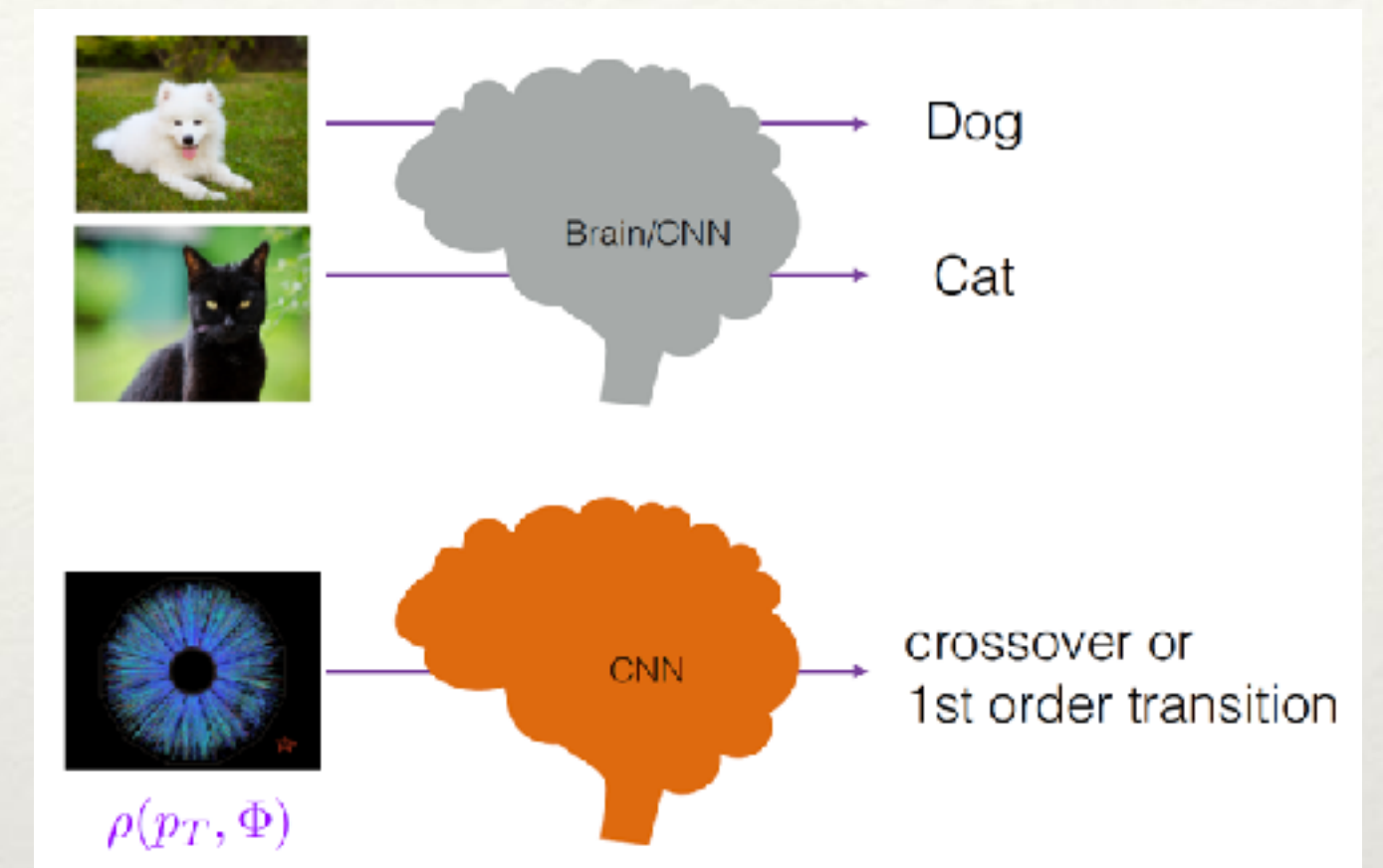


An EoS-meter of QCD transition from deep learning

Pang, Zhou, NS, Peterson, Stöcker & Wang, Nat. Commun. 9, 210 (2018)

- ❖ Conventional neural networks (CNN) & hydrodynamic simulations
- ❖ An EoS-meter of QCD transition from CNN
 - ❖ Direct map between $\rho(p_T, \Phi)$ and the type of EoS — model independent
 - ❖ 1st application of deep learning to high-energy nuclear physics
 - ❖ 2018 Nat. Commun. “Top 50: Physics”; SAMSON AI Grant

TESTING DATA	GROUP 0	GROUP 1	GROUP 2
Number of events	4000	7343	10953
Accuracy	$99.88 \pm 0.04\%$	$93.46 \pm 1.35\%$	$93.91 \pm 3.92\%$



Quantum computing

Computing power of high performance computing (HPC) — challenge!

- ❖ 2022 Nobel Prize in Physics: Alain Aspect, John F. Clauser and Anton Zeilinger
 - ❖ "for experiments with entangled photons, establishing the violation of Bell inequalities and pioneering quantum information science"
- ❖ **Saturation** of conventional HPC's computing power — **bottleneck** to many applications, e.g.
 - ❖ Condensed matter physics — geometrical frustration: very time & power consuming
 - ❖ Telecommunication — network routing: very power consuming, low efficiency
 - ❖ Materials design — optimal formulation: extremely time & power consuming, suboptimal solutions
 - ❖ Intelligent transportation — entire network navigation: beyond the scope of conventional HPC
- ❖ Quantum computing as a next generation HPC technology — potentially efficient solutions

Birth of quantum computing

- ❖ Physics of Computation Conference, MIT & IBM, May 6-8, 1981
- ❖ Richard Feynman: “Nature isn’t classical, dammit, and if you want to make a simulation of nature, you’d better make it quantum mechanical...”
- ❖ Quantum characteristics: superposition, entanglement, tunneling, etc



