



Chimera: Improving Generalist Model with Domain-Specific Experts

Supplementary Material

Due to the eight-page limitation of the main text, we provide more details and visualizations from the following aspects:

- Sec. 1: Selection strategy for pre-training tasks and expert models.
- Sec. 2: Dataset Details in training.
- Sec. 3: Chimera’s performance on general tasks.
- Sec. 4: Details about preference optimization.
- Sec. 5: Experiment results on mask ratio selection.
- Sec. 6: Introduction of Table-SE and Doc-SE.
- Sec. 7: Experiments of scaling up more experts.
- Sec. 8: Comparison with existing works.
- Sec. 9: More information of implementation details.
- Sec. 10: Visualization of Chimera’s visual content extraction performance.

1. Pre-training Tasks and Expert Models

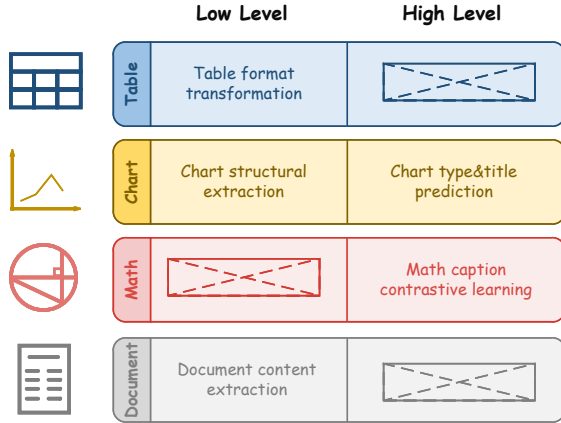


Figure 1. Pre-training tasks of expert models considered by Chimera.

The type of pre-training task significantly affects model performance, which we consider when selecting expert models. As shown in Fig. 1, we categorize low-level tasks as the precise extraction of domain-specific visual content and structure (e.g., Table2LaTeX, Chart2Markdown, Doc2Markdown), while high-level tasks involve understanding and summarizing image content. We select expert models with diverse pre-training task configurations. For the table expert, we use the encoder from StructE-qTable [39], which effectively converts table images into LaTeX/HTML. For the chart expert, we choose the encoder from ChartVLM [40], which excels in structural extraction and chart type classification. For the math expert, we adopt Math-CLIP [43], trained on extensive geometry and func-

Stage 1	General: ShareGPT4v [4], ShareGPT4-o [4]
	Table: TableX [39]
	Chart: ChartQA [27], PlotQA [30], ChartX [40], SimChart [38]
	Math: MAVIS-Caption [43]
Stage 2:	Language: Kaggle-science-exam [21], MathInstruct [41], MathQA [1], SciInstruct [42], Orcamath [31]
	General: ShareGPT4v [4], ShareGPT4-o [4], LLaVAR [44], AI2D (GPT4V) [18], AI2D (InternVL [5]), AI2D (Original) [16], MathVision [36], IconQA [24], MapQA [2], ScienceQA [32], ArxivQA [19], TQA [17], CLEVR-Math [10], Super-CLEVR [20], Cambrian Data Engine [35]
	Table: TableX [39], TabMWP [26], MMTab [45]
	Chart: PlotQA [30], ChartX [40], SimChart [38], Chart2Text [14], ChartQA [27], LRV Chart [22], ChartGemma [28], DVQA [12], FigureQA [13], VisText [34]
	Math: MAVIS-Caption [43], Geo170K [7], GeoMVerse [15], MAVIS Manual Collection [43], MAVIS Data Engine [43], Geometry3K [23], GeoQA+ [3], InterGPS [23]

Table 1. Dataset used for multi-modal reasoning scenario. **Stage 1** and **Stage 2** represent Domain-General Knowledge Alignment and Visual Instruction Tuning separately.

Stage 1	ChartQA [27], PlotQA [30], ChartX [40], SimChart [38], TableX [39]
Stage 2	DocGenome [39], DocStruct4M [9], DocVQA [29]

Table 2. Datasets used for visual content extraction scenario. **Stage 1** represents Domain-General Knowledge Alignment, and **Stage 2** represents Visual Instruction Tuning.

tion caption data. For document structural extraction, we employ the encoder from the latest model, GOT [37].

2. Dataset Details

The datasets used for Chimera is presented in Tab. 1 and Tab. 2. All the datasets we used come from publicly accessible datasets.

3. Evaluation on General Tasks

We evaluate Chimera’s general capabilities using the perception set from the general benchmark MME [6], with results presented in Tab. 3. Across different model sizes, Chimera and InternVL exhibit varying strengths across different tasks, achieving overall comparable performance.

	InternVL2		Chimera	
	4B	8B	4B	8B
Existence	200.00	190.00	200.00	195.00
Count	123.33	158.33	130.00	155.00
Position	143.33	155.00	123.33	148.33
Color	165.00	175.00	160.00	190.00
Posters	158.84	168.03	159.86	164.97
Celebrity	125.00	148.53	145.29	162.65
Scene	158.75	152.50	163.50	157.75
Landmark	167.25	178.25	167.25	177.75
Artwork	144.75	154.50	144.00	153.00
OCR	147.50	162.50	117.50	132.50

Table 3. Performance on perception sub-tasks of MME.

This suggests that the Chimera framework introduces minimal degradation to the model’s general task capabilities. Meanwhile, Chimera demonstrates strong expertise in domains such as tables, math, charts, and documents, further validating that our proposed approach effectively enhances a generalist LMM’s domain-specific knowledge without compromising its general performance.

4. Details about preference optimization

For preference optimization, we adopt a commonly used approach: we randomly sample 10k problems from MathV-360K, generating 16 responses per problem using Chimera. Each response is classified based on correctness using rule-based answer matching, and after filtering, we construct 60k preference pairs for Direct Preference Optimization (DPO) training. Then we perform DPO training on 60K data for 1 epoch.

5. Mask Ratio Selection

Model	Ratio	ALL	General	Chart	Table	Math
InternVL2-4B [5]	N/A	57.0	50.1	66.2	65.7	58.3
InternVL2-4B-NF [5]	N/A	58.5	51.5	67.1	74.3	58.6
Chimera-4B-0.0	0.0	59.4	50.8	66.2	67.1	65.5
Chimera-4B	0.3	61.3	54.0	64.8	72.9	66.9
Chimera-4B-0.5	0.5	60.4	51.3	68.5	70.0	65.8
Chimera-4B-1.0	1.0	56.2	51.5	63.5	72.9	53.6

Table 4. Ablation results on different visual content domain on the *testmini* subset of MathVista. InternVL2-4B-NF represents naive finetune of baseline with same settings, Chimera-4B-*R* means Chimera model trained with mask ratio *R* in GSCM.

We conducted an ablation study on 4B scale models to assess our approach’s effectiveness, as shown in Table 4. It should be noted that model with mask ratio 1.0 does not have access to the general encoder during training, contrary to our intentions. Thus, we modified this case to give the model an 80% probability of masking all general features. The results show that naively finetuning the LMM leads to limited performance improvement. By incorporat-

ing domain knowledge from expert models, even the case without GSCM still yields better results than naive finetuning. As the mask ratio increases, the model’s performance improves initially and then declines. This indicates that slightly masking helps balance encoder optimization, leading to better alignment. However, as the mask ratio increase, we believe excessive masking prevents the model from effectively learning to utilize both features for reasoning. Based on the above observations, we set the mask ratio to 0.3 in Chimera’s implementation. We also observed that performance trends vary across domains as the mask ratio changes, suggesting that the alignment difficulty of expert models differs by domain and task, which we leave for future exploration.

6. Details of Table-SE and Doc-SE

In Tables 7 and 8 of the main text, we conduct the experiments on Table Structural Extraction (Table-SE) task and Document Structural Extraction (Doc-SE) task, respectively. In this section, we primarily introduce the evaluation dataset construction method and provide detailed information about the dataset.

6.1. Data Source

	Count
Document Categories	
PPT2PDF	43
Academic Literature	42
Book	13
Colorful Textbook	37
Magazine	30
Exam Paper	7
Note	18
Newspaper	15
Language	
Simplified Chinese	128
English	77
Layout	
1 and More Column	27
Single Column	91
Other Layout	43
Double Column	40
Three Column	4
# Total	205

Table 5. Statistical information of Doc-SE.

Our benchmark was developed through a systematic sampling process from an initial collection of 200,000 PDF documents sourced from Common Crawl, Google, Baidu search engines, and internal repositories. We initially extracted visual features using ResNet-50 [8] and performed clustering algorithm using Faiss [11] to identify diverse document patterns. From the 10 cluster centers, we sampled 6,000 visually diverse pages, which were then man-

	Count
Background	
w/o Background	80
w/ Background	20
Equation	
w/o Equation	78
w/ Equation	22
Language	
English	45
English & Chinese Mixed	5
Chinese	50
Table Format	
Three-line Table	47
Full-bordered Table	39
Partial-bordered Table	14
w/o Merged Cells	58
w/ Merged Cells	42
Layout	
Horizontal	97
Vertical	3
# Total	100

Table 6. Statistical information of Table-SE.

ually annotated with attributes such as page type, layout type, and language. As illustrated in Table 5 and Table 6, the final benchmark includes 205 page-level PDF images and 100 table images, ensuring comprehensive representation of real-world document scenarios with various layouts and attributes.

6.2. Annotation Process

For ensuring annotation quality and efficiency, we design separate standardized processes for page-level PDF documents and tables.

