

ED4D

Disruptive Digitalization
FOR Decarbonization



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FUTURA EXPO
4 ottobre 2022

FUTURA EXPO
ECONOMIA X L'AMBIENTE 2 · 3 · 4 OTTOBRE 2022





Tecnologie digitali

Emissioni

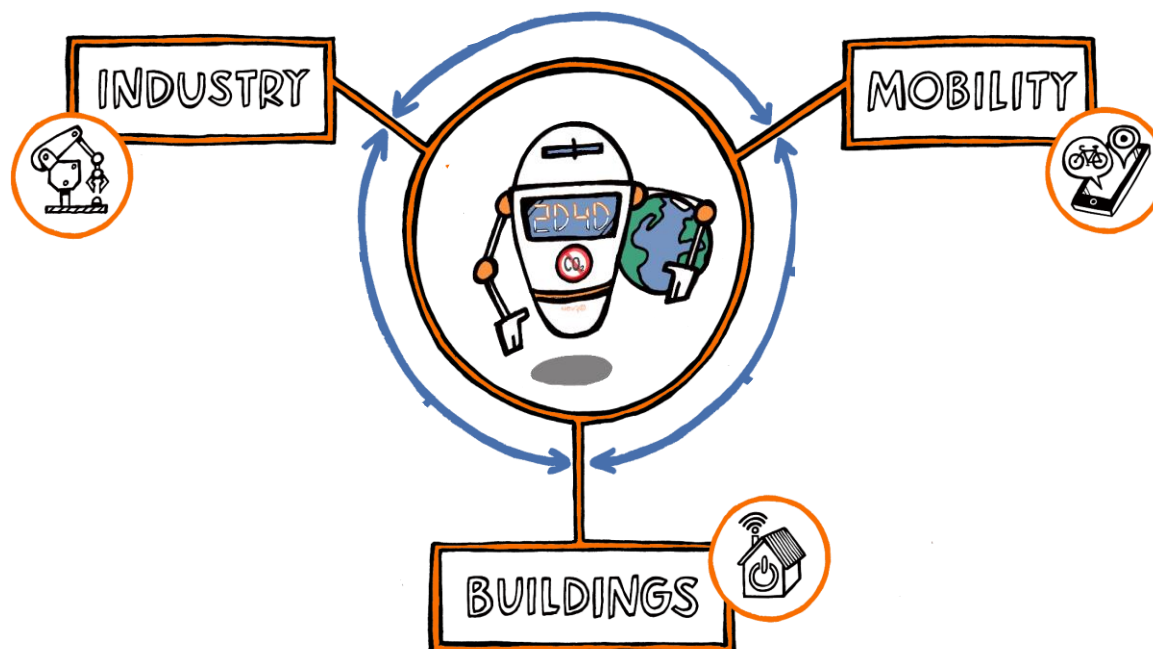


Disruptive Digitalization
FOR Decarbonization

Impatti socio-economici



Transizione energetica





Tecnologie digitali