Physics of Computation Conference Endicott House MIT May 6-8, 1981

1 Freeman Dyson	13 Frederick Kantor	25 Robert Suaya	37 George Michals
2 Gregory Chaitin	14 David Leinweber	26 Stan Kugel	38 Richard Feynman
3 James Gentilefield	15 Konrad Zuse	27 Bill Gosper	39 Laurie Lingham
4 Norman Packard	16 Bernard Zeigler	28 Lutz Priese	40 Thiruganjan
5 Panos Tsigomeras	17 Carl Adam Petri	29 Madhu Gupta	41 ?
6 Jerome Rothstein	18 Anato. Holt	30 Paul Benioff	42 Gerard Vichniac
7 Carl Hewitt	19 Roland Vollmar	31 Hans Moravec	43 Leonid Levin
8 Norman Hardy	20 Hans Bremerman	32 Ian Richards	44 Lev Levitin
9 Edward Fredkin	21 Donald Grospan	33 Marian Pour El	45 Peter Gacs
10 Tom Toffoli	22 Markus Buechtiker	34 Danny Hillis	46 Dan Greenberger
11 Rolf Landauer	23 Otto Floberth	35 Arthur Bucks	
12 John Wheeler	24 Robert Lewis	36 John Cocke	

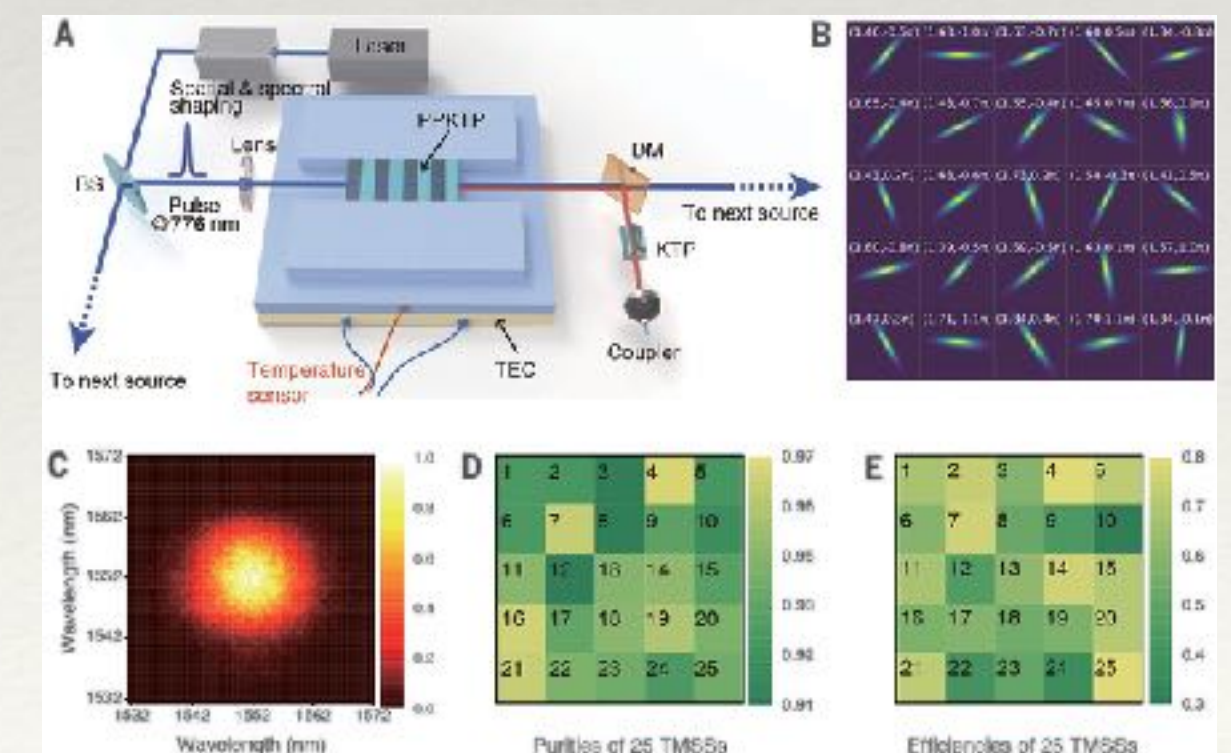
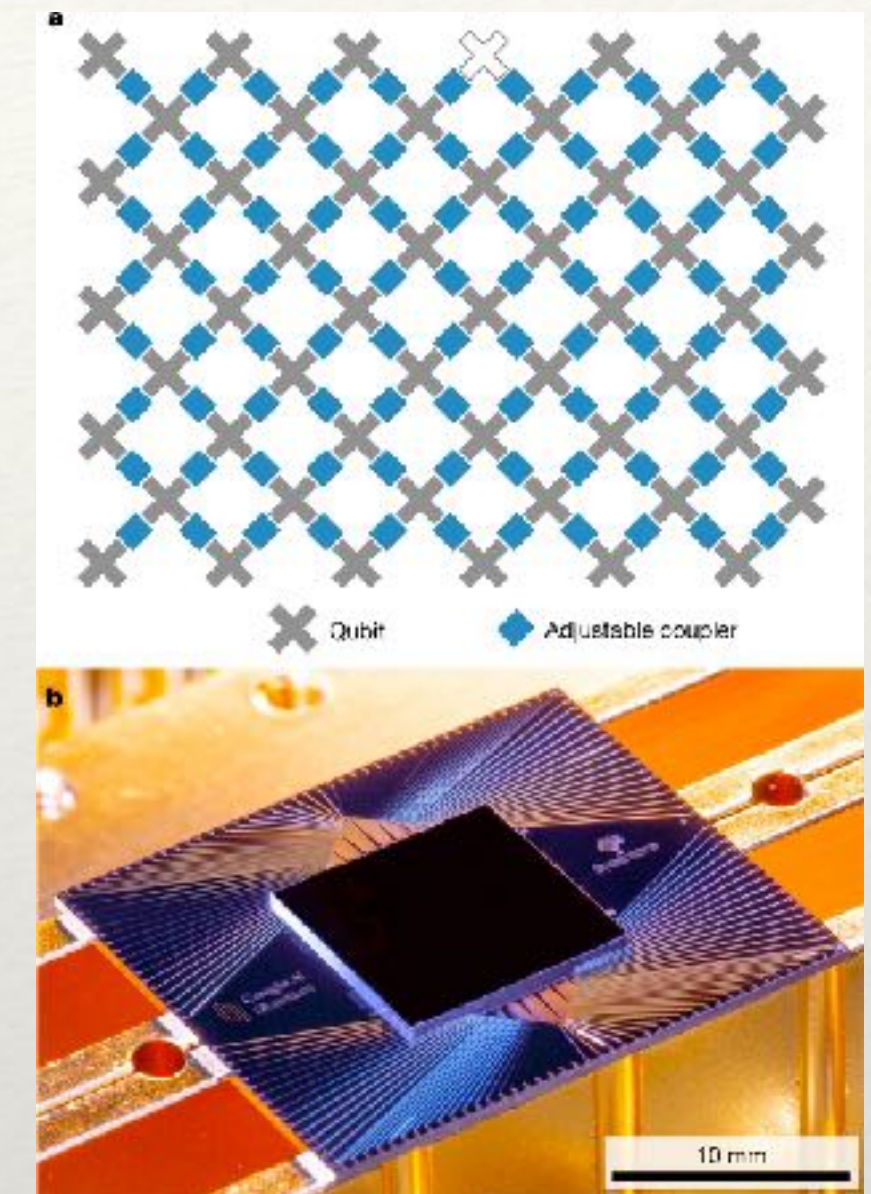
Quantum advantage