For page-level PDF documents, our process consists of three stages: (1) We first employ fine-tuned LayoutLMv3 for layout detection and PaddleOCR for text recognition as intelligent pre-annotation. (2) Professional annotators then refine the detection boxes, verify text content accuracy, and enhance annotations with reading order and affiliation details. (3) Finally, researchers review the annotations to ensure overall quality and accuracy.

For table annotations, we follow a similar but specialized three-stage approach: (1) We utilize GPT-4o and PaddleOCR for initial table annotations. (2) Annotators then verify and correct the table structure and content, using specialized tools like Tables Generator for verification. (3) Finally, experts through table annotations re-rendering to ensure correct HTML and LaTeX code labels.

6.3. Showcase

We provide several visualization examples of Table-SE in Fig. 2 and Fig. 3, where each item contains a visual table and its corresponding LaTeX code.

7. Experiments of Scaling Up More Experts

Model	ALL	General	Chart	Table	Math
InternVL2-4B	57.0	50.1	66.2	65.7	58.3
InternVL2-4B w/ Chart Expert	59.4	52.0	68.0	72.9	60.8
Chimera-4B	61.3	54.0	64.8	72.9	66.9

Table 7. Accuracy scores of different visual content domain on the *testmini* subset of MathVista. Those do not belong to the last three domains are uniformly classified as General for simplicity. InternVL2-4B w/ Chart Expert represent the case only integrating chart expert model.

To further validate the impact of scaling up the number of expert models, we provide ablation results introducing only the chart expert. In this case, non-chart data is encoded solely by the general encoder during training. As shown in Table 7, incorporating only the chart expert obtains lower MathVista [25] overall score by 1.9 points than Chimera-4B.

Specifically, InternVL2-4B w/ Chart Expert also shows improvements in general scenarios, though less significant than Chimera, which integrates three expert models. In the chart domain, InternVL2-4B w/ Chart Expert achieves notable gains by avoiding conflicts among multiple experts with large task gaps. However, Chimera’s integration of multiple experts enhances performance across diverse domains, boosting overall results. In the math domain, InternVL2-4B w/ Chart Expert scores 6.1 points lower than Chimera, demonstrating the strong mathematical reasoning capabilities derived from integrating the math expert.

8. Comparison with existing works

Integrating specialist experts that contain specialized prior knowledge presents a promising approach to improve the specific capabilities of generalist model. MoVA [46] proposes a multi-turn method that relies on the language model to call an expert in the first round and generates responses in the second, which reduces conciseness and efficiency. MoME [33] uses soft-weighting to fuse multiple visual encoders, enabling VLMs to benefit from leveraging representations from different encoders. However, this approach lacks explicit guidance for encoder selection and introduces additional concerns such as inference efficiency and uniform visual feature sizes.

9. Training Configuration

The training strategy is summarized in Tab 8 and Tab 9. During the two-stage training process, we gradually increase the maximum image resolution and the number of visual tokens of the general visual encoder E^g . In the Domain-General Knowledge Alignment stage, we use

Visual Table

	Beef meat				Chicken meat				Pork meat			
	Cases	Controls	OR	95% CI	Cases	Controls	OR	95% CI	Cases	Controls	OR	95% CI
Never Exposed (Ref group)	1,823	2,273	1.00		1,823	2,273	1.00		1,823	2,273	1.00	
Ever Exposed	117	108	1.22	0.90-1.67	136	129	1.19	0.91-1.55	145	143	1.09	0.83-1.42
Duration of exposure												
<5 years	40	37	1.45	0.92-2.31	30	40	0.97	0.60-1.58	39	41	1.25	0.80-1.96
6-15 years	29	43	0.79	0.47-1.31	42	41	1.21	0.78-1.88	44	58	0.84	0.55-1.28
>16 years	48	28	1.63	0.95-2.88	64	48	1.36	0.90-2.06	61	43	1.28	0.81-2.03
p-value of test for linear trend (with ref cat)			0.23				0.11				0.54	
Intensity of exposure												
Low	60	59	1.26	0.86-1.83	71	68	1.24	0.88-1.75	70	72	1.15	0.82-1.62
Medium	42	35	1.22	0.73-2.04	47	46	1.11	0.72-1.71	55	52	1.03	0.68-1.58
High	15	14	0.91	0.35-2.40	18	15	1.22	0.56-2.65	20	19	0.89	0.40-1.94
p-value of test for linear trend (with ref cat)			0.36				0.29				0.78	

AED	MES test		Pentylenetetrazol		Bicuculline		Picrotoxin		Strychnine	
	ED ₅₀ (95% CI)	PI	ED ₅₀ (95% CI)	PI	ED ₅₀ (95% CI)	PI	ED ₅₀ (95% CI)	PI	ED ₅₀ (95% CI)	PI
Rofenofen	15.1 (12.5-18.1)	>32.2	54.0 (37.8-72.1)	>9.3	76.3 (54.8-97.7)	>9.3	10.1 (7.0-13.2)	>9.3	10.1 (7.0-13.2)	>9.3
AED	600 (515-72.1)	9.5 (8-10.4)	4.9 (3.9-6.0)	300 (230-390)	1.8 (1.5-2.1)	1.8 (1.5-2.1)	1.8 (1.5-2.1)	1.8 (1.5-2.1)	1.8 (1.5-2.1)	1.8 (1.5-2.1)
Phenytoin	6.9 (5.8-7.9)	3.2	12.2 (8.5-15.9)	3.2	37.7 (26.5-48.9)	1.8	37.7 (26.5-48.9)	2.5	35-100 ^a	NA
Phenobarbital	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Ethosuximide	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Valproic acid	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Carbamazepine	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Topiramate	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Levetiracetam	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Primidone	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
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Clonazepam	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Valproic acid	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Carbamazepine	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Topiramate	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
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Topiramate	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Levetiracetam	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	18.6 (16.1-21.1)	3.4	250 (191.3-365.0)	0.7
Primidone	44.8 (38.1-50.3)	1.0	18.6 (16.1-21.1							

ED₅₀, the dose of drug required to produce the desired end point in 50% of animals; CI, confidence interval; PI, protective index (ratio of TD₅₀ to ED₅₀).

Ground Truth: LaTeX

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\begin{tabular}{c}
\hline
\multicolumn{4}{c}{Beef meat} & \multicolumn{4}{c}{Chicken meat} & \multicolumn{4}{c}{Pork meat} \\
\hline
\hline
Cases & Controls & OR & 95\% CI & Cases & Controls & OR & 95\% CI & Cases & Controls & OR & 95\% CI \\
\hline
Never Exposed (Ref group) & 1,823 & 2,273 & 1.00 & --- & 1,823 & 2,273 & 1.00 & --- & 1,823 & 2,273 & 1.00 & --- \\
Ever Exposed & 117 & 108 & 1.22 & 0.90-1.67 & 136 & 129 & 1.19 & 0.91-1.55 & 145 & 143 & 1.09 & 0.83-1.42 \\
Duration of exposure & & & & & & & & & & & & \\
<5 years & 40 & 37 & 1.45 & 0.92-2.31 & 30 & 40 & 0.97 & 0.60-1.58 & 39 & 41 & 1.25 & 0.80-1.96 \\
6-15 years & 29 & 43 & 0.79 & 0.47-1.31 & 42 & 41 & 1.21 & 0.78-1.88 & 44 & 58 & 0.84 & 0.55-1.28 \\
>16 years & 48 & 28 & 1.63 & 0.95-2.88 & 64 & 48 & 1.36 & 0.90-2.06 & 61 & 43 & 1.28 & 0.81-2.03 \\
p-value of test for linear trend (with ref cat) & & & 0.23 & & & & 0.11 & & & & & 0.54 \\
Intensity of exposure & & & & & & & & & & & & \\
Low & 60 & 59 & 1.26 & 0.86-1.83 & 71 & 68 & 1.24 & 0.88-1.75 & 70 & 72 & 1.15 & 0.82-1.62 \\
Medium & 42 & 35 & 1.22 & 0.73-2.04 & 47 & 46 & 1.11 & 0.72-1.71 & 55 & 52 & 1.03 & 0.68-1.58 \\
High & 15 & 14 & 0.91 & 0.35-2.40 & 18 & 15 & 1.22 & 0.56-2.65 & 20 & 19 & 0.89 & 0.40-1.94 \\
p-value of test for linear trend (with ref cat) & & & 0.36 & & & & 0.29 & & & & & 0.78
\end{tabular}