Emissioni

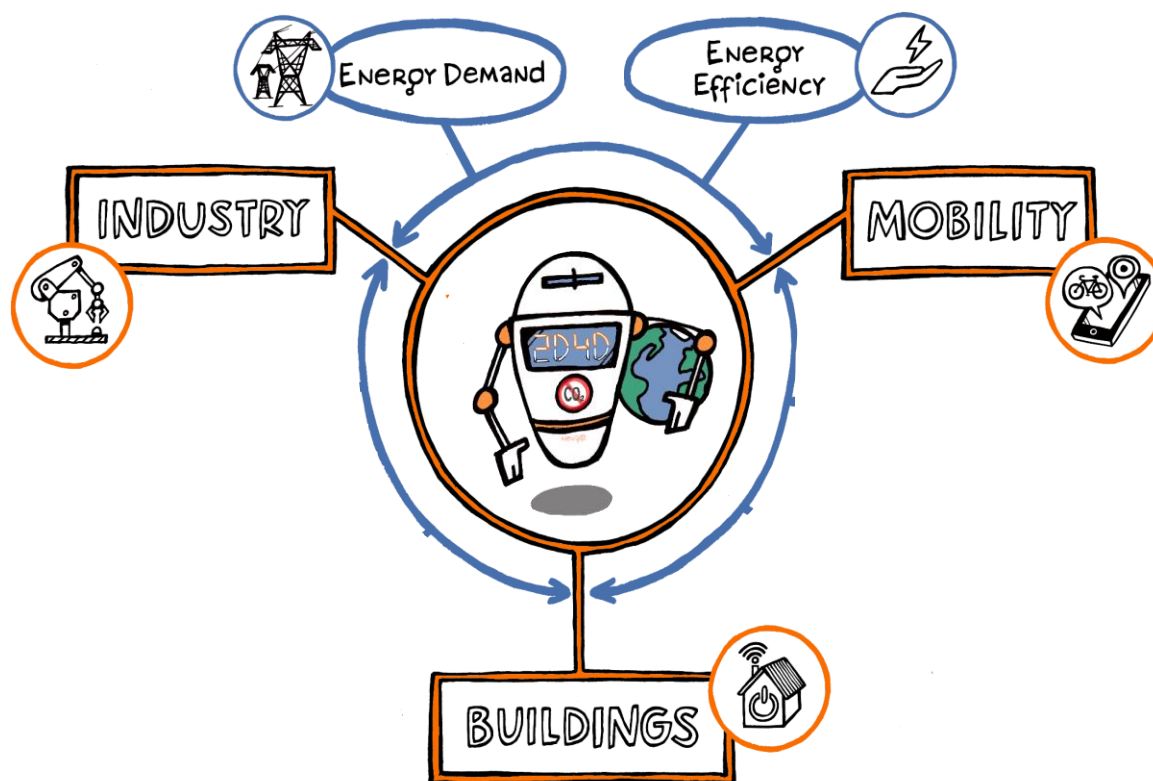


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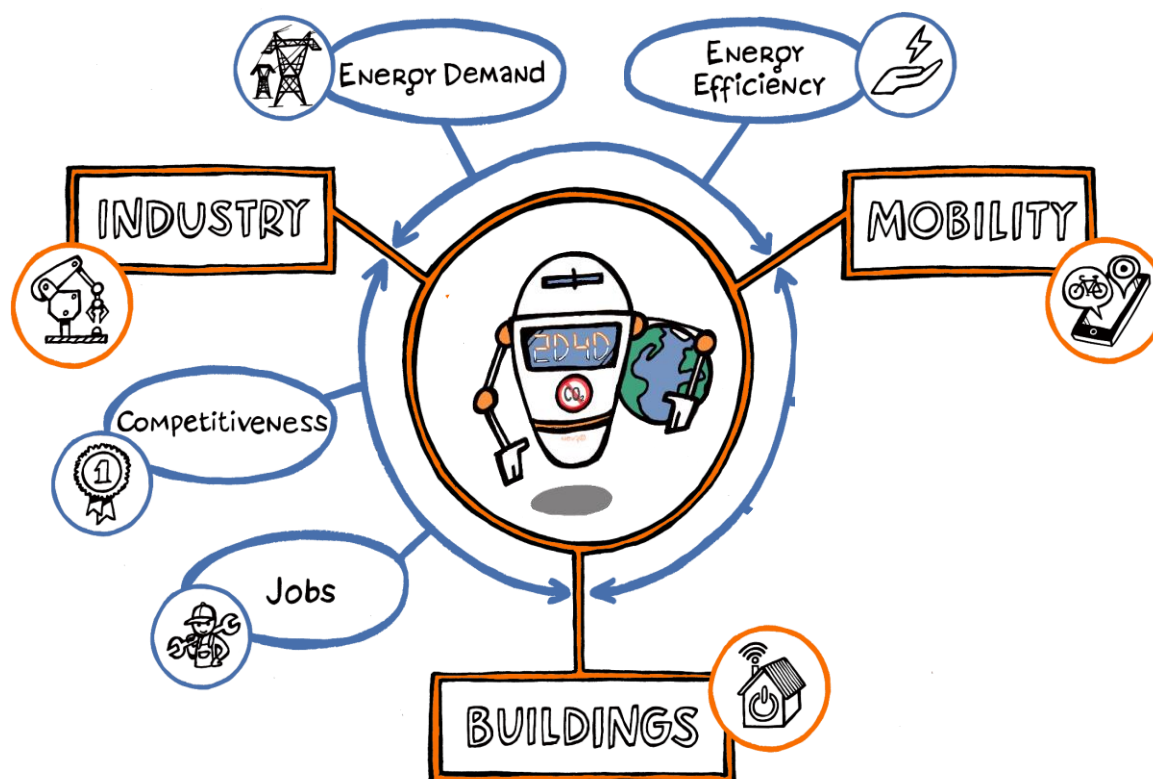
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