Google Quantum AI (2019)

- ❖ “Quantum supremacy using a programmable superconducting processor” [Nature 574, 505 (2019)]
 - 53 superconducting qubits Sycamore processor
 - 200 seconds random sampling — 10,000 years for HPC
 - Can be tackled by HPC within days by IBM Watson [arXiv:1910.09534] & NSC Wuxi [arXiv:2111.01066]
 - Can be tackled by HPC within hours by ITP CAS [PRL 129, 090502 (2022)]

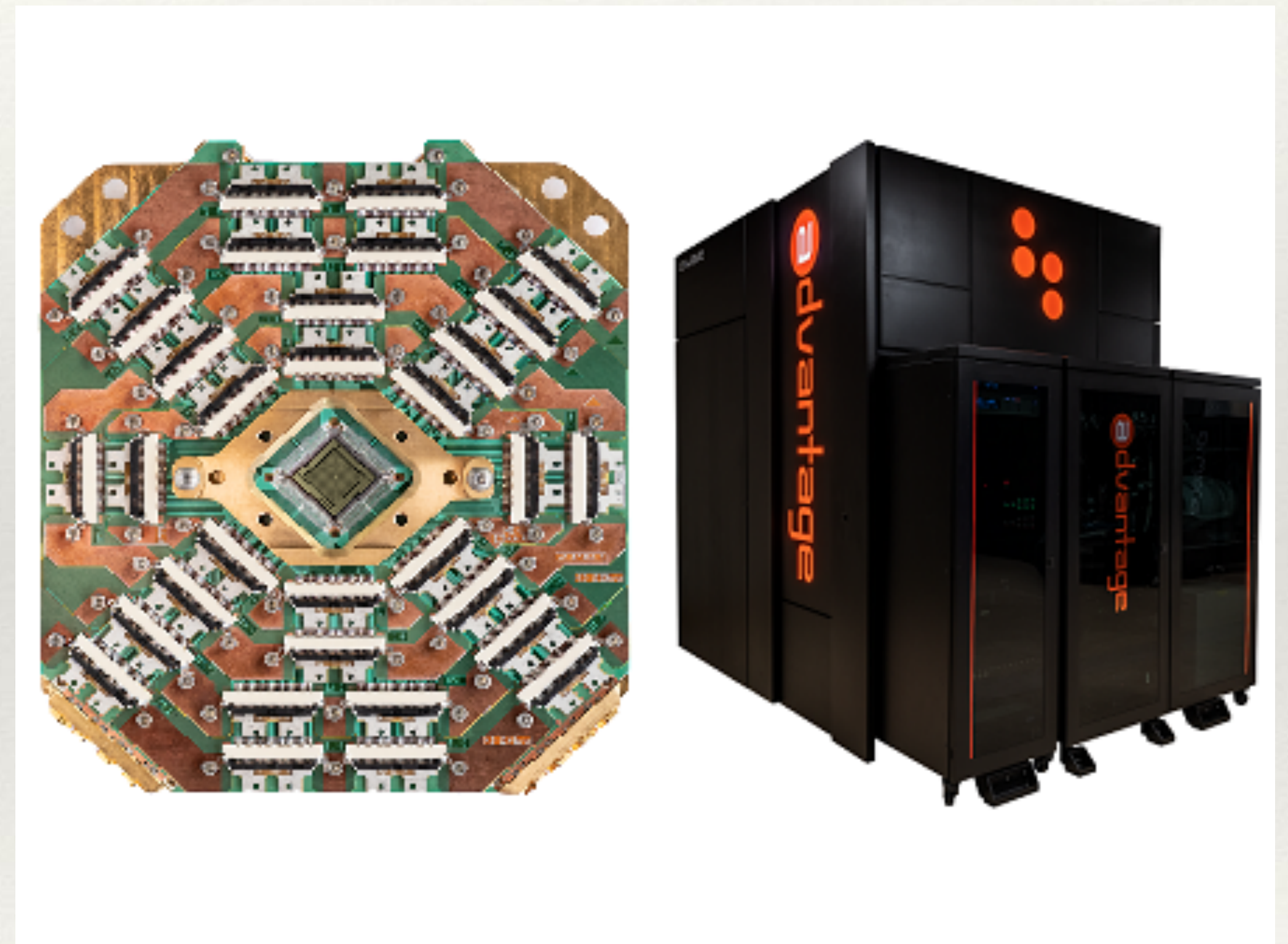
University of Science and Technology of China (2020)

- ❖ “Quantum computational advantage using photons” [Science 370, 1460 (2020)]
 - 76 photonics qubits Jiuzhang quantum computer
 - 1st time Boson sampling: 10^{14} faster than HPC (1 second Jiuzhang ~ 3M years HPC)
 - Jiuzhang 2.0 — 113 qubits, 10^{24} faster than HPC [PRL 127, 180502 (2021)]



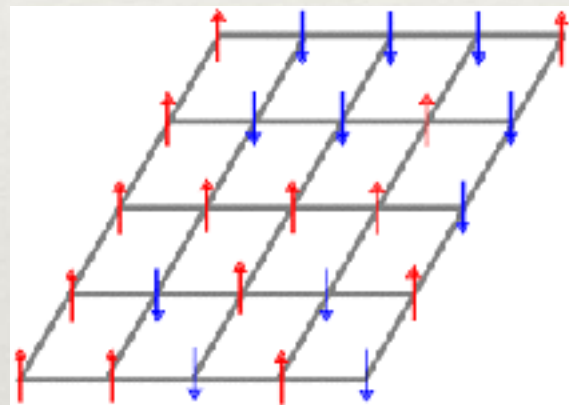
Quantum annealer

- ❖ Gate-model **universal** QC limited in size and error correction
- ❖ QC hardware implementation based on quantum annealing, **specific computer** for combinatorial optimization problems
- ❖ Combinatorial optimization: find an optimal object from a finite set of objects, e.g. traveling salesman problem — **NP-hard(!)**
- ❖ January 17, 2022, Jülich Supercomputing Center, D-Wave Advantage System, 5,000+ qubits
- ❖ QA emulator: Fujitsu Digital Annealer (~10,000 times faster than HPC), 4rd generation (May 18, 2022), emulating 100,000 qubits
- ❖ **Ready for practical applications**