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\begin{tabular}{c}
\hline
\multicolumn{12}{c}{Table 1. Anticonvulsant activity and protective index of intraperitoneal AEDs in mice} \\
\hline
\hline
\multicolumn{12}{c}{MES test} \\
\hline
\multicolumn{12}{c}{Phenytoin} \\
\hline
\multicolumn{12}{c}{Bicuculline} \\
\hline
\multicolumn{12}{c}{Strychnine} \\
\hline
\multicolumn{12}{c}{ED50 (95\% CI) mg/kg & ED50 (95\%CI) mg/kg & PI & ED50 (95\%CI) mg/kg & PI & ED50 (95\%CI) mg/kg & PI & ED50 (95\%CI) mg/kg & PI & ED50 (95\%CI) mg/kg & PI} \\
\hline
\multicolumn{12}{c}{Rufinamide & >500<1,000 & 15.5(12.5-18.1) & >32.2 & 54.0(38.1-74.9) & >9.3 & 50.5(23.9-87.8) & >9.9 & 76.3 (64.0-90.7) & >6.6 & 125\textit{textasciicircum}{a} & NA} \\
\hline
\multicolumn{12}{c}{Phenytoin & 65.5(52.5-72.1) & 9.5(8.1-10.4) & 6.9 & 300 no protection & <0.2 & 100 no protection & <0.7 & 100 no protection & <0.7 & 55-100b & <0.7} \\
\hline
\multicolumn{12}{c}{Phenobarbital & 69.0(62.8-72.9) & 21.8(15.0-25.5) & 3.2 & 13.2(5.9-15.9) & 5.2 & 37.7(26.5-47.4) & 1.8 & 27.5 (20.9-34.8) & 2.5 & 95.3 (91.3-99.5) & 0.7} \\
\hline
\multicolumn{12}{c}{Valproate & 425.8(369-450) & 272(247-338) & 1.6 & 148.6(123-177) & 2.9 & 359.9(294-439) & 1.2 & 387.2 (341-444) & 1.1 & 292.9 (261-323) & 1.5} \\
\hline
\multicolumn{12}{c}{Ethosuximide & 440.8(383-485) & 1,000 no protection & 0.4 & 130.3(111-151) & 3.4 & 459.0(350-633) & 1.0 & 243 (228-255) & 1.8 & 250-1,000c & <0.4} \\
\hline
\multicolumn{12}{c}{\{a: Maximum protection, 37.5\%\}} \\
\multicolumn{12}{c}{\{b: protection, 50.0\%\}} \\
\multicolumn{12}{c}{\{c: Maximum protection, 62.5\%\}} \\
\multicolumn{12}{c}{AED, antiepileptic drug; MES, maximal electroshock; TD50, the dose eliciting evidence of minimal neurotoxicity in 50\% of animals; CI, confidence interval; ED50, the dose of drug required to produce the desired end point in 50\% of animals; and PI, protective index (ratio of TD50 to ED50)} \\
\hline
\end{tabular}

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Figure 2. Showcase of Table-SE.

Visual Table

企业类型	目的	模式和特点	优势	典型企业
云服务提供商	以物联网为抓手带动上层应用服务业绩增长	目前多以提供底层计算资源、提供应用使能平台为主	在互联网领域中积累了丰富的技术、商业、生态优势经验 底层IaaS能力突出、共性技术能力提炼	阿里巴巴、腾讯云、百度云、亚马逊AWS IoT等
通信领域厂商	获得流量业务收入、战略布局物联网、把握新增市场机遇	多以连接管理、应用使能平台为主要功能服务为主	在连接管理平台具有绝对优势、具有全球通用连接能力	电信运营商、通信设备厂商、中国移动ONENet、中国联通物联网平台、华为云IoT等
软件系统服务商	解决内部开发效率的问题、优化产品服务	以应用开发平台为主要服务内容为主	擅长软件设计、生产、管理、运维等服务、具备丰富的行业软件开发及服务经验	紫光云、广联达筑联等
垂直领域传统厂商	利用自身对行业的理解与经验、打造垂直平台、实现传统企业的转型升级	垂直专业领域的物联网平台	深刻的行业理解和行业技术、对行业有深度应用、拥有行业数据和客户资源	西门子、工业富联、美的M-Smart等企业
初创企业	看好物联网未来的发展潜力	目前阶段很多初创平台企业多以SaaS解决方案公司的形式存在	拥有与选定细分行业相关的软件、硬件经验 服务延伸到通用型平台厂商难以触及的细分领域、形成错位竞争	涂鸦智能、云智易、机智云、艾拉物联等

Site-B	Site-E	DEV	DD	Model
3-HCY	11-DGG	1.49	66.5	d-1
3-HCY	12-GGS	1.49	66.5	
4-CNY	1-GGH	1.29	74.8	
4-CNY	11-DGG	1.29	74.8	
4-CNY	12-GGS	1.29	74.8	d-2
5-YNT	1-GGH	1.39	70.3	
5-YNT	11-DGG	1.39	70.3	
5-YNT	12-GGS	1.39	70.3	
5-YNT	18-GGC	1.39	76.6	d-3
9-NND	1-GGH	1.85	69.2	
9-NND	12-GGS	1.85	69.2	
9-NND	18-GGC	1.85	73.9	

Ground Truth: LaTeX

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\begin{tabular}{|l|l|l|l|l|} \hline
企业类型 & 目的 & 模式和特点 & 优势 & 典型企业 \\ \hline
云服务提供商 & 以物联网为抓手带动上层应用服务业绩增长 & 目前多以提供底层计算资源、提供应用使能平台为主 & 在互联网领域中积累了丰富的技术、商业、生态优势经验 & 阿里巴巴、腾讯云、百度云、亚马逊AWS IoT等 \\ \hline
通信领域厂商 & 获得流量业务收入、战略布局物联网、把握新增市场机遇 & 多以连接管理、应用使能平台为主要功能服务为主 & 在连接管理平台具有绝对优势、具有全球通用连接能力 & 电信运营商、通信设备厂商、中国移动ONENet、中国联通物联网平台、华为云IoT等 \\ \hline
软件系统服务商 & 解决内部开发效率的问题、优化产品服务 & 以应用开发平台为主要服务内容为主 & 擅长软件设计、生产、管理、运维等服务、具备丰富的行业软件开发及服务经验 & 紫光云、广联达筑联等 \\ \hline
垂直领域传统厂商 & 利用自身对行业的理解与经验、打造垂直平台、实现传统企业的转型升级 & 垂直专业领域的物联网平台 & 深刻的行业理解和行业技术、对行业有深度应用、拥有行业数据和客户资源 & 西门子、工业富联、美的M-Smart等企业 \\ \hline
初创企业 & 看好物联网未来的发展潜力 & 目前阶段很多初创平台企业多以SaaS解决方案公司的形式存在 & 拥有与选定细分行业相关的软件、硬件经验 & 涂鸦智能、云智易、机智云、艾拉物联等 \\ \hline
\end{tabular}