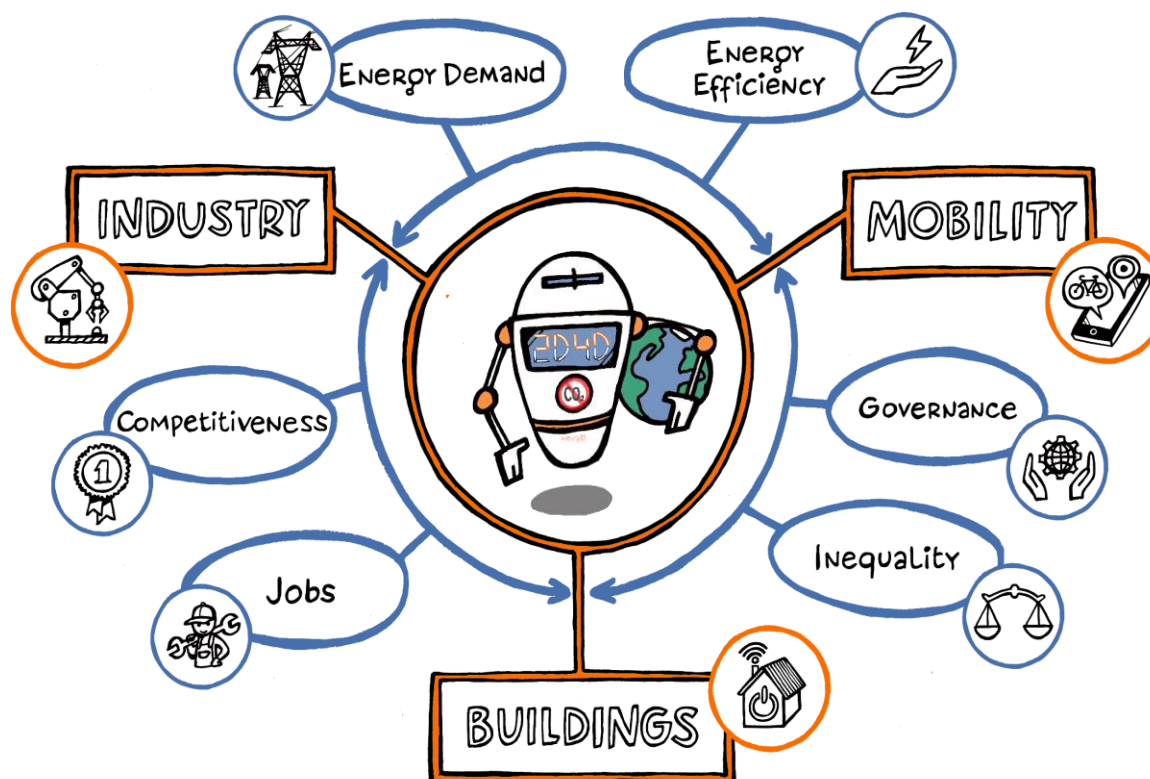
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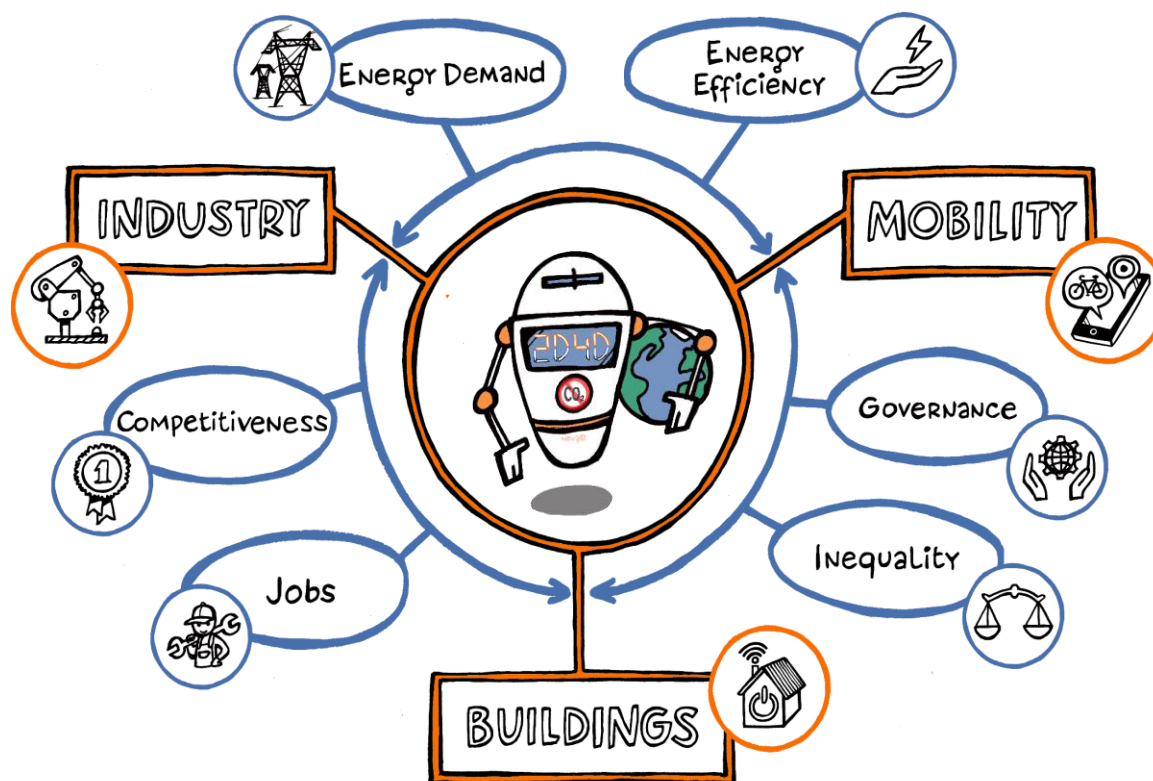
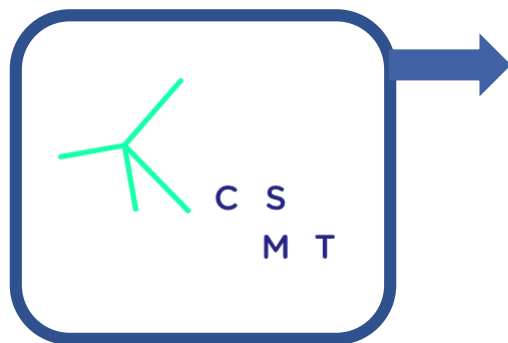