Quantum annealing in a nutshell

- ❖ Combinatorial optimization: finding global minimum of Ising Problem

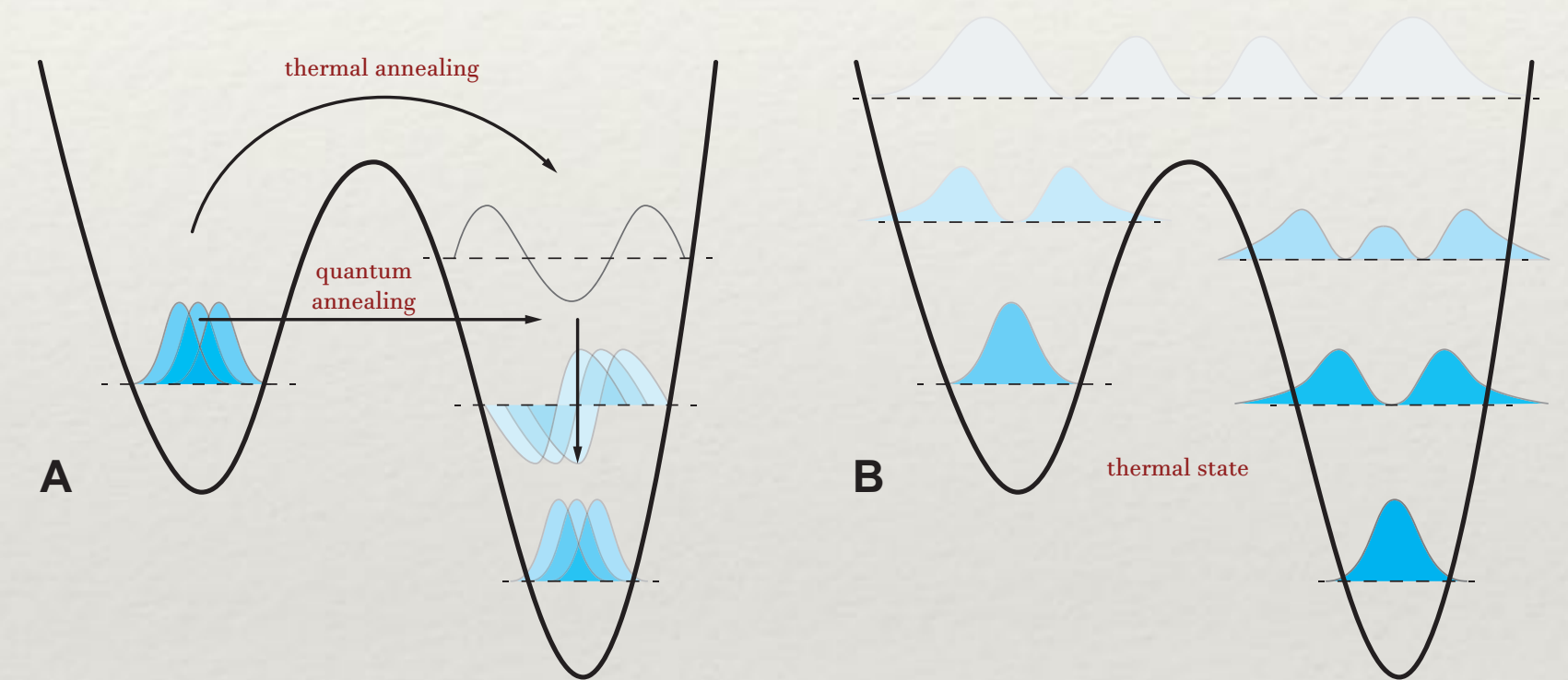


$$\min \left(\sum_i \sum_{j>i} J_{ij} \sigma_i \sigma_j + \sum_i h_i \sigma_i \right), \text{ with } \sigma_i \in \{-1, 1\}$$