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\begin{tabular}{|l|l|l|l|l|} \hline
Site-B & Site-E & DEV & DD & Model \\ \hline
3-HCY & 11-DGG & 1.49 & 66.5 & \\ \hline
3-HCY & 12-GGS & 1.49 & 66.5 & d-1 \\ \hline
4-CNY & 1-GGH & 1.29 & 74.8 & \\ \hline
4-CNY & 11-DGG & 1.29 & 74.8 & \\ \hline
4-CNY & 12-GGS & 1.29 & 74.8 & \\ \hline
5-YNT & 1-GGH & 1.39 & 70.3 & d-2 \\ \hline
5-YNT & 11-DGG & 1.39 & 70.3 & \\ \hline
5-YNT & 12-GGS & 1.39 & 70.3 & \\ \hline
5-YNT & 18-GGC & 1.39 & 76.6 & \\ \hline
9-NND & 1-GGH & 1.85 & 69.2 & d-3 \\ \hline
9-NND & 12-GGS & 1.85 & 69.2 & \\ \hline
9-NND & 18-GGC & 1.85 & 73.9 & \\ \hline
\end{tabular}

```

Figure 3. Showcase of Table-SE.

thumbnail images as input for E^g without employing the widely-used Dynamic High Resolution (DHR) technique [5, 18]. In the Visual Instruction Tuning stage, DHR is introduced, allowing up to six times more visual tokens. At this stage, we apply the Generalist-Specialist Collaboration Masking (GSCM) mechanism with a masking ratio of 0.3 to constrain E^g , encouraging the model to leverage domain-specific information from expert models. For train-

able modules, the Domain-General Knowledge Alignment stage updates only the General Projector P^g and Expert Projector Set S^e . In subsequent stages, the General Projector P^g , Expert Projector Set S^e , and Language Model f are updated.

		Domain-General Knowledge Alignment	Visual Instruction Tuning	
Vision	Resolution	General Encoder E^g	448	$448 \times \{\{1,2,3,4,5,6\} \times 1, 1 \times \{2,3,4,5,6\}, 2 \times \{2,3\}, 3 \times 2\}\}$
		Table Encoder E^{table}	N/A	N/A
		Chart Encoder E^{chart}	N/A	N/A
		Math Encoder E^{math}	336	336
	#Tokens	General Encoder E^g	256	Max 256×6
		Table Encoder E^{table}	2048	2048
		Chart Encoder E^{chart}	2048	2048
		Math Encoder E^{math}	576	576
	Total Tokens		$256 + \{0, 2048, 2048, 576\}$	Max $256 \times 6 + \{0, 2048, 2048, 576\}$
	Training	#Samples	1.1M	2.6M
GSCM ratio		N/A	0.3	
Dynamic High Res [5]		False	True	
Trainable		General Projector P^g , Expert Projector Set S^e	General Projector P^g , Expert Projector Set S^e , Language Model f	
Batch Size		256/128	128	
LR		4e-5/2e-5	2e-5/1e-5	
Warm Up		100 steps	0.03 ratio	
LR Scheduler		Consine	Consine	
Max Length		4096	8192	
Weight Decay		0.01	0.01	
Epoch		1	1	

Table 8. Detailed configuration for each training stage of Chimera in multi-modal reasoning scenario. The table outlines the progression of vision parameters, dataset characteristics and training hyperparameters. For elements containing “/”, the left side represents configurations used by the 2B and 4B model, while the right side represents configurations used by the 8B model.

			Domain-General Knowledge Alignment	Visual Instruction Tuning
Vision	Resolution	General Encoder E^g	448	$448 \times \{\{1,2,3,4,5,6\} \times 1, 1 \times \{2,3,4,5,6\}, 2 \times \{2,3\}, 3 \times 2\}\}$
		Document Encoder $E^{document}$	1024	1024
	#Tokens	General Encoder E^g	256	Max 256×6
		Document Encoder $E^{document}$	256	256
	Total Tokens		$256 + 256$	Max $256 \times 6 + 256$
Training	#Samples	995K	275K	
	GSCM ratio	N/A	0.3	
	Dynamic High Res [5]	False	True	
	Trainable	General Projector P^g , Expert Projector Set S^e	General Projector P^g , Expert Projector Set S^e , Language Model f	
	Batch Size	256	128	
	LR	4e-5	2e-5	
	Warm Up	100 steps	0.03 ratio	
	LR Scheduler	Consine	Consine	
	Max Length	4096	8192	
	Weight Decay	0.01	0.01	
	Epoch	1	1	

Table 9. Detailed configuration for each training stage of Chimera in visual content extraction scenario. The table outlines the progression of vision parameters, dataset characteristics and training hyperparameters.

10. Visualization of Chimera on Visual Content Extraction

10.1. Table Format Transformation

We provide the rendered table of the output results of Chimera-8B to show its table format transformation performance. As shown in Fig. 4, Fig. 5 and Fig. 6, Chimera excels in extracting and formatting table content from both Arxiv-style and more diverse table layouts with high accuracy.

10.2. Chart Structural Extraction

We provide the rendered table of the output results of Chimera-8B to show its chart structural extraction performance. As shown in Fig. 7, Fig. 8 and Fig. ??, Chimera can identify and extract information from various types of charts, such as pie charts, line graphs, bar charts, etc., and output this information in a structured format accurately.

Input: Visual Table

资产负债表					
	2020	2021	2022E	2023E	2024E
货币资金	703.3	604.1	2,111.4	2,582.9	2,817.4
交易性金融资产	-	-	-	-	-
应收账款	1,066.7	1,307.4	1,507.8	1,720.8	2,801.0
应收票据	85.6	163.1	72.2	157.3	147.1
预付账款	56.4	54.9	89.0	86.7	116.8
存货	779.0	648.3	1,009.8	1,053.6	1,519.6
其他流动资产	670.3	448.3	414.6	511.1	274.8
可供出售金融资产	-	-	-	-	-
持有至到期投资	-	-	-	-	-
长期股权投资	60.7	36.9	36.9	36.9	36.9
投资性房地产	5.0	13.2	13.2	13.2	13.2
固定资产	445.5	502.8	459.0	415.2	371.5
在建工程	24.4	29.0	29.0	29.0	29.0
无形资产	107.8	98.8	86.7	74.6	62.5
其他非流动资产	1,274.9	1,705.5	1,498.7	1,490.2	1,290.6
资产总额	5,279.6	5,612.2	7,328.3	8,171.5	9,480.2
短期债务	736.6	809.5	1,460.2	2,054.8	1,482.0
应付账款	1,554.8	1,651.2	2,027.4	1,942.5	2,822.0
应付票据	109.3	124.5	23.3	153.6	60.8
其他流动负债	516.5	613.7	581.4	580.4	584.1
长期借款	240.4	60.1	422.8	189.0	-
其他非流动负债	53.4	111.7	65.8	76.9	84.8
负债总额	3,210.9	3,370.6	4,580.8	4,997.2	5,033.8
少数股东权益	208.0	111.5	118.1	129.1	142.4
股本	774.8	772.8	772.8	772.8	772.8
留存收益	1,290.0	1,662.1	2,059.7	2,557.2	3,134.9
股东权益	2,068.6	2,241.6	2,747.6	3,174.3	4,446.4

主要财务指标				
	2021A	2022E	2023E	2024E
营业收入	19515	23852	27760	31884
收入同比 (%)	45.1%	22.2%	16.4%	14.9%
归属母公司净利润	1150	1339	1671	2002
净利润同比 (%)	43.9%	16.4%	24.8%	19.8%
毛利率 (%)	12.6%	12.7%	13.6%	13.9%
ROE (%)	15.8%	15.9%	17.0%	17.4%
每股收益 (元)	2.19	2.52	3.15	3.77
P/E	24.45	14.59	11.69	9.76
P/B	3.91	2.33	1.99	1.70
EV/EBITDA	15.29	6.19	4.85	3.80

	RPCB多层板	软板+模组	HDI	IC载板
2022(E)	387.21	138.42	117.63	174.15
2023(F)	373.40	134.27	115.28	160.73
2024(F)	381.79	141.31	122.25	174.41
2025(F)	419.39	148.72	129.65	189.26
2026(F)	444.30	156.52	137.49	205.38
2027(F)	450.48	164.73	145.81	222.86
2022-2027 CAGR	3.1%	3.5%	4.4%	5.1%

Output: LaTeX

资产负债表					
	2020	2021	2022E	2023E	2024E
货币资金	703.3	604.1	2,111.4	2,582.9	2,817.4
交易性金融资产	-	-	-	-	-
应收账款	1,066.7	1,307.4	1,507.8	1,720.8	2,801.0
应收票据	85.6	163.1	72.2	157.3	147.1
预付账款	56.4	54.9	89.0	86.7	116.8
存货	779.0	648.3	1,009.8	1,053.6	1,519.6
其他流动资产	670.3	448.3	414.6	511.1	274.8
可供出售金融资产	-	-	-	-	-
持有至到期投资	-	-	-	-	-
长期股权投资	60.7	36.9	36.9	36.9	36.9
投资性房地产	5.0	13.2	13.2	13.2	13.2
固定资产	445.5	502.8	459.0	415.2	371.5
在建工程	24.4	29.0	29.0	29.0	29.0
无形资产	107.8	98.8	86.7	74.6	62.5
其他非流动资产	1,274.9	1,705.5	1,498.7	1,490.2	1,290.6
资产总额	5,279.6	5,612.2	7,328.3	8,171.5	9,480.2
短期负债	736.6	809.5	1,460.2	2,054.8	1,482.0
应付账款	1,554.8	1,651.2	2,027.4	1,942.5	2,822.0
应付票据	109.3	124.5	23.3	153.6	60.8
其他流动负债	516.5	613.7	581.4	580.4	584.1
长期借款	240.4	60.1	422.8	189.0	-
其他非流动负债	53.4	111.7	65.8	76.9	84.8
负债总额	3,210.9	3,370.6	4,580.8	4,997.2	5,033.8
少数股东权益	208.0	111.5	118.1	129.1	142.4
股本	774.8	772.8	772.8	772.8	772.8
留存收益	1,290.0	1,662.1	2,059.7	2,557.2	3,134.9
股东权益	2,068.6	2,241.6	2,747.6	3,174.3	4,446.4

主要财务指标				
	2021A	2022E	2023E	2024E
营业收入	19515	23852	27760	31884
收入同比增长 (%)	45.1%	22.2%	16.4%	14.9%
归属于母公司净利润	1150	1339	1671	2002
净利润同比增长 (%)	43.9%	16.4%	24.8%	19.8%
毛利率 (%)	12.6%	12.7%	13.6%	13.9%
ROE (%)	15.8%	15.9%	17.0%	17.4%
每股收益 (元)	2.19	2.52	3.15	3.77
P/E	24.45	14.59	11.69	9.76
P/B	3.91	2.33	1.99	1.70
EV/EBITDA	15.29	6.19	4.85	3.80

	RPCB 多层板	软板 + 模组	HDI	IC 载板
2022(E)	387.21	138.42	117.63	174.15
2023(F)	373.40	134.27	115.28	160.73
2024(F)	381.79	141.31	122.25	174.41
2025(F)	419.39	148.72	129.65	189.26
2026(F)	444.30	156.52	137.49	205.38
2027(F)	450.48	164.73	145.81	222.86
2022-2027 CAGR	3.1%	3.5%	4.4%	5.1%

Figure 4. Output of Chimera-8B on Table Format Transformation.

10.3. Document Context Extraction

We provide the rendered page of the output results of Chimera to show its document content extraction performance. As shown in Fig. 10, Fig. 11, Fig. 12 and Fig. 13, Chimera demonstrates exceptional content extraction ca-

pabilities on both single-column and double-column documents, effectively extracting structured information end-to-end from text-dense visual inputs.

Input: Visual Table