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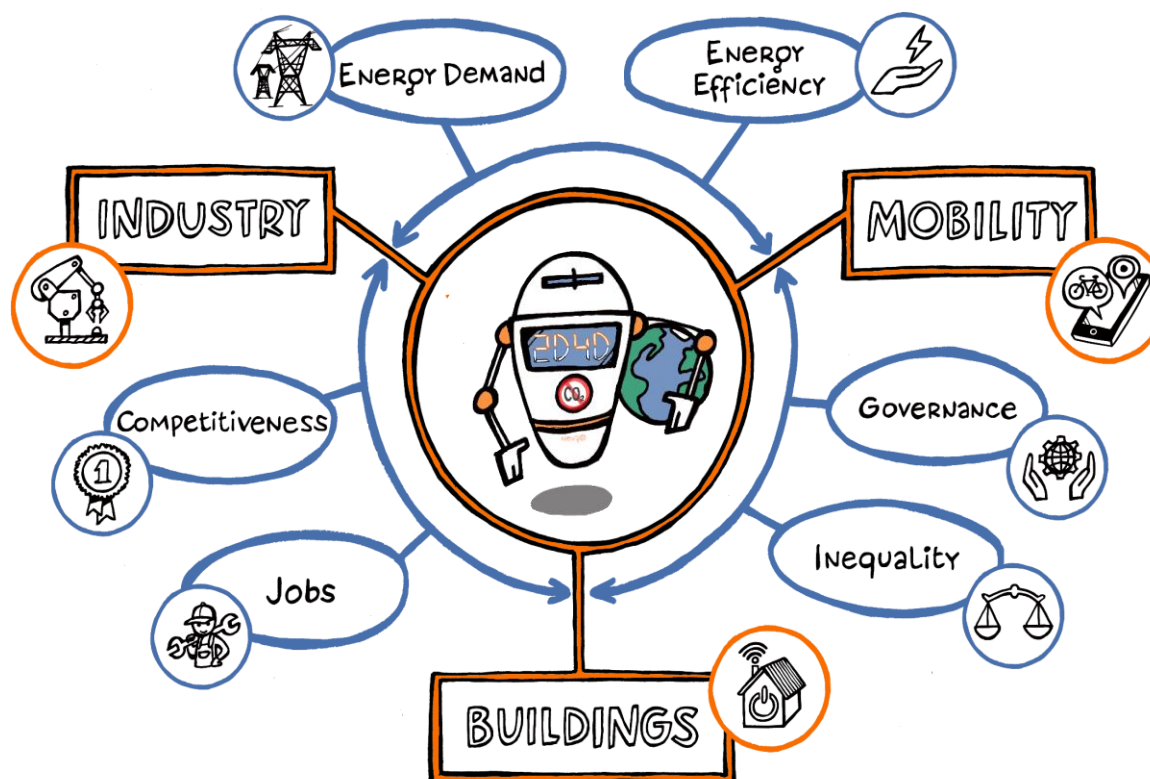
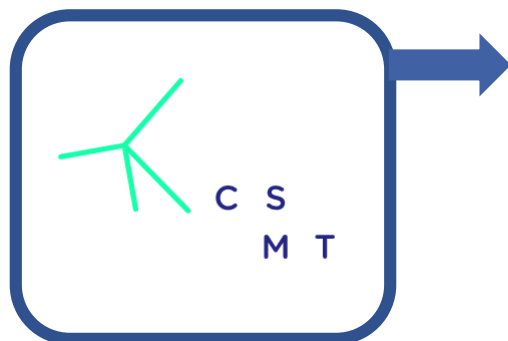
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Transizione energetica