- ❖ Quantum annealing: quantum version of simulated annealing

- Application of quantum many-body theory

- ❖ D-Wave: $T_{\text{anneal}} = 20\mu s, 25\text{kW}$

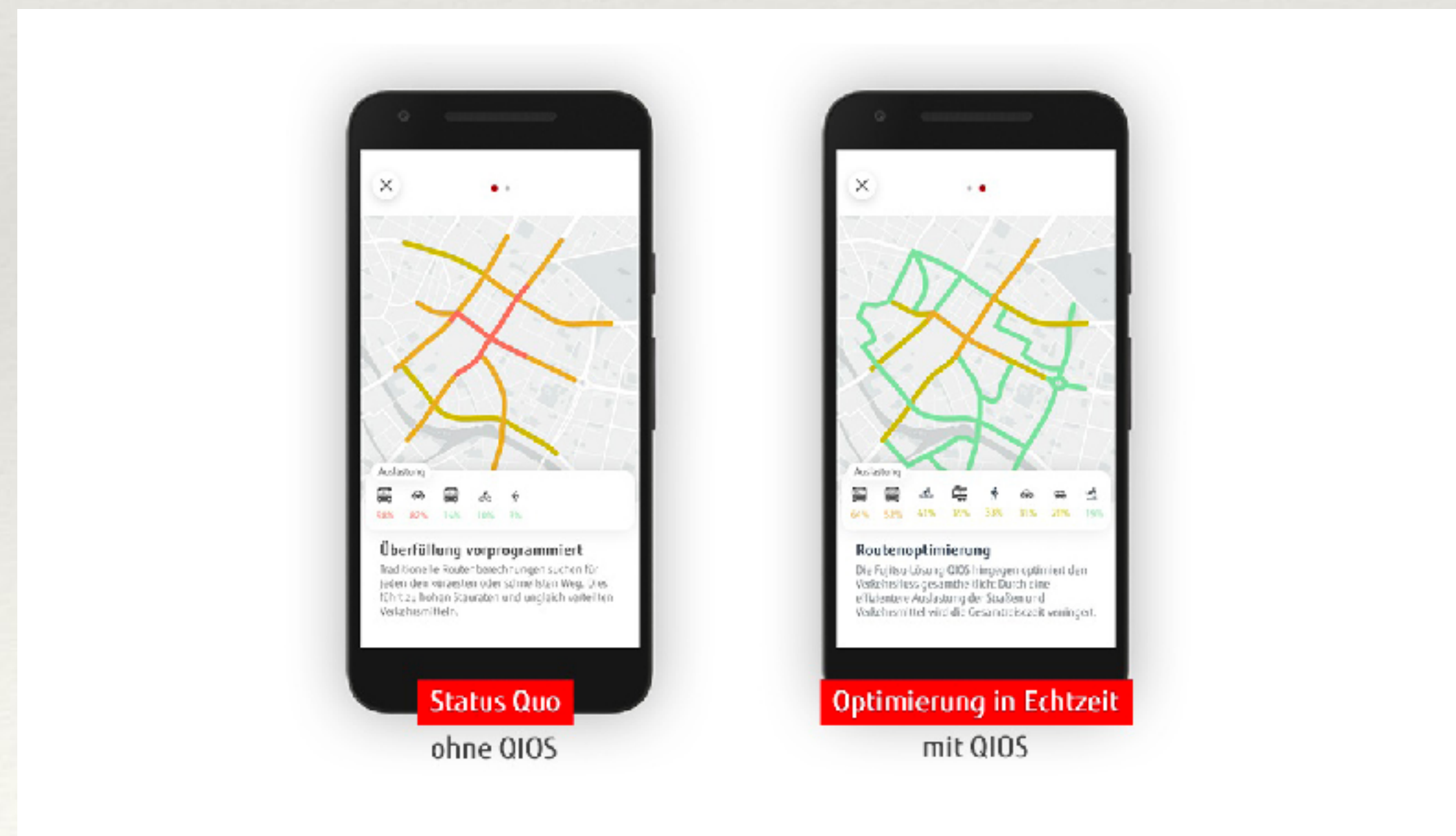
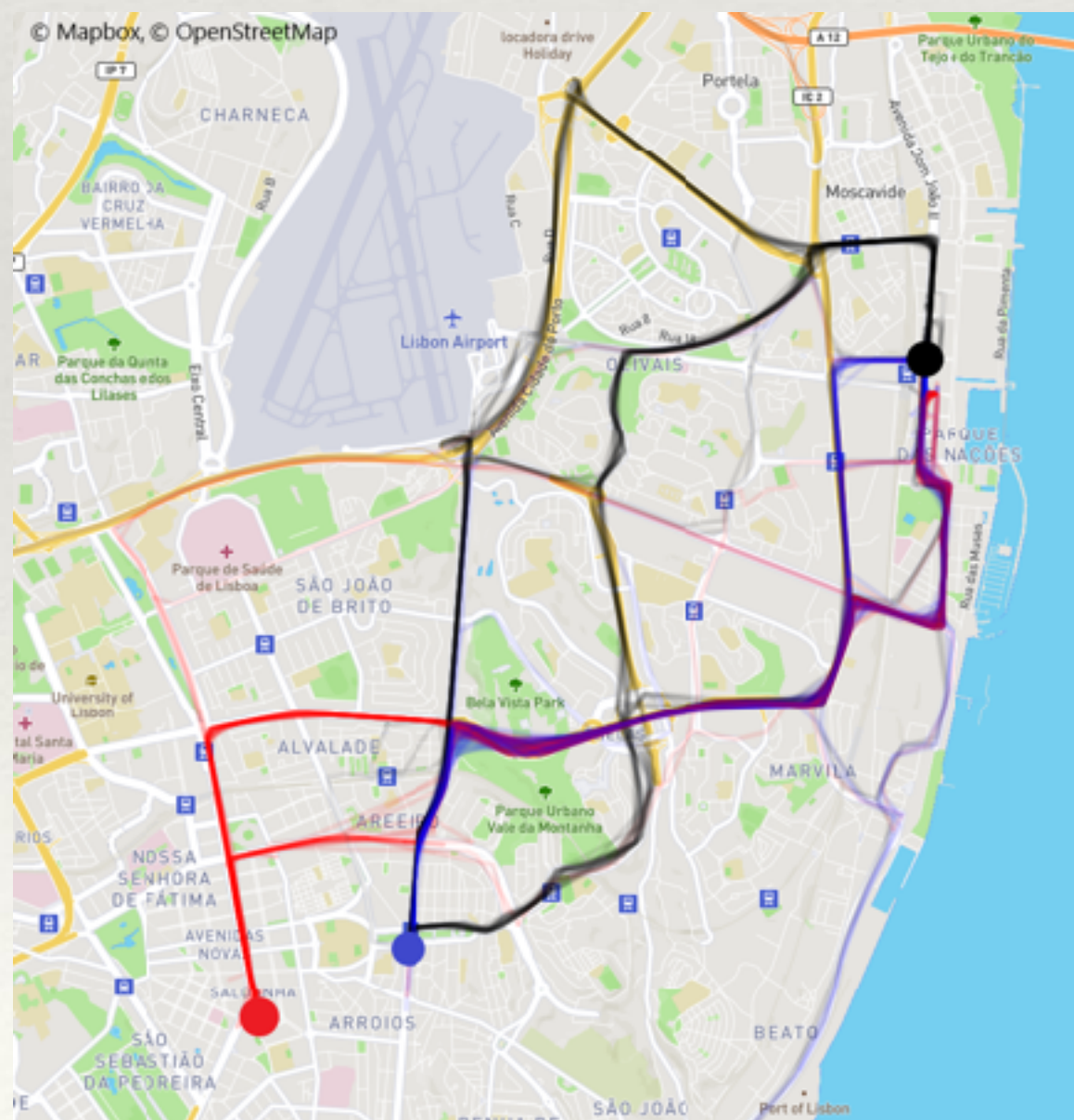