Star	Sp. type	P _{rot} (days)	P _{cyc} (range) (years)	lt ^c	P _{cyc} , Balinuas et al. (years)
simple cycles					
HD 3651	K0V ^a	44 ^b	11.6(9.57–13.7)-lt	yes	13.8±0.4
HD 4628	K2.5V ^a	41.6	8.94(8.30–9.5)	no	8.37±0.08
HD 10476	K1V ^a	33.7	9.8(9.01–9.85)	no	9.6±0.1
HD 16160	K3V ^a	57	12.1	no	13.2±0.2
HD 26965	K1V ^a	43 ^b	10.0(9.57–10.5)	no	10.1±0.1
HD 32147	K3+V ^a	39.3	10.6(9.85–11.3), lt	yes	11.1±0.2
HD 81809	G5V ^a	39.3	8.69(8.10–9.28)	no	8.17±0.08
HD 103995	K1V ^a	36.5	6.95(6.9–7.0), lt	yes	7.30±0.08
HD 160346	K2.5V ^a	35.3	7.35(7.2–7.5), lt	yes	7.00±0.08
HD 166620	K2V ^c	41.5	13.6(9.6–17.6)	no	15.8±0.3
HD 201091	K5V	37.1	6.95(6.7–7.2), lt	yes	7.3±0.1
HD 219834B	K2V	34.0	9.29(9.01–9.57)	no	10.0±0.2
Sum	G2V	27.275	11(9–14), 3.65(3.3–4.0) ^d	yes	
complicated cycles					
HD 1835	G2V ^a	7.84	7.6(7.3–7.9), 2.4(2.50–2.28), 3.97(4.85–3.09), lt	yes	9.1±0.3
HD 20630	G5V	9.08	5.32(5.36–5.27)	no	5.6±0.1
HD 76151	G3V ^a	15.2	5.32(6.07–4.56), lt	yes	2.52±0.02
HD 78366	G0V	9.7	13.45(12.6–14.3)-lt, 4.0(3.85–4.15)	yes	12.2±0.4, 5.9±0.1
HD 95735	M2V	54	3.90(3.75–4.04), 12.7(10.3–15.0)-lt	yes	
HD 100180	F9.5V ^a	14.6	13.2(16.6–9.85), 3.63	no	3.56±0.04, 12.9±0.5
HD 114710	G0V	12.9	16.6(17.6–13.7–18.6), 7.7(7.2–8.2), 5.25(5.1–5.4)	no	16.6±0.6, 9.6±0.3
HD 115404	K2.5V ^a	18.8	10.8(9.57–12.1), 5.08(5.83–4.33), 3.4(3.96–2.84) lt	yes	12.4±0.4
HD 131156A	G8V	6.25	3.7(3.6–3.8), 12, lt	yes	
HD 131156B	K4V	11.05	4.2(4.64–3.76), 2.28, lt	yes	
HD 149661	K0V ^a	21.3	4.0, 6.2(5.6–6.8), 12.25(11.3–13.2)-lt	yes	17.4±0.7, 4.00±0.04
HD 152391	G8+V ^a	11.4	9.9(12.6–7.17), 2.8(2.5–3.1)	no	10.9±0.2
HD 156026	K5V	29.2	4.33(3.95–4.71), 8.1, lt(≈20)	yes	21.0±0.9
HD 165341A	K0V ^a	18.9	11.4(9–13.8), 5.17, lt	yes	5.1±0.1
HD 165341B	K6V	–	–	yes	Var
HD 190406	G0V ^a	15.5	15.8(15.0–17.6)-lt, 8.35(7.69–9.01), 4.65(4.44–4.86), 2.33, 3.01	yes	2.60±0.02, 16.9±0.8
HD 201092	K7V	34.1	13.4(10.1–16.6)-lt, 4.45(4.71–4.18), (2.07:)	yes	11.7±0.4

Output: LaTeX

Star	Sp. type	P _{rot} (days)	P _{cyc} (range) (years)		lt ^c	P _{cyc} , Balinuas et al. (years)			
simple cycles									
HD 3651	K0V ^a	44 ^b	11.6(9.57–13.7)-lt		yes	13.8±0.4			
HD 4628	K2.5V ^a	41.6	8.94(8.30–9.5)		no	8.37±0.08			
HD 10476	K1V ^a	33.7	9.8(9.01–9.85)		no	9.6±0.1			
HD 16160	K3V ^a	57	12.1		no	13.2±0.2			
HD 26965	K1V ^a	43 ^b	10.0(9.57–10.5)		no	10.1±0.1			
HD 32147	K3+V ^a	39.3	10.6(9.85–11.3), lt		yes	11.1±0.2			
HD 81809	G5V ^a	39.3	8.69(8.10–9.28)		no	8.17±0.08			
HD 103995	K1V ^a	36.5	6.95(6.9–7.0), lt		yes	7.30±0.08			
HD 160346	K2.5V ^a	35.3	7.35(7.2–7.5), lt		yes	7.00±0.08			
HD 166620	K2V ^c	41.5	13.6(9.6–17.6)		no	15.8±0.3			
HD 201091	K5V	37.1	6.95(6.7–7.2), lt		yes	7.3±0.1			
HD 219834B	K2V	34.0	9.29(9.01–9.57)		no	10.0±0.2			
Sum	G2V	27.275	11(9–14), 3.65(3.3–4.0) ^d		yes				
complicated cycles									
HD 1835	G2V ^a	7.84	7.6(7.3–7.9), 2.4(2.50–2.28), 3.97(4.85–3.09), lt		yes	9.1±0.3			
HD 20630	G5V	9.08	5.32(5.36–5.27)		no	5.6±0.1			
HD 76151	G3V ^a	15.2	5.32(6.07–4.56), lt		yes	2.52±0.02			
HD 78366	G0V	9.7	13.45(12.6–14.3)-lt, 4.0(3.85–4.15)		yes	12.2±0.4, 5.9±0.1			
HD 95735	M2V	54	3.90(3.75–4.04), 12.7(10.3–15.0)-lt		yes				
HD 100180	F9.5V ^a	14.6	13.2(16.6–9.85), 3.63		no	3.56±0.04, 12.9±0.5			
HD 114710	G0V	12.9	16.6(17.6–13.7–18.6), 7.7(7.2–8.2), 5.25(5.1–5.4)		no	16.6±0.6, 9.6±0.3			
HD 115404	K2.5V ^a	18.8	10.8(9.57–12.1), 5.08(5.83–4.33), 3.4(3.96–2.84) lt		yes	12.4±0.4			
HD 131156A	G8V	6.25	3.7(3.6–3.8), 12, lt		yes				
HD 131156B	K4V	11.05	4.2(4.64–3.76), 2.28, lt		yes				
HD 149661	K0V ^a	21.3	4.0, 6.2(5.6–6.8), 12.25(11.3–13.2)-lt		yes	17.4±0.7, 4.00±0.04			
HD 152391	G8+V ^a	11.4	9.9(12.6–7.17), 2.8(2.5–3.1)		no	10.9±0.2			
HD 156026	K5V	29.2	4.33(3.95–4.71), 8.1, lt(approximately 20)		yes	21.0±0.9			
HD 165341A	K0V ^a	18.9	11.4(9–13.8), 5.17, lt		yes	5.1±0.1			
HD 165341B	K6V	—	—		yes	Var			
HD 190406	G0V ^a	15.5	15.8(15.0–17.6)-lt, 8.35(7.69–9.01), 4.65(4.44–4.86), 2.33, 3.01		yes	2.60±0.02, 16.9±0.8			
HD 201092	K7V	34.1	13.4(10.1–16.6)-lt, 4.45(4.71–4.18), (2.07:)		yes	11.7±0.4			
		Controls		Hodgkin lymphoma		Non-Hodgkin lymphoma			
				Cases	OR	95% CI	Cases	OR	95% CI
Never Exposed (reference group)		2,273		315	1.00	–	1,823	1.00	–
Ever Exposed		189		24	1.06	0.65–1.71	184	1.18	0.95–1.46
Duration of exposure									
<5 years		52		12	1.14	0.57–2.30	49	1.25	0.84–1.86
6–15 years		62		8	0.98	0.44–2.20	52	1.04	0.71–1.51
≥16 years		73		4	1.02	0.36–2.90	82	1.27	0.92–1.76
p-value of test for linear trend						0.90			0.13
Weighted duration of exposure									
<6 months		62		14	1.54	0.79–2.99	57	1.10	0.76–1.59
7 months to 1 year		35		3	0.60	0.17–2.13	40	1.39	0.87–2.20
>1 year		90		7	0.84	0.37–1.91	86	1.17	0.87–1.59
p-value of test for linear trend						0.75			0.13
Intensity of exposure									
Low		84		11	1.05	0.52–2.12	85	1.24	0.91–1.70
Medium		70		10	1.19	0.56–2.49	66	1.11	0.79–1.57
High		35		3	0.80	0.23–2.74	32	1.14	0.70–1.85

Figure 5. Output of Chimera-8B on Table Format Transformation.

Input: Visual Table

会计年度	2021A	2022A	2023E	2024E	2025E
营业收入(百万元)	11,339	11,008	15,123	16,828	18,396
YoY(%)	14.6	-2.9	37.4	11.3	9.3
净利润(百万元)	101	113	1,392	1,991	2,200
YoY(%)	-8.7	12.8	1126.8	43.0	10.5
毛利率(%)	34.2	33.1	41.1	47.6	47.7
EPS(摊薄/元)	0.09	0.11	1.30	1.86	2.06
ROE(%)	1.7	1.4	8.7	11.2	11.1
P/E(倍)	408.4	362.1	29.5	20.6	18.7
P/B(倍)	2.5	2.5	2.3	2.1	1.9
净利率(%)	0.9	1.0	9.2	11.8	12.0

8月排名	证券简称	占比(%)	7月排名	证券简称	占比(%)
1	中天科技	8.37%	1	中天科技	8.43%
2	中际旭创	5.07%	2	中际旭创	4.43%
3	新易盛	4.59%	3	烽火通信	4.36%
4	普天科技	4.02%	4	中国联通	4.24%
5	中国移动	3.84%	5	新易盛	4.15%
6	中国联通	3.68%	6	中兴通讯	3.82%
7	华测导航	3.58%	7	华测导航	3.60%
8	中兴通讯	3.50%	8	天孚通信	3.52%
9	亿联网络	3.03%	9	普天科技	3.41%
10	烽火通信	2.87%	10	中国移动	3.32%

公司名称	2023年11月累计保费收入 (亿元)	同比增速	当月保费收入 (亿元)	环比增速
人保财险	4727	6.8%	329	8.8%
车险	2543	5.5%	247	7.7%
非车险	2184	8.3%	82	12.1%
平安财险	2736	1.4%	243	3.4%
太保财险	1748	11.8%	138	9.7%
合计	9211	6.0%	709	7.0%

Output: LaTeX

会计年度	2021A	2022A	2023E	2024E	2025E
营业收入 (百万元)	11,339	11,008	15,123	16,828	18,396
YoY(%)	14.6	-2.9	37.4	11.3	9.3
净利润 (百万元)	101	113	1,392	1,991	2,200
YoY(%)	-8.7	12.8	1126.8	43.0	10.5
毛利率 (%)	34.2	33.1	41.1	47.6	47.7
EPS(摊薄/元)	0.09	0.11	1.30	1.86	2.06
ROE(%)	1.7	1.4	8.7	11.2	11.1
P/E(倍)	408.4	362.1	29.5	20.6	18.7
P/B(倍)	2.5	2.5	2.3	2.1	1.9
净利率 (%)	0.9	1.0	9.2	11.8	12.0

8月排名	证券简称	占比 (%)	7月排名	证券简称	占比 (%)
1	中天科技	8.37%	1	中天科技	8.43%
2	中际旭创	5.07%	2	中际旭创	4.43%
3	新易盛	4.59%	3	烽火通信	4.36%
4	普天科技	4.02%	4	中国联通	4.24%
5	中国移动	3.84%	5	新易盛	4.15%
6	中国联通	3.68%	6	中兴通讯	3.82%
7	华测导航	3.58%	7	华测导航	3.60%
8	中兴通讯	3.50%	8	天孚通信	3.52%
9	亿联网络	3.03%	9	普天科技	3.41%
10	烽火通信	2.87%	10	中国移动	3.32%

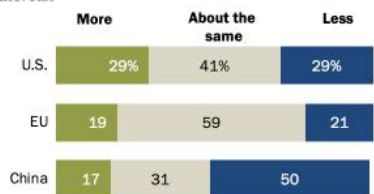
公司名称	2023年11月累计保费收入	同比增长	当月保费收入	环比增速
人保财险	4727	6.8%	329	8.8%
车险	2543	5.5%	247	7.7%
非车险	2184	8.3%	82	12.1%
平安财险	2736	1.4%	243	3.4%
太保财险	1748	11.8%	138	9.7%
合计	9211	6.0%	709	7.0%

Figure 6. Output of Chimera-8B on Table Format Transformation.

Input: Visual Chart

Many think China's global influence will decline after the coronavirus outbreak

% who say each will have ___ influence in world affairs after the coronavirus outbreak compared to before the outbreak



Note: No answer responses not shown.

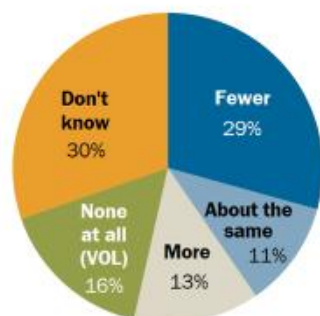
Source: Survey of U.S. adults conducted April 29-May 5, 2020.

*Americans Give Higher Ratings to South Korea and Germany Than U.S. for Dealing With Coronavirus

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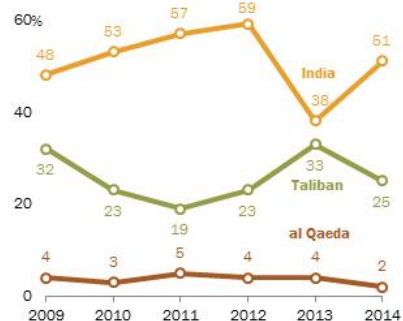
Little appetite for allowing more immigrants into India

India should allow ___ immigrants



Concerns about India on the Rise Again

Which is the greatest threat to Pakistan?



Source: Spring 2014 Global Attitudes survey. Q117.

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Output: Markdown

Entity	More	About the same	Less
China	17	31	50
EU	19	59	21
U.S.	29	41	29

Entity	Value
About the same	11
More	13
None at all (VOL)	16
Don't know	30
Fewer	29

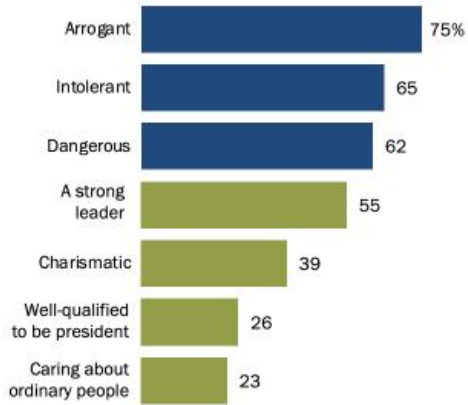
Entity	2009	2010	2011	2012	2013	2014
al Qaeda	4.0	3.0	5.0	4.0	4.0	2.0
Taliban	32.0	23.0	19.0	23.0	33.0	25.0
India	48.0	53.0	57.0	59.0	38.0	51.0

Figure 7. Output of Chimera-8B on Chart Structural Extraction.

Input: Visual Chart

Global views of Trump's characteristics

% who say they think of President Donald Trump as ...

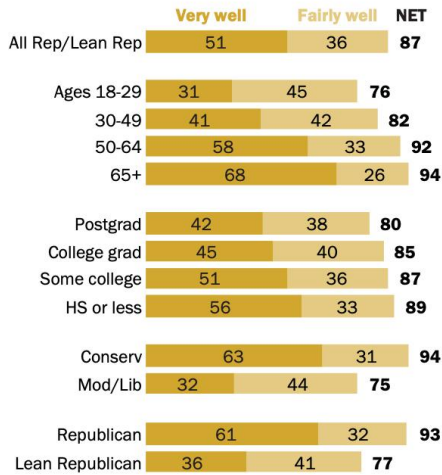


Note: Percentages are global medians based on 37 countries.