Legislazione e policy

Domanda di
competenze nella forza
lavoro

Impatto di innovazioni
su competitività



Tecnologie digitali

Emissioni



Impatti socio-economici



Transizione energetica

Normativa per il riciclo batterie al litio

Legislazione e policy

Tassonomia di tecnologie digitali ad alto potenziale di mitigazione nel settore industriale e analisi del loro impatto su competenze richieste dall'impresa

Domanda di competenze nella forza lavoro

Casi studio e best practices

Impatto di innovazioni su competitività



Tecnologie digitali

Emissioni



Impatti socio-economici

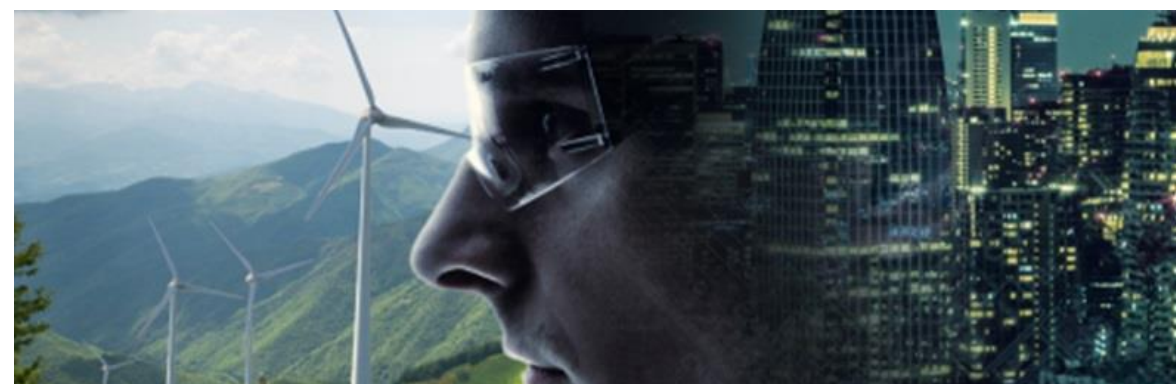
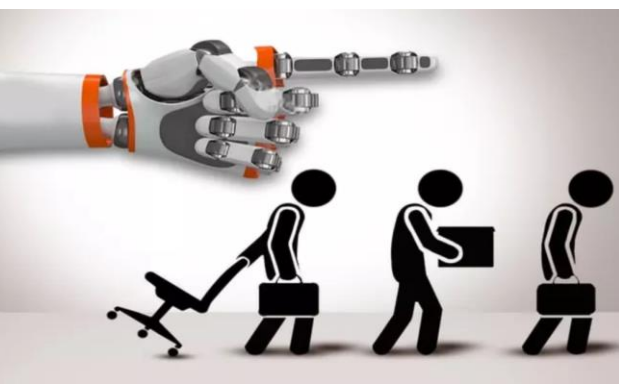


Transizione energetica

Effetto dislocamento, effetto produttività
ed effetto reintegrazione

Evidenza scientifica esiste solo per robotizzazione. Impatti di altre tecnologie digitali non sono noti

Domanda di
competenze nella forza
lavoro





Tecnologie digitali

Emissioni



Impatti socio-economici



Transizione energetica

Technology	Labor market effects
Industrial Robots	Negative ↓ (Acemoglu and Restrepo, 2020): Robots and humans compete. One robot/'000 workers reduces empl/pop ratio by 0.2 pp and wages by 0.42%; (Grigoli et al., 2020): Routinization reduces labour force participation among prime-age men and women No effect ⇌; ∠ (Graetz and Michaels, 2018): 0.36 pp annual labor productivity growth (higher TFP, lower output price); No implications for aggregate hours worked; Reduction of hours worked for low skill (relative to middle and high) Positive ↑; ∠ (Koch et al., 2021): Net job creation (10% rate). Note: gains for all worker types, including low-skilled and manufacturing. However, negative effects on non-adopting firms (empl., output, survival) Δ skills (Dauth et al., 2021): Job reduction in manufacturing (new entrants) but offset by new jobs in services. Incumbents have more stable employment and change tasks (higher quality jobs). Beneficial for complementary high skill occupations (managers, technicians). Young workers adapt education (less vocational training vs university and college).
Additive Manufacturing	Positive, non causal + (Felice et al., 2021): AM adoption and employment are positively associated at the industry level; Heterogeneous mechanisms across industries Δ skills, non causal (construction sector) (Hossain et al., 2020): AM would imply changes in the skill composition, from manual construction work to 3D programming and designing skills (Highlights from literature review) + Complement to labour productivity; higher workers' safety

Industrial Internet of Things (IIoT)

Artificial Intelligence (AI) and Big Data Analytics

∅
Negative ↓ (Acemoglu et al., 2020) Evidence of negative hiring effects (establishment level), but modest and non-robust effect sizes. No stat. signif. impact at occupation or industry level. No significant effect on changes in skills. Classification: Exposure (replacement risk) Frey and Osborne 2017: 47% of employment in the U.S.; (Weller et al., 2019): 24% in Latin America ∠ (Webb, 2019): Most exposed: high skill occupations (e.g., clinical laboratory technicians, chemical engineers, and power plant operators); smaller proportion of low skill jobs (inspection and quality control) Least exposed: occupations with high reasoning skills (e.g.researchers); interpersonal skill (e.g. managers, instructors, chefs, therapists) Brynjolfsson et al. 2018: occupations exposed to ML in different degrees. No full replacement

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