Application: entire network navigation

S. Yarkoni et al., *Quantum Shuttle: Traffic Navigation with Quantum Computing*, arXiv:2006.14162

<https://www.fujitsu.com/de/microsite/its-world-congress/mozart/index.html>

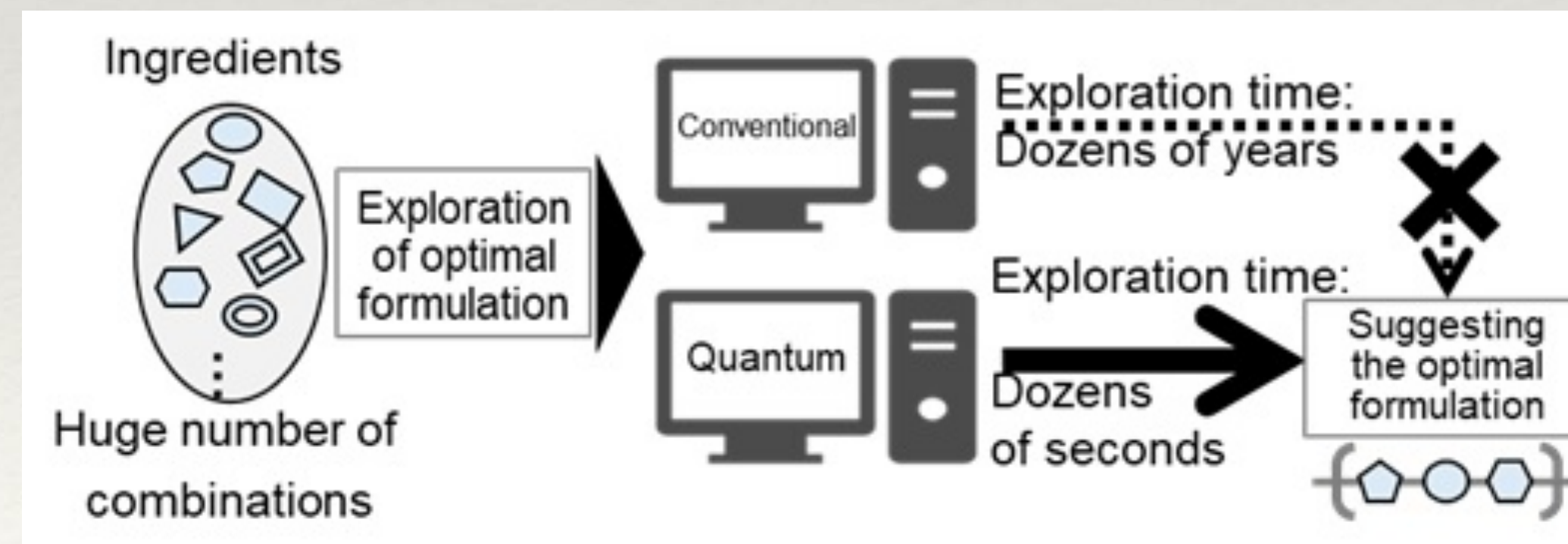
- ❖ Quantum Shuttle: Volkswagen, Hexad (2019); MOZART-Projekt: Fujitsu, Hamburg Port Authority, TU Graz (2021)
- ❖ Quantum navigation boosts the efficiency of the entire network: saving time, traffic, emission
 - Much more optimized results than HPC, ability of entire network coordination



Application: materials design

<https://www.sdk.co.jp/english/news/2022/41713.html>

- ❖ The theoretical combinations of ingredients and mixing ratios in semiconductor materials design are more than 10^{50}
 - Conventional AI computation takes dozens of years, so only part of all the combinations was analyzed (suboptimal)
- ❖ February 2022: Showa Denko
 - Performing such AI analysis with Fujitsu's Digital Annealer only takes dozens of seconds, 72,000 times faster than conventional AI (with limited combinations explored)
 - Obtained semiconductor from optimal formulation is expected to have 30% higher performance than the one from conventional AI

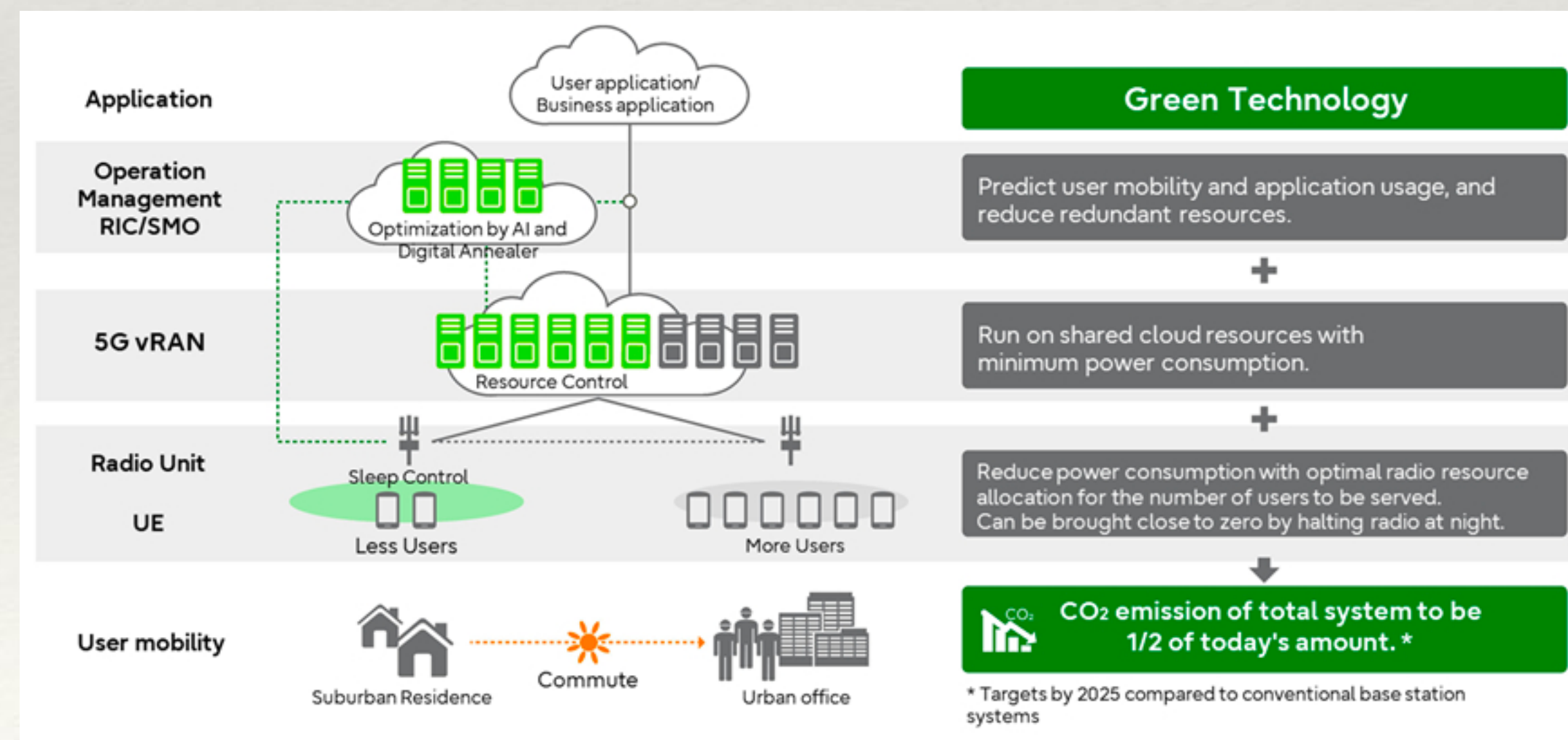


Application: 5G wireless network power & emission reduction

<https://www.fujitsu.com/global/about/resources/news/press-releases/2022/0224-01.html>