Source: Spring 2017 Global Attitudes Survey.

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Older Republicans especially likely to see Trump as fighting for their beliefs

Among **Republicans and Republican leaners**, % who say the phrase 'fights for what I believe in' describes Trump ...



Note: Based on Republicans and Republican-leaning independents.
Source: Survey of U.S. adults conducted Feb. 4-15, 2020.

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Output: Markdown

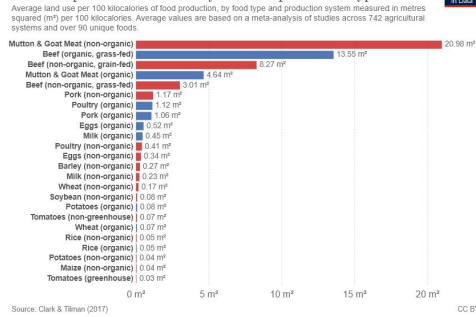
Entity	Value
Caring about ordinary people	23
Well-qualified to be president	26
Charismatic	39
A strong leader	55
Dangerous	62
Intolerant	65
Arrogant	75

Entity	Very well	Fairly well	NET
Lean Republican	36	41	77
Republican	61	32	93
Mod/Lib	32	44	75
Conserv	63	31	94
HS or less	56	33	89
Some college	51	36	87
College grad	45	40	85
Postgrad	42	38	80
65+	68	26	94
50-64	58	33	92
30-49	41	42	82
Ages 18-29	31	45	76
All Rep/Lean Rep	51	36	87

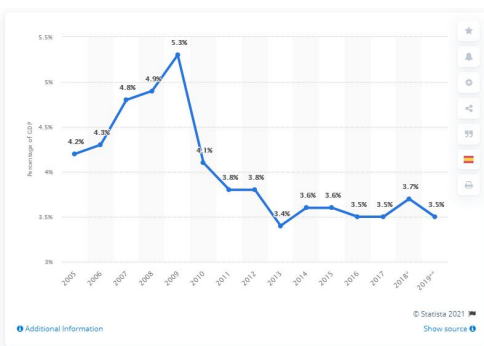
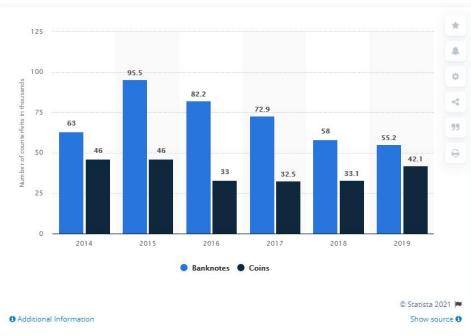
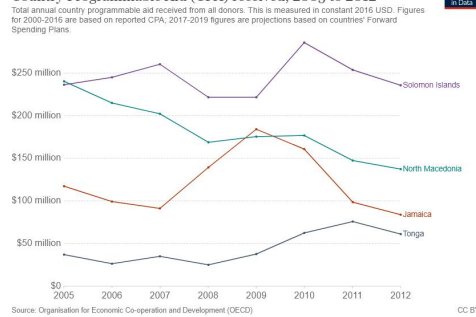
Figure 8. Output of Chimera-8B on Chart Structural Extraction.

Input: Visual Chart

Land use per 100 kilocalories by food and production type



Country Programmable Aid (CPA) received, 2005 to 2012



Output: Markdown

Country	Land use per 100 kilocalories by food and production type
Mutton & Goat Meat (non-organic)	20.98 m²
Beef (organic, grass-fed)	13.55 m²
Beef (non-organic, grain-fed)	8.27 m²
Mutton & Goat Meat (organic)	4.64 m²
Beef (non-organic, grass-fed)	3.01 m²
Pork (non-organic)	1.17 m²
Poultry (organic)	1.12 m²
Pork (organic)	1.06 m²
Eggs (organic)	0.52 m²
Milk (organic)	0.45 m²
Poultry (non-organic)	0.41 m²
Eggs (non-organic)	0.34 m²
Barley (non-organic)	0.27 m²
Milk (non-organic)	0.23 m²
Wheat (non-organic)	0.17 m²
Soybean (non-organic)	0.08 m²
Potatoes (organic)	0.08 m²
Tomatoes (non-greenhouse)	0.07 m²
Wheat (organic)	0.07 m²
Rice (non-organic)	0.05 m²
Rice (organic)	0.05 m²
Potatoes (non-organic)	0.04 m²
Maize (non-organic)	0.04 m²
Tomatoes (greenhouse)	0.03 m²

Entity	2005	2006	2007	2008	2009	2010	2011	2012
Solomon Islands	239.0	246.0	265.0	223.0	223.0	286.0	256.0	237.0
North Macedonia	239.0	214.0	202.0	171.0	179.0	179.0	148.0	138.0
Jamaica	119.0	99.0	93.0	141.0	186.0	163.0	99.0	86.0
Tonga	38.0	28.0	38.0	26.0	37.0	63.0	77.0	62.0

Characteristic	Banknotes	Coins
2019	55.2	42.1
2018	58.0	33.1
2017	72.9	32.5
2016	82.2	33.0
2015	95.5	46.0
2014	63.0	46.0

Characteristic	Percentage of GDP
2019**	3.5%
2018*	3.7%
2017	3.5%
2016	3.5%
2015	3.6%
2014	3.6%
2013	3.4%
2012	3.8%
2011	3.8%
2010	4.1%
2009	5.3%
2008	4.9%
2007	4.8%
2006	4.3%
2005	4.2%

Figure 9. Output of Chimera-8B on Chart Structural Extraction.

Input: Document Page

HYDROLOGICAL PROCESSES
Hydrol. Process. **16**, 3131–3150 (2002)
Published online 7 August 2002 in Wiley InterScience (www.interscience.wiley.com). DOI: 10.1002/hyp.1092

Daily streamflow modelling and assessment based on the curve-number technique

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² Division of Biological Resources and Material Engineering, Seoul National University, Suwon, Kyonggi 441-744, South Korea

Abstract:

A cell-based long-term hydrological model (CELTHYM) that can be integrated with a geographical information system (GIS) was developed to predict continuous stream flow from small agricultural watersheds. The CELTHYM uses a cell-by-cell soil moisture balance approach. For surface runoff estimation, the curve number technique considering soil moisture on a daily basis was used, and release rate was used to estimate baseflow. Evapotranspiration was computed using the FAO modified Penman equation that considered land-use-based crop coefficients, soil moisture and the influence of topography on radiation. A rice paddy field water budget model was also adapted for the specific application of the model to East Asia. Model sensitivity analysis was conducted to obtain operational information about the model calibration parameters. The CELTHYM was calibrated and verified with measured runoff data from the WS#1 and WS#3 watersheds of the Seoul National University, Department of Agricultural Engineering, in Hwasong County, Kyonggi Province, South Korea. The WS#1 watershed is comprised of about 35.4% rice paddy fields and 42.3% forest, whereas the WS#3 watershed is about 85.0% forest and 11.5% rice paddy fields. The CELTHYM was calibrated for the parameter release rate, K , and soil moisture storage coefficient, STC, and results were compared with the measured runoff data for 1986. The validation results for WS#1 considering all daily stream flow were poor with R^2 , E^2 and RMSE having values of 0.40, -6.63 and 9.69 (mm), respectively, but validation results for days without rainfall were statistically significant ($E^2 = 0.66$). Results for WS#3 showed good agreement with observed data for all days, and R^2 , E^2 and RMSE were 0.92, 0.91 and 2.23 (mm), respectively, suggesting potential for CELTHYM application to other watersheds. The direct runoff and water balance components for watershed WS#1 with significant areas of paddy fields did not perform well, suggesting that additional study of these components is needed. Copyright © 2002 John Wiley & Sons, Ltd.

KEY WORDS watershed modelling; GIS; soil moisture balance; grid-based modelling; paddy field water balance; model calibration; sensitivity analysis; model assessment

INTRODUCTION

Water resources development and watershed management require an understanding of hydrological variations owing to changes in watershed characteristics over long-term periods (Bhaduri *et al.*, 2000), and spatial variability of watershed characteristics that affect hydrological phenomenon also must be evaluated in heterogeneous land-use watersheds. However, hydrological model operations that reflect long-term watershed changes often have limitations owing to difficulties obtaining measured hydrological data and data quantifying land use and soil characteristics. Therefore, simulation of stream flow on a daily basis using a long-term hydrological model that is simple to operate with readily available data is needed.

Continuous models, also called long-term hydrological models or continuous stream flow models, typically are focused on estimating water yield from a watershed. Owing to complications and difficulties related to data preparation and operation, however, these hydrological models largely have been used for daily, 10-day and

11 | One View, London Borough of Barking and Dagenham, London, UK

Gaining staff support is also key to ensuring the tool's continued improvement. Addressing the challenge of data linkage specifically, Katy Brown said: "We are doing more to communicate to users why, in some instances, they might not be able to view a record or why it is matched incorrectly (i.e., conflicting information in different case management systems), which is not the fault of One View directly. At a high level, this is to try and lift staff understanding of the process and to maintain trust/understanding in the system."

- **Clear feedback mechanism:** It is important to the project team that improvements to One View are driven by frontline staff feedback. Jill Gallagher described how this works in practice: "Frontline staff give us some great narratives around what's going on for them and how the needs have changed for the people they're serving. We are able to take that information and then adapt One View to meet that need." This may involve going back to the individual services to address data quality issues: "We have in some areas gone back to services to address some data quality issues and improve things from the point of data entry," said Rhodri Rowlands. This was echoed by Katy Brown: "If staff find information in a case summary they know to be wrong, then we are doing a big push for staff to report it, so we can look at it in the background and amend."

For example, staff feedback has led to increased frequency of data collection. As described in the Key challenges section, data is extracted at different frequencies and can be out of date. Debt information used to be extracted monthly, but LBBDD have increased the frequency of these extracts to weekly based on staff feedback. "We are speaking to staff to identify the priority datasets and do what we can to increase their frequency and mitigate the impact of 'time' on the data presented," said Katy Brown.

- **Information governance (IG):** for the service delivery team, it's been very reassuring to know that there is robust IG. Gill Wilson describes her confidence in the tool's IG compliance: "One View has got a very, very good structure to it that embeds the Data Protection Act, the EU's General Data Protection Regulation, so that you know whenever you go in, you only get access to what is at your level. The formal process around the data protection, but also the formal process around the governance structure with the board and everything, it all feeds into the management board. It's very good. It's very strong."

Output: Markdown

Daily stream flow modelling and assessment based on the curve-number technique

Jin-Yong Choi,¹ Bernard A. Engel¹ and HaWooChung² Department of Agricultural and Biological Engineering,Purdue University,West Lafayette,IN47907-1146, USA, Z Division of Biological Resources and Material Engineering,Soupi National University,Shonvik 41-744, South Korea

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KEYWORDS

Water resources development and watershed management require an understanding of hydraulic variations owing to changes in watershed characteristics over long-term periods (Badur *et al.*, 2000), and spatial variability of weather shaded characteristics that affect hydrological phenomenon also must be evaluated in heterogeneous land-use watersheds. However, hydrological model operations that reflect long-term watershed changes often have limitations owing to difficulties obtaining measured hydrological data and data quantifying land use and soil characteristics. Therefore, multination of stream flow on a daily basis using a long-term hydrological model that is simple to operate with readily available data is needed.

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Figure 10. Output of Chimera on Document Context Extraction.

Output: Markdown

The apparent discrepancy between natural history studies and the results of the present study, the observational study and the prospective cohort on CIS and clinical trials that do not show differences in outcomes among different topographies may be explained by our observational study, which demonstrates that, overall, patients with ON may have a better outcome because, as a group, they have more chances for normal baseline MRI results than patients with other CISs. Differential diagnosis in patients with subacute visual loss is comprehensible, and other causes that mimic inflammatory-demyelinating ON are difficult to identify. In our study, the results of the baseline and follow-up MRI results, his or her prognosis does not differ from that of other patients with different CISs. MRI at baseline, not CIS topography, appears to be the crucial issue at MS presentation.

1. Confavreux C, Vukusic S, Adeleine P. Early clinical predictors and progression of irreversible disability in multiple sclerosis: an amnesic process. *Brain* 2003;126:770-782.
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The apparent discrepancy between natural history studies that claimed that ON has a better outcome and the prospective cohort on CSIs and clinical trials that do not show differences in outcomes among different topographies may be explained by our observational study, which demonstrates that overall, patients with ON may have a better outcome because, as a group, they have more chances for normal baseline MRI results than patients with other CSIs. Diffuse cortical lesions in patients with subvisual loss is comprehensible, and other causes that mimic inflammatory/demyelinating ON are difficult to identify. Nevertheless, if a patient with ON has abnormal baseline MRI results, his or her prognosis does not differ from that of other patients with different CSIs. MRI at baseline, not CIS topology, appears to be the crucial issue at MS presentation.

1. Confavreux C, Vukobratovic S, Ales J, P. Early clinical predictors and progression of irreversible disability in multiple sclerosis: an antimetabolite process. *Bank* 2003;126:770-782.
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Figure 11. Output of Chimera on Document Context Extraction.

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Y-Chromosome Variation in Southern African Khoe-San Populations Based on Whole-Genome Sequences

Thijssen Naidoo^{1,2,3,4,5}, Jingzi Xu^{1,6}, Mario Vicente¹, Helena Malmström^{1,5}, Himla Soodyall^{7,8},
Matthias Jakobsson^{1,3,5}, and Carina M. Schliebusch^{1,3,5,*}

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⁶Division of Human Genetics, School of Pathology, Faculty of Health Sciences, University of the Witwatersrand, Johannesburg, South Africa

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Data deposition: The complete Y-chromosome sequences were deposited on the European Genome Phenome Archive (<https://www.ebi.ac.uk/egp/>), accession number EGAS00001004459, and are available for research use under controlled access policies.

Abstract

Although the human Y chromosome has effectively shown utility in uncovering facets of human evolution and population histories, the ascertainment bias present in early Y-chromosome variant data sets limited the accuracy of diversity and TMRCA estimates obtained from them. The advent of next-generation sequencing, however, has removed this bias and allowed for the discovery of thousands of new variants for use in improving the Y-chromosome phylogeny and computing estimates that are more accurate. Here, we describe the high-coverage sequencing of the whole Y-chromosome in a data set of 19 male Khoe-San individuals in comparison with existing whole Y-chromosome sequence data. Due to the increased resolution, we potentially resolve the source of haplogroup B-P70 in the Khoe-San, and reconcile recently published haplogroup A-M51 data with the most recent version of the ISOGG Y-chromosome phylogeny. Our results also improve the positioning of tentatively placed new branches of the ISOGG Y-chromosome phylogeny. The distribution of major Y-chromosome haplogroups in the Khoe-San and other African groups coincide with the emerging picture of African demographic history, with E-M2 linked to the agriculturalist Bantu expansion, E-M35 linked to pastoralist eastern African migrations, B-M112 linked to earlier east-south gene flow, A-M14 linked to shared ancestry with central African rainforest hunter-gatherers, and A-M51 potentially unique to the Khoe-San.

Key words: Y chromosome, next-generation sequencing, haplogroups, Khoe-San, southern Africa.

Introduction

The male-specific portion of the Y chromosome (MSY) has long been regarded as an effective tool in the study of human evolutionary history (Underhill and Kivisild 2007). It has proved useful mainly due to a lack of recombination along its length, making it the longest

haplotypic block in the human genome (Sozarkari et al. 2012), and its paternal mode of inheritance. The transmission of an intact haplotype from father to son, changing only through mutation, preserves a simpler record of its history and allows us to study the male contribution to the shaping of humanity.

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Attitudes towards Rainwater Harvesting

Roofwater Harvesting

Attitudes towards Rainwater Harvesting

Annex 1: Letter and questionnaire to donor organisations

Dear water professional,

I'm writing this letter in the hope that you may be able to assist us in our research programme, called „Domestic Roofwater Harvesting in the Humid Tropics“. It is a 3-year programme to generate reliable information for water policy planners, water supply professionals and ultimately householders. The programme just started and is funded by the EU. It involves 4 partners from India, Sri Lanka, England and Germany. Links are being developed with practitioners in Central America and East Africa. The programme will examine literature and practice from many parts of the world, but it is expected that those from humid tropical areas will be of most use, since Domestic Rainwater Harvesting technology and economics are dominated by factors like climate and culture.