❖ February 2022: Fujitsu

- New virtualized radio access network (vRAN) software technology leveraging Fujitsu's AI and Digital Annealer technologies to optimize computing resources of the entire 5G wireless network
- Estimation of reducing overall system CO2 emissions by 50% or more comparing to conventional base station systems by 2025

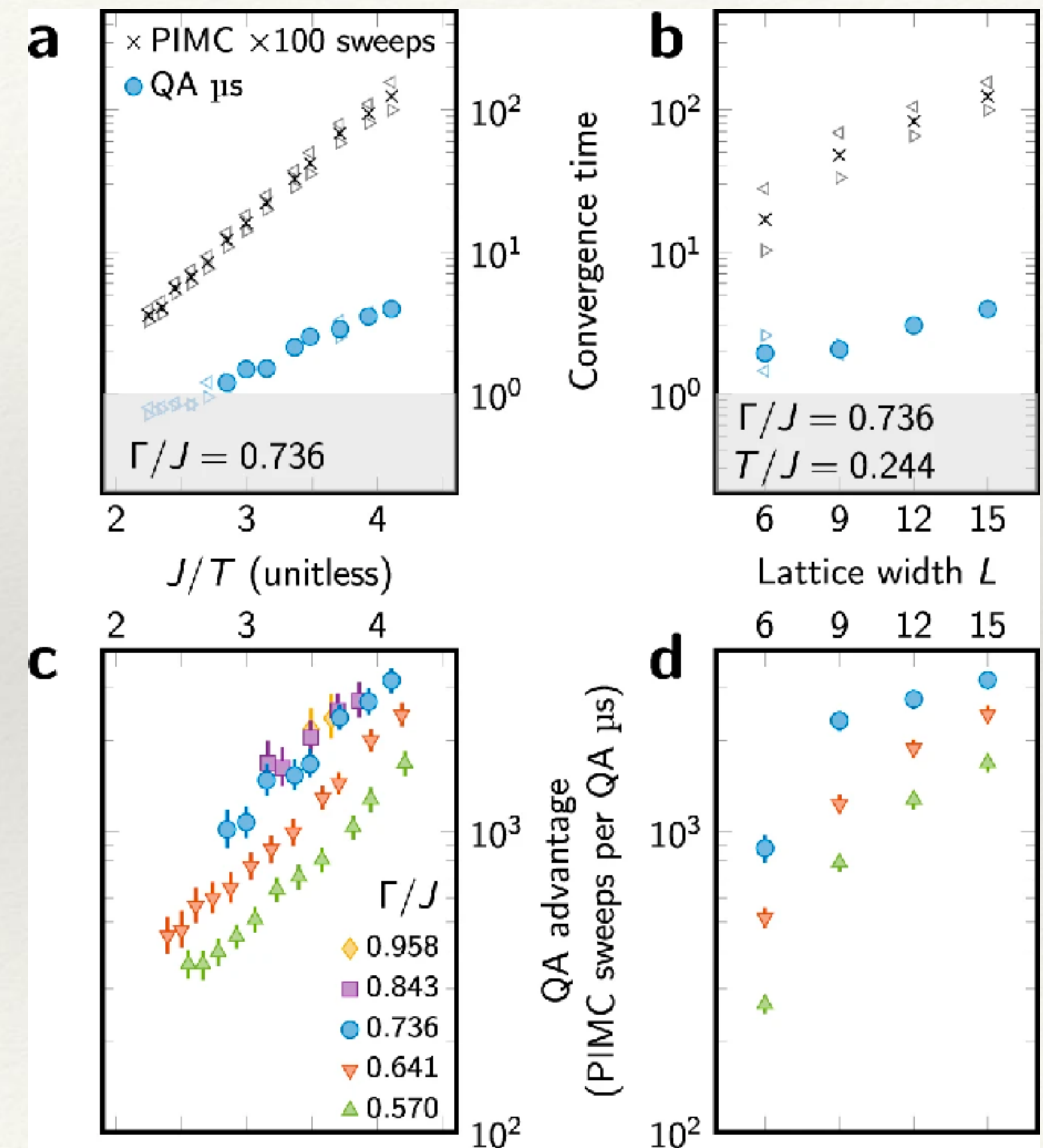
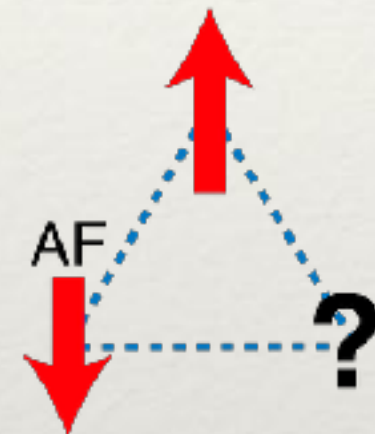


Application: quantum simulation of geometrically frustrated magnets

King, Raymond, Lanting *et al.*, Nat Commun. 12, 1113 (2021)