In the view of water resources getting scarce, it is becoming obvious that we should use every available water resource as e.g. rainfall. Rainwater harvesting has been and is successfully practised for millennia around the Mediterranean as a supplementary source of water or the only one available. In many countries of Asia, Africa and Latin America, it is currently newly introduced or its use widened. One of several components of the programme is to define the information needs of organisations active in the water sector. We would therefore like to ask you about your funding policy for the water sector and how it is implemented in the different countries. We will compile our research findings (we just have started) and will make them available to you if you wish so. They will also be available on a web site, which will soon be established. Please feel free to contact me for any additional information you might want to get.

Thanking you in advance for your time and efforts invested
- also in the name of the other partners involved

Hans Hartung

Responsible Task Manager for Task B: Institutional Values and Decision Making

Output: Markdown

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[†]These authors contributed equally to this work. Data deposition: The complete Y-chromosome sequences were deposited on the European Genome Phenome Archive (<https://www.ebi.ac.uk/egp/>), accession number EGAS00001004459, and are available for research use under controlled access policies.

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Figure 12. Output of Chimera on Document Context Extraction.

Input: Document Page

areas. In the first bid book, however, it was noticeable that what the whole of Germany associated with Chemnitz during the application phase, i.e. from August 2018 onward, hardly played a role in the application, namely the riots of right-wing extremist and populists in August 2018. Nevertheless, Chemnitz presented an extensive application overall and demonstrated a good network of cultural actors in the city and the region, so that the selection panel recommended the city for the next round.

Dresden (dropped out of competition in 2019)

Under the motto "New Home Dresden" ("Neue Heimat Dresden"), the Saxon capital presented concepts and projects in its bid book I that were intended to counteract the social division of the city. In numerous artistic actions, some of which were supported by the project office with microcredits, as well as in panel discussions and participative projects such as a postcard campaign, the concept of "home" ("Heimat") was played with in order to find new narratives for the city. This was not only about an open attitude towards immigrants, but also about major present and future tasks that fundamentally change people's lives, such as digitisation. Although Dresden's approach was highly innovative, relevant and authentic, various influential opinion makers such as the German Cultural Council (Deutscher Kulturrat 2019: 14) and the Süddeutsche Zeitung (Heidtmann/ Nimb 2019: online) contrasted the two Saxon competing cities in a questionably simplistic way: poor Chemnitz, which is preceded by a problematic reputation, versus dazzling, privileged Dresden. Whether Dresden really dropped out of the competition at an early stage against the background of this simplistic black-and-white portrayal in the media cannot be proven. In any case, many could not understand the selection panel's assessment (see the Expert Panel's Report of the Pre-Selection Stage / Selection of the European Capital of Culture 2025 in Germany, https://ec.europa.eu/programmes/creative-europe/actions/capitals-culture_en, and a comment at Jacobsen 2020: 22).

Gera (dropped out of competition in 2019)

The Thuringian city with only 95,000 inhabitants dared to go off the beaten track. An association of committed citizens initiated the application and pushed it forward years before the city made a - relatively modest - contribution to the application. Their bid book I presented the historical heritage, and perhaps this common understanding of the interpretation of the past was already a benefit to the city society. However, too few feasible visions of the future were presented, which could be achieved with the ECoc title. All in all, Gera had probably shown too little commitment in the application process to meet the demanding and far-reaching requirements, especially with regard to the other top-class competing cities.

Hanover

The capital of Lower Saxony entered the competition relatively late and then immediately faced special cultural policy challenges: a new mayor was elected during this period, and there was also a change in the cultural administration after a few squabbles, when the former head of the culture department was brought to court. Nevertheless, Hanover managed to submit an exceptionally artistic bid book I, which was awarded a prize not least for its design. Positive aspects of the content were its distinct European dimension and a professionally positioned management, which led to the expectation that the ECoc programme would be very feasible in the fair and Expo city.

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INTRODUCTION.

Consider a general-relativistic system — a compact core immersed in a steadily accreting self-gravitating gas. The gravitational binding energy of the infalling gas can be converted to radiation. Assume that an external distant observer can measure total luminosity, asymptotic temperature and redshifts of the radiation. Let be known: the total (asymptotic) mass of the system and the physics of the mixture of gas and radiation. Then it would be natural to ask: what mass is within the compact body? Alternatively, the mass of the core would be known and the total mass would require determination.

The main goal of this paper is the numerical investigation of this problem for stationary flows.

We assume spherical symmetry and adopt thin gas approximation in the transport equation. It is already known from studies of newtonian radiation hydrodynamics [1] — [3] that supersonic flows are generically not fixed by total luminosity, asymptotic temperature and redshift. To each set of such data there can correspond two solutions with different gas abundances. Changing luminosity one obtains two curves, on the luminosity-(gas abundance) diagram, that originate at a bifurcation point. This point is unique, for given boundary data. General-relativistic supersonic flows with small redshifts are similar to newtonian ones in that they also branch from a bifurcation transonic flow. In the case of high redshifts supersonic general-relativistic flows can be absent. A similar picture appears in transonic flows of perfect gases, newtonian or general-relativistic, without radiation. In this case boundary data can consist of the mass accretion rate and the asymptotic speed of sound [4] and the only unique solution — a branching point — corresponds to the maximal accretion.

Accretion systems with subsonic flows are not determined by the data described hitherto. One needs additional information, for instance the asymptotic gas density, in order to specify the solution completely. We discover, however, an interesting fact valid in the newtonian case and in the low-redshift regime of general relativity: transonic flows encompass, on the luminosity-(gas abundance) diagram, the set filled with subsonic flows. Therefore the two transonic branches provide estimates of the mass abundance of corresponding subsonic solutions. In particular, numerical analysis suggests that the most luminous flow is supersonic. This picture is valid in the newtonian level and also in the general-relativistic case, for small redshifts. If redshifts are large, then the boundary of the set of subsonic solutions may consist of transonic or subsonic flows, but it is remarkable that the shape of the set of subsonic solutions is only weakly dependent on redshifts. In

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