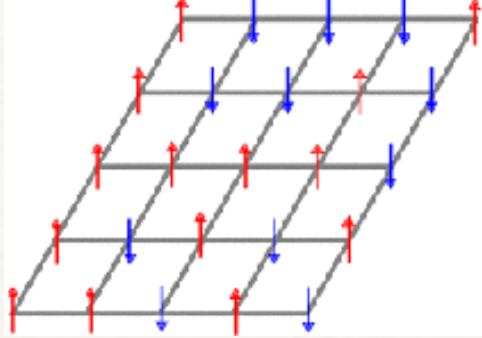
❖ D-Wave

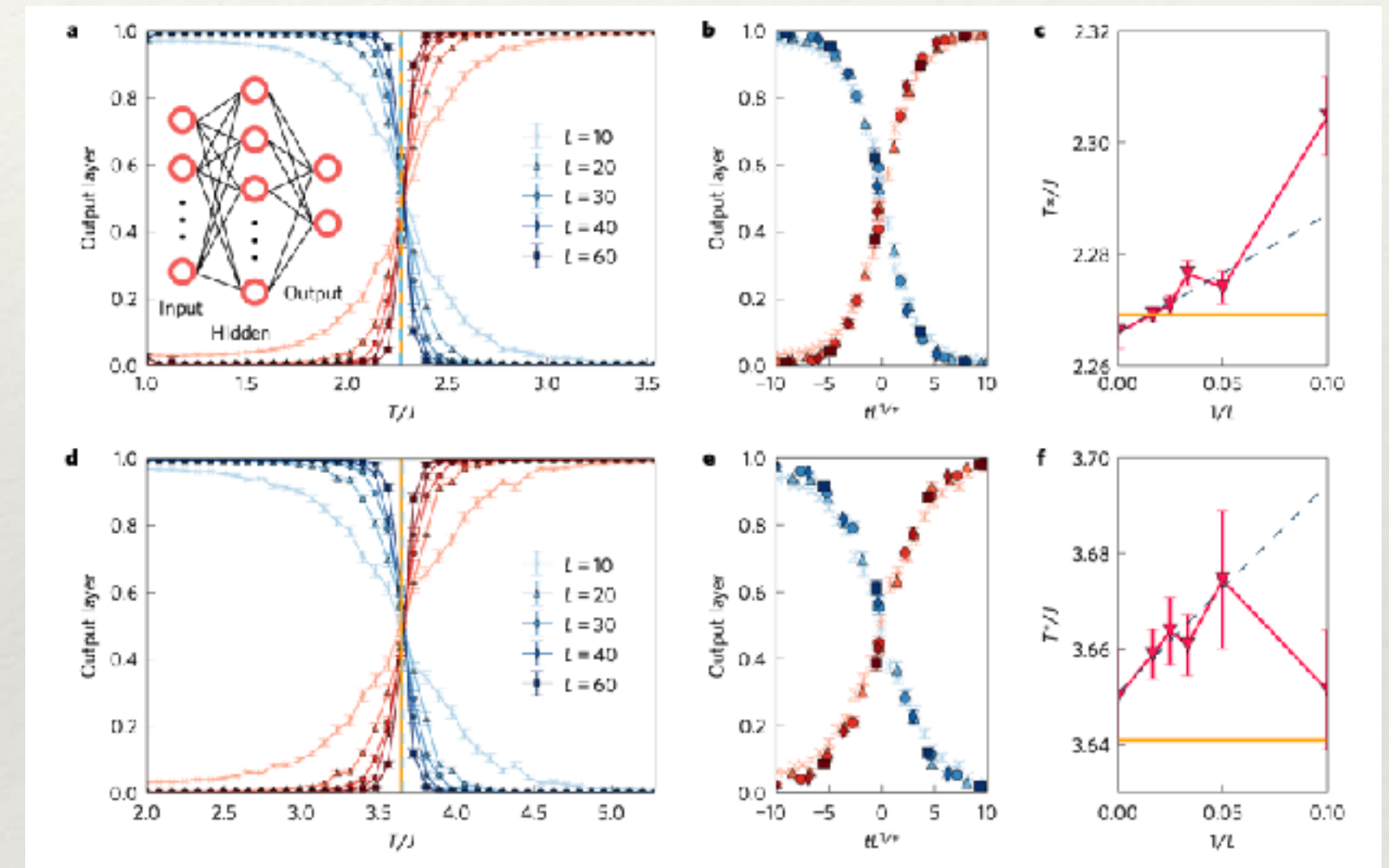
- Geometrical frustration — challenging!
- “Scaling advantage over path-integral Monte Carlo in quantum simulation of geometrically frustrated magnets”
- QA equilibration is 3 million times faster than PIMC on a CPU



On the generalizability of artificial neural networks in spin models

Yau, NS, SciPost Phys. Core, 5 032 (2022)

- ❖ Ising model: $\mathcal{H} = -J \sum_{\langle i,j \rangle} \sigma_i \sigma_j$, where $\sigma \in \{-1, 1\}$

- ❖ “Machine learning phases of matter” [Carrasquilla & Melko, Nat. Phys. 13, 431 (2017)]
 - Supervised learning, big data training: critical exponent $\nu = 1.0 \pm 0.2$
- ❖ Exponential reduction of training data state space
 - $2^N \rightarrow 3$, with N the number of spins, typically $\mathcal{O}(10^2) \sim \mathcal{O}(10^6)$
 - $\{-1, \dots, -1\}$ & $\{1, \dots, 1\}$ for ordered phase; $\{0, \dots, 0\}$ for disordered phase
 - Critical exponent $\nu = 1.01 \pm 0.11$
- ❖ QAML outperforms conventional ML in few data training [Mott *et al.*, Nature 550, 375 (2017)]
 - Training strategy for spin model related problems, e.g. condensed matter physics, materials science, complex networks, etc



Conclusions and outlook

Conclusions and outlook

- ❖ Quantum many-body systems offer promising opportunities
- ❖ Thermal quantum field theory: incorporation of confinement to thermal QCD for a proper understanding of strong force
 - ❖ QCD collective excitations: full analysis
 - ❖ Improved HTL effective theory incorporating both electric and magnetic scales (heavy ions, astrophysics, cosmology, etc)
 - ❖ Strongly coupled systems in condensed matter physics, e.g. quantum spin liquids
- ❖ Quantum computing: applications of quantum many-body theories
 - ❖ Lattice simulation: high energy physics, condensed matter physics
 - ❖ Telecommunication: wireless and optical network optimizations
 - ❖ Energy & sustainability: energy network optimizations, trading strategies

Shandong Institute of Advanced Technology (SDIAT)

- ❖ Established in November 2019
- ❖ Honorary Director: Prof. Samuel Chao Chung Ting
- ❖ 150,000 m² campus to support 1,200 staffs
 - ❖ Particle Physics Research Center
 - ❖ Thermal Science Research Center
 - ❖ AMS Data Center
 - ❖ Material Science Research Center
- ❖ Faculty positions at all ranks
 - ❖ Competitive salaries & starting